# Package 'cxr' 

October 26, 2023

## Type Package

Title A Toolbox for Modelling Species Coexistence in R
Version 1.1.1
Description Recent developments in modern coexistence theory have advanced our understanding on how species are able to persist and co-occur with other species at varying abundances. However, applying this mathematical framework to empirical data is still challenging, precluding a larger adoption of the theoretical tools developed by empiricists. This package provides a complete toolbox for modelling interaction effects between species, and calculate fitness and niche differences. The functions are flexible, may accept covariates, and different fitting algorithms can be used. A full description of the underlying methods is available in García-Callejas, D., Godoy, O., and Bartomeus, I. (2020) [doi:10.1111/2041-210X.13443](doi:10.1111/2041-210X.13443). Furthermore, the package provides a series of functions to calculate dynamics for stage-structured populations across sites.
License MIT + file LICENSE
URL https://github.com/RadicalCommEcol/cxr
BugReports https://github.com/RadicalCommEcol/cxr/issues
Depends R (>= 3.5)
Imports Matrix, mvtnorm, optimx, stats
Suggests BB, DEoptimR, dfoptim, dplyr, GenSA, ggplot2, knitr, magrittr, minqa, nloptr, rmarkdown, stringr, testhat (>= $0.8 .0)$, tidyr, ucminf
VignetteBuilder knitr
Encoding UTF-8
RoxygenNote 7.2.3
NeedsCompilation no
Author David Garcia-Callejas [aut, cre] ([https://orcid.org/0000-0001-6982-476X](https://orcid.org/0000-0001-6982-476X)), Ignasi Bartomeus [aut] ([https://orcid.org/0000-0001-7893-4389](https://orcid.org/0000-0001-7893-4389)),

Oscar Godoy [aut] ([https://orcid.org/0000-0003-4988-6626](https://orcid.org/0000-0003-4988-6626)), Maxime Lancelot [ctb], Maria Paniw [ctb]<br>Maintainer David Garcia-Callejas [david.garcia.callejas@gmail.com](mailto:david.garcia.callejas@gmail.com)<br>Repository CRAN

Date/Publication 2023-10-26 21:30:02 UTC

## $R$ topics documented:

abundance ..... 4
abundance_projection ..... 4
avg_fitness_diff ..... 5
BH_er_lambdacov_global_effectcov_global_responsecov_global ..... 7
BH_er_lambdacov_none_effectcov_none_responsecov_none ..... 8
BH_pm_alpha_global_lambdacov_none_alphacov_none ..... 9
BH_pm_alpha_none_lambdacov_none_alphacov_none ..... 9
BH_pm_alpha_pairwise_lambdacov_global_alphacov_global ..... 10
BH_pm_alpha_pairwise_lambdacov_global_alphacov_pairwise ..... 11
BH_pm_alpha_pairwise_lambdacov_none_alphacov_none ..... 12
BH_project_alpha_global_lambdacov_none_alphacov_none ..... 13
BH_project_alpha_none_lambdacov_none_alphacov_none ..... 14
BH_project_alpha_pairwise_lambdacov_global_alphacov_global ..... 14
BH_project_alpha_pairwise_lambdacov_global_alphacov_pairwise ..... 15
BH_project_alpha_pairwise_lambdacov_none_alphacov_none ..... 16
build_param ..... 17
calculate_densities ..... 18
competitive_ability ..... 18
cxr_er_bootstrap ..... 20
cxr_er_fit ..... 21
cxr_generate_test_data ..... 24
cxr_pm_bootstrap ..... 27
cxr_pm_fit ..... 28
cxr_pm_multifit ..... 30
densities_to_df ..... 33
fill_demography_matrix ..... 34
fill_dispersal_matrix ..... 34
fill_transition_matrix ..... 35
fitness_ratio ..... 36
generate_vital_rate_coefs ..... 36
glm_example_coefs ..... 38
LV_er_lambdacov_global_effectcov_global_responsecov_global ..... 38
LV_er_lambdacov_none_effectcov_none_responsecov_none ..... 39
LV_pm_alpha_global_lambdacov_none_alphacov_none ..... 40
LV_pm_alpha_none_lambdacov_none_alphacov_none ..... 40
LV_pm_alpha_pairwise_lambdacov_global_alphacov_global ..... 41
LV_pm_alpha_pairwise_lambdacov_global_alphacov_pairwise ..... 42
LV_pm_alpha_pairwise_lambdacov_none_alphacov_none ..... 43
LV_project_alpha_global_lambdacov_none_alphacov_none ..... 44
LV_project_alpha_none_lambdacov_none_alphacov_none ..... 45
LV_project_alpha_pairwise_lambdacov_global_alphacov_global ..... 45
LV_project_alpha_pairwise_lambdacov_global_alphacov_pairwise ..... 46
LV_project_alpha_pairwise_lambdacov_none_alphacov_none ..... 47
LW_er_lambdacov_global_effectcov_global_responsecov_global ..... 48
LW_er_lambdacov_none_effectcov_none_responsecov_none ..... 49
LW_pm_alpha_global_lambdacov_none_alphacov_none ..... 49
LW_pm_alpha_none_lambdacov_none_alphacov_none ..... 50
LW_pm_alpha_pairwise_lambdacov_global_alphacov_global ..... 51
LW_pm_alpha_pairwise_lambdacov_global_alphacov_pairwise ..... 52
LW_pm_alpha_pairwise_lambdacov_none_alphacov_none ..... 53
LW_project_alpha_global_lambdacov_none_alphacov_none ..... 54
LW_project_alpha_none_lambdacov_none_alphacov_none ..... 54
LW_project_alpha_pairwise_lambdacov_global_alphacov_global ..... 55
LW_project_alpha_pairwise_lambdacov_global_alphacov_pairwise ..... 56
LW_project_alpha_pairwise_lambdacov_none_alphacov_none ..... 57
metapopulation_example_param ..... 58
neigh_list ..... 58
niche_overlap ..... 59
RK_er_lambdacov_global_effectcov_global_responsecov_global ..... 60
RK_er_lambdacov_none_effectcov_none_responsecov_none ..... 61
RK_pm_alpha_global_lambdacov_none_alphacov_none ..... 62
RK_pm_alpha_none_lambdacov_none_alphacov_none ..... 62
RK_pm_alpha_pairwise_lambdacov_global_alphacov_global ..... 63
RK_pm_alpha_pairwise_lambdacov_global_alphacov_pairwise ..... 64
RK_pm_alpha_pairwise_lambdacov_none_alphacov_none ..... 65
RK_project_alpha_global_lambdacov_none_alphacov_none ..... 66
RK_project_alpha_none_lambdacov_none_alphacov_none ..... 67
RK_project_alpha_pairwise_lambdacov_global_alphacov_global ..... 67
RK_project_alpha_pairwise_lambdacov_global_alphacov_pairwise ..... 68
RK_project_alpha_pairwise_lambdacov_none_alphacov_none ..... 69
salinity_list ..... 70
spatial_sampling ..... 70
species_fitness ..... 71
species_rates ..... 72
summary.cxr_er_fit ..... 73
summary.cxr_pm_fit ..... 73
summary.cxr_pm_multifit ..... 74
vec_permutation_matrices ..... 74
vital_rate ..... 75
Index ..... 76

Abundance measurements

## Description

A dataset containing abundances for each plant species, where each species was sampled at its developmental peak.

- plot: one of 9 plots of the study area
- subplot: one of $361 \times 1 \mathrm{~m}$ subplots of each plot
- species: plant species
- individuals: number of individuals observed


## Usage

data(abundance)

## Format

A data frame with 5184 rows and 4 variables

## Note

For details, see Lanuza et al. 2018 Ecology Letters.
abundance_projection Project abundances from population dynamics models

## Description

The function projects a number of steps of a time-discrete model, with model parameters taken from a 'cxr_pm_multifit' object or as function arguments.

## Usage

abundance_projection(
cxr_fit = NULL,
model_family = NULL,
alpha_form = NULL, lambda_cov_form = NULL, alpha_cov_form = NULL, lambda = NULL, alpha_matrix = NULL, lambda_cov = NULL, alpha_cov = NULL,

```
    covariates = NULL,
    timesteps = 2,
    initial_abundances = 0
```

)

## Arguments

```
cxr_fit object of type 'cxr_pm_multifit'. If this is not specified, all parameters below
are needed.
model_family acronym for model family. Included by default in 'cxr' are 'BH' (Beverton-
    Holt), 'RK' (Ricker), 'LW' (Law-Watkinson), 'LV' (Lotka-Volterra).
    alpha_form character, either "none","global", or "pairwise".
lambda_cov_form
                            character, either "none" or "global".
alpha_cov_form character, either "none","global", or "pairwise".
lambda named vector with lambda values for all taxa to be projected.
alpha_matrix square matrix with taxa names in rows and columns.
lambda_cov optional named matrix with covariates in columns and taxa in rows, representing
    the effect of each covariate on the lambda parameter of each taxa.
alpha_cov optional list. Each element of the named list represents the effects of a covariate
    over alpha values. Thus, each list element contains a square matrix of the same
    dimensions as 'alpha_matrix', as returned from the function 'cxr_pm_fit'. Note
    that for alpha_cov_form = "global", all columns in this matrix are the same, as
    there is a single value per species.
    covariates matrix or dataframe with covariates in columns and timesteps in rows.
    timesteps number of timesteps to project.
    initial_abundances
    named vector of initial abundances for all taxa.
```


## Value

named matrix with projected abundance values for each taxa at each timestep.

```
avg_fitness_diff Average fitness differences
```


## Description

computes the average fitness differences among two or more species according to the formulation of the MCT (Chesson 2012, Godoy and Levine 2014), and according to the structural approach (Saavedra et al. 2017). For the MCT version, the average fitness ratio is decomposed in a 'demographic ratio' and a 'competitive response ratio', the product of which is the average fitness ratio (Godoy and Levine 2014). This formulation is only valid for competitive interaction coefficients (i.e. positive alpha values in the interaction matrix). The structural analog can be computed for any interaction matrix, on the other hand. Note that the 'demographic ratio' is model-specific (Hart et al. 2018).

## Usage

avg_fitness_diff(
cxr_multifit = NULL,
cxr_sp1 = NULL,
cxr_sp2 = NULL,
pair_lambdas = NULL,
pair_matrix = NULL,
model_family = NULL
)

## Arguments

cxr_multifit
cxr_sp1 cxr_pm_fit object giving the parameters from the first species.
cxr_sp2 cxr_pm_fit object giving the parameters from the second species.
pair_lambdas numeric vector of length 2 giving lambda values for the two species.
pair_matrix $2 \times 2$ matrix with intra and interspecific interaction coefficients between the two species.
model_family model family for which to calculate fitness differences.

## Details

This function, as in niche_overlap and competitive_ability, accepts three different parameterizations:

- A cxr_pm_multifit object, from which average fitness differences will be computed across all species pairs.
- two cxr_pm_fit objects, one for each species.
- explicit lambda and alpha values, as well as the model family from which these parameters were obtained.

If using the third parameterization, the function will try to find a model-specific function for obtaining the demographic ratio, by looking at the 'model_family' parameter. If this specific function is not found, it will resort to the standard Lotka-Volterra formulation (lambda in the numerator term). Overall, we strongly suggest that you use the standard formulation ONLY if you are completely confident that your custom model is consistent with it. Otherwise, you should include your own formulation of the demographic ratio (see vignette 4).

## Value

data frame with variable number of rows, and columns specifying the different components of the MCT average fitness ratio, as well as its structural analog. The average fitness ratio informs quantitatively about the better competitor. If the ratio is $<1, \mathrm{sp} 2$ is the better competitor; if $=1$, both species are equivalent competitors, if $>1, \mathrm{sp} 1$ is the better competitor.

## Examples

```
avg_fitness_diff(pair_lambdas = runif(2,1,10),
    pair_matrix = matrix(runif(4,0,1), nrow = 2),
    model_family = "BH")
```

```
BH_er_lambdacov_global_effectcov_global_responsecov_global
    Effect response Beverton-Holt model with covariate effects on lambda,
    effect, and response
```


## Description

The function for calculating fecundity given effect and response values is taken from Godoy et al. (2014). Note that, as e and $r$ are not pair-specific, all species parameters are fit in the same function.

## Usage

BH_er_lambdacov_global_effectcov_global_responsecov_global(
par,
fitness,
target,
density,
covariates,
fixed_parameters
)

## Arguments

| par | 1d vector with initial parameters in the order: lambda,lambda_cov, effect,effect_cov,response,response_co |
| :--- | :--- |
| fitness | $1 d$ vector with fitness observations |
| target | matrix with species in rows, observations in columns. Value is 1 if a species is <br> focal for a given observation, 0 otherwise. |
| density | matrix with species in rows, observations in columns. Value is density of each <br> sp as neighbour for each observation. |
| covariates | numeric dataframe or matrix with observations in rows and covariates in columns. <br> Each cell is the value of a covariate in a given observation |
| fixed_parameters |  | optional list specifying values of fixed parameters, with components "lambda","lambda_cov","effect","eff "response","response_cov".

## Value

log-likelihood value

```
BH_er_lambdacov_none_effectcov_none_responsecov_none
    Effect response model without covariate effects
```


## Description

The function for calculating fecundity given effect and response values is taken from Godoy et al. (2014). Note that, as e and $r$ are not pair-specific, all species parameters are fit in the same function.

## Usage

```
    BH_er_lambdacov_none_effectcov_none_responsecov_none(
        par,
        fitness,
        target,
        density,
        covariates,
        fixed_parameters
    )
```


## Arguments

par 1d vector with initial parameters in the order: lambda,effect,response,sigma.
fitness 1d vector with fitness observations.
target matrix with species in rows, observations in columns. Value is 1 if a species is focal for a given observation, 0 otherwise.
density matrix with species in rows, observations in columns. Value is density of each sp as neighbour for each observation.
covariates included for compatibility, not used in this model.
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","effect","response".

## Value

log-likelihood value

BH_pm_alpha_global_lambdacov_none_alphacov_none
Beverton-Holt model with a global alpha and no covariate effects

## Description

Beverton-Holt model with a global alpha and no covariate effects

## Usage

BH_pm_alpha_global_lambdacov_none_alphacov_none(
par,
fitness,
neigh_intra_matrix = NULL,
neigh_inter_matrix,
covariates,
fixed_parameters
)

## Arguments

par 1d vector of initial parameters: lambda, alpha, and sigma.
fitness $\quad 1 d$ vector of fitness observations, in $\log$ scale.
neigh_intra_matrix
included for compatibility, not used in this model.
neigh_inter_matrix
matrix of arbitrary columns, number of neighbours for each observation. As in this model there is a single alpha argument, do not distinguish neighbour identity
covariates included for compatibility, not used in this model.
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","alpha_inter".

## Value

log-likelihood value

```
BH_pm_alpha_none_lambdacov_none_alphacov_none
```

Beverton-Holt model with no alphas and no covariate effects

## Description

Beverton-Holt model with no alphas and no covariate effects

## Usage

```
BH_pm_alpha_none_lambdacov_none_alphacov_none(
    par,
    fitness,
    neigh_intra_matrix = NULL,
    neigh_inter_matrix,
    covariates,
    fixed_parameters
)
```


## Arguments

par | 1d vector of initial parameters: lambda and sigma |
| :--- |
| fitness |
| neigh_intra_matrix |
| included for compatibility, not used in this model. |

| neigh_inter_matrix |
| ---: | :--- |
| included for compatibility, not used in this model. |


| covariates | included for compatibility, not used in this model |
| :--- | :--- |
| fixed_parameters |  | included for compatibility, not used in this model

## Value

log-likelihood value

```
BH_pm_alpha_pairwise_lambdacov_global_alphacov_global
    Beverton-Holt model with pairwise alphas and global covariate effects
    on lambda and alpha
```


## Description

Beverton-Holt model with pairwise alphas and global covariate effects on lambda and alpha

## Usage

```
BH_pm_alpha_pairwise_lambdacov_global_alphacov_global(
    par,
    fitness,
    neigh_intra_matrix = NULL,
    neigh_inter_matrix,
    covariates,
    fixed_parameters
)
```


## Arguments

par
1d vector of initial parameters: lambda, lambda_cov, alpha, alpha_cov, and sigma
fitness $\quad 1 d$ vector of fitness observations, in $\log$ scale
neigh_intra_matrix
optional matrix of one column, number of intraspecific neighbours for each observation
neigh_inter_matrix matrix of arbitrary columns, number of interspecific neighbours for each observation
covariates optional matrix with observations in rows and covariates in columns. Each cell is the value of a covariate in a given observation
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","alpha_intra","alpha_inter"

## Value

log-likelihood value

```
BH_pm_alpha_pairwise_lambdacov_global_alphacov_pairwise
    Beverton-Holt model with pairwise alphas, covariate effects on
    lambda, and pairwise covariate effects on alpha
```


## Description

Beverton-Holt model with pairwise alphas, covariate effects on lambda, and pairwise covariate effects on alpha

## Usage

```
BH_pm_alpha_pairwise_lambdacov_global_alphacov_pairwise(
    par,
        fitness,
        neigh_intra_matrix = NULL,
        neigh_inter_matrix,
        covariates,
        fixed_parameters
)
```


## Arguments

par
1d vector of initial parameters: lambda, lambda_cov, alpha, alpha_cov, and sigma
fitness $\quad 1 d$ vector of fitness observations, in log scale

```
neigh_intra_matrix
```

                    optional matrix of one column, number of intraspecific neighbours for each ob-
                        servation
    neigh_inter_matrix
matrix of arbitrary columns, number of interspecific neighbours for each obser-
vation
covariates optional matrix with observations in rows and covariates in columns. Each cell
is the value of a covariate in a given observation
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","alpha_intra","alpha_inter"

## Value

log-likelihood value

```
BH_pm_alpha_pairwise_lambdacov_none_alphacov_none
    Beverton-Holt model with pairwise alphas and no covariate effects
```


## Description

Beverton-Holt model with pairwise alphas and no covariate effects

## Usage

BH_pm_alpha_pairwise_lambdacov_none_alphacov_none(
par,
fitness,
neigh_intra_matrix = NULL,
neigh_inter_matrix,
covariates,
fixed_parameters
)

## Arguments

par 1d vector of initial parameters: 'lambda', 'alpha_intra' (optional), 'alpha_inter', and 'sigma'
fitness $\quad 1 d$ vector of fitness observations, in log scale
neigh_intra_matrix
optional matrix of one column, number of intraspecific neighbours for each observation
neigh_inter_matrix
matrix of arbitrary columns, number of interspecific neighbours for each observation
covariates included for compatibility, not used in this model
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","alpha_intra","alpha_inter"

## Value

log-likelihood value

BH_project_alpha_global_lambdacov_none_alphacov_none
Beverton-Holt model for projecting abundances, with a global alpha and no covariate effects

## Description

Beverton-Holt model for projecting abundances, with a global alpha and no covariate effects

## Usage

```
    BH_project_alpha_global_lambdacov_none_alphacov_none(
        lambda,
        alpha_intra,
        alpha_inter,
        lambda_cov,
        alpha_cov,
        abundance,
        covariates
    )
```


## Arguments

lambda numeric lambda value.
alpha_intra included for compatibility, not used in this model.
alpha_inter single numeric value.
lambda_cov included for compatibility, not used in this model.
alpha_cov included for compatibility, not used in this model.
abundance named numeric vector of abundances in the previous timestep.
covariates included for compatibility, not used in this model.

## Value

numeric abundance projected one timestep

```
BH_project_alpha_none_lambdacov_none_alphacov_none
    Beverton-Holt model for projecting abundances, with no alpha and no
    covariate effects
```


## Description

Beverton-Holt model for projecting abundances, with no alpha and no covariate effects

## Usage

BH_project_alpha_none_lambdacov_none_alphacov_none( lambda, alpha_intra, alpha_inter, lambda_cov, alpha_cov, abundance, covariates )

## Arguments

| lambda | numeric lambda value. |
| :--- | :--- |
| alpha_intra | included for compatibility, not used in this model. |
| alpha_inter | included for compatibility, not used in this model. |
| lambda_cov | included for compatibility, not used in this model. |
| alpha_cov | included for compatibility, not used in this model. |
| abundance | named numeric vector of abundances in the previous timestep. |
| covariates | included for compatibility, not used in this model. |

## Value

numeric abundance projected one timestep

```
BH_project_alpha_pairwise_lambdacov_global_alphacov_global
    Beverton-Holt model for projecting abundances, with specific alpha
    values and global covariate effects on alpha and lambda
```


## Description

Beverton-Holt model for projecting abundances, with specific alpha values and global covariate effects on alpha and lambda

## Usage

```
BH_project_alpha_pairwise_lambdacov_global_alphacov_global(
        lambda,
        alpha_intra,
        alpha_inter,
        lambda_cov,
        alpha_cov,
        abundance,
        covariates
)
```


## Arguments

| lambda | numeric lambda value. |
| :--- | :--- |
| alpha_intra | single numeric value. |
| alpha_inter | numeric vector with interspecific alpha values. |
| lambda_cov | numeric vector with effects of covariates over lambda. |
| alpha_cov | named list of numeric values with effects of each covariate over alpha. |
| abundance | named numeric vector of abundances in the previous timestep. <br> covariates |
|  | matrix with observations in rows and covariates in columns. Each cell is the <br> value of a covariate in a given observation. |

## Value

numeric abundance projected one timestep

BH_project_alpha_pairwise_lambdacov_global_alphacov_pairwise
Beverton-Holt model for projecting abundances, with specific alpha values and global covariate effects on alpha and lambda

## Description

Beverton-Holt model for projecting abundances, with specific alpha values and global covariate effects on alpha and lambda

## Usage

```
BH_project_alpha_pairwise_lambdacov_global_alphacov_pairwise(
    lambda,
    alpha_intra,
    alpha_inter,
    lambda_cov,
    alpha_cov,
    abundance,
    covariates
)
```


## Arguments

| lambda | named numeric lambda value. |
| :--- | :--- |
| alpha_intra | single numeric value. |
| alpha_inter | numeric vector with interspecific alpha values. |
| lambda_cov | numeric vector with effects of covariates over lambda. <br> alpha_cov <br> named list of named numeric vectors with effects of each covariate over alpha <br> values. |
| abundance | named numeric vector of abundances in the previous timestep. <br> covariates |
| matrix with observations in rows and covariates in named columns. Each cell is <br> the value of a covariate in a given observation. |  |

## Value

numeric abundance projected one timestep

```
BH_project_alpha_pairwise_lambdacov_none_alphacov_none
    Beverton-Holt model for projecting abundances, with specific alpha
    values and no covariate effects
```


## Description

Beverton-Holt model for projecting abundances, with specific alpha values and no covariate effects

## Usage

```
BH_project_alpha_pairwise_lambdacov_none_alphacov_none(
    lambda,
    alpha_intra,
    alpha_inter,
    lambda_cov,
    alpha_cov,
    abundance,
    covariates
    )
```


## Arguments

lambda numeric lambda value.
alpha_intra included for compatibility, not used in this model.
alpha_inter single numeric value.
lambda_cov included for compatibility, not used in this model.
alpha_cov included for compatibility, not used in this model.
abundance named numeric vector of abundances in the previous timestep.
covariates included for compatibility, not used in this model.

## Value

numeric abundance projected one timestep
build_param Build param structure

## Description

Builds a nested list for the parameters of a given metapopulation

## Usage

build_param(sp, sites, rates, env, num.params = NULL)

## Arguments

sp character vector with species names
sites character vector with site names
rates character vector, vital rate names
env boolean, whether environment is accounted for
num. params optional, integer giving the number of parameters to account for. If not specified, it will include environment interactions with all species densities. E.g. if 3 sp and env = TRUE, there will be 7 params (intercept +6 betas)

## Value

nested list of the form 'list[[sp]][[site]]'. Each of these elements is a NA matrix with vital rates in rows and expected parameters in columns.

## Examples

```
sp <- c("s1","s2","s3")
sites <- c("sa","sb")
rates <- c("Sj","Sn","Sr","Rn","Rr","D","0")
env <- TRUE
param <- build_param(sp = sp,sites = sites,rates = rates,env = env)
```

calculate_densities Obtain species densities from transition matrices

## Description

Using the vec-permutation approach as defined in: Hunter and Caswell 2005, doi:10.1016/j.ecolmodel.2005.05.002, Ozgul et al. 2009, doi: 10.1086/597225 In particular, it uses the arrangement by patches, and calculates first demography, then dispersal (Table 1 of Hunter and Caswell 2005).

## Usage

calculate_densities(focal.sp, vpm, current.densities)

## Arguments

| focal.sp | integer, focal species |
| :--- | :--- |
| vpm | data structure holding all vector-permutation matrices; see 'vec_permutation_matrices'. <br> If not in an appropriate format, it is likely to fail without warning. |
| current. densities |  |$\quad$| list of length sp, each element is a matrix sites*stages. If not in that format, it is |
| :--- |
|  |
|  |
| likely to fail without warning. |

## Value

matrix of sites $x$ stages, each element is the density of a given life stage (juvenile, non-reproductive adult, reproductive adult) at a given site.

## Description

Computes the competitive ability among two species, as defined by Hart et al. (2018). This metric, as others in MCT, is model-specific; the formulation for a series of Lotka-Volterra-like models is given in table A1 of Hart et al. (2018). We include in cxr by default the formulation for BevertonHolt, Ricker, Law-Watkinson, and Lotka-Volterra families.

```
Usage
    competitive_ability(
        cxr_multifit = NULL,
        cxr_sp1 = NULL,
        cxr_sp2 = NULL,
        lambda = NULL,
        pair_matrix = NULL,
        model_family = NULL
    )
```


## Arguments

| cxr_multifit | cxr_pm_multifit object, with parameters for a series of species. |
| :--- | :--- |
| cxr_sp1 | cxr_pm_fit object giving the parameters from the first species. |
| cxr_sp2 | cxr_pm_fit object giving the parameters from the second species. |
| lambda | numeric lambda value of the focal species. |
| pair_matrix | $2 \times 2$ matrix with intra and interspecific interaction coefficients between the focal <br> and competitor species. |
| model_family | model family for which to calculate competitive ability. |

## Details

The function, as in avg_fitness_diff and niche_overlap, accepts three different parameterizations:

- A cxr_pm_multifit object, from which competitive ability of a focal species relative to a given competitor will be computed across all species pairs.
- two cxr_pm_fit objects, one for a focal species and one for a competitor.
- explicit lambda and alpha values, as well as the model family from which these parameters were obtained.

If the third parameterization is used, the function will try to find a model-specific function for obtaining the competitive ability, by looking at the 'model_family' parameter. If this specific function is not found, it will resort to the standard Lotka-Volterra formulation (lambda - 1 in the numerator term, Hart et al. 2018). Overall, we strongly suggest that you use the standard formulation ONLY if you are completely confident that the model from which you obtained your parameters is consistent with it. Otherwise, you should include your own formulation of competitive ability (see vignette 4).

## Value

data frame with variable number of rows and three columns, specifying taxa identity and the competitive ability of focal species (sp1) relative to the competitor ( sp 2 ).

## Examples

```
competitive_ability(lambda = runif(1, 1, 10),
    pair_matrix = matrix(runif(4,0,1),nrow = 2),
    model_family = "BH")
```


## Description

Computes bootstrap standard errors for a given effect/response function. This function is provided for completeness, but error calculation is integrated in the function cxr_er_fit.

## Usage

```
cxr_er_bootstrap(
    fitness_model,
    optimization_method,
    data,
    covariates,
    init_par,
    lower_bounds,
    upper_bounds,
    fixed_parameters,
    bootstrap_samples
)
```


## Arguments

fitness_model effect/response function, see cxr_er_fit
optimization_method
numerical optimization method.
data either a list of dataframes or a single dataframe. if 'data' is a list, each element is a dataframe with the following columns:

- fitness: fitness metric for each observation
- neighbours: named columns giving the number of neighbours of each column the names of the list elements are taken to be the names of the focal species.
If 'data' is a dataframe, it also needs a 'focal' column. Regardless of the data structure, all focal species need to have the same number of observations (i.e. same number of rows), and the set of neighbour species needs to be the same as the set of focal species, so that the neighbours columns correspond to the names of the list elements or, if 'data' is a dataframe, to the values of the 'focal' column. Future versions will relax this requirement.
covariates a data structure equivalent to 'data', in which each column are the values of a covariate.
init_par initial values for parameters
lower_bounds optional list with single values for "lambda", "effect","response", and optionally "lambda_cov", "effect_cov", "response_cov".
upper_bounds optional list with single values for "lambda", "effect","response", and optionally "lambda_cov", "effect_cov", "response_cov".
fixed_parameters
list with values for fixed parameters, or NULL.
bootstrap_samples
number of bootstrap samples for error calculation. Defaults to 0 , i.e. no error is calculated.


## Value

1d vector, the standard error of each parameter in init_par

```
cxr_er_fit General optimization for effect-response models
```


## Description

Estimates parameters of user-specified models of competitive effects and responses. NOTE: including covariates on competitive effects is still under development, in this version it is suggested not to use that feature.

## Usage

```
cxr_er_fit(
        data,
        model_family = c("BH"),
        covariates = NULL,
        optimization_method = c("Nelder-Mead", "BFGS", "CG", "ucminf", "L-BFGS-B", "nlm",
            "nlminb", "Rcgmin", "Rvmmin", "spg", "bobyqa", "nmkb", "hjkb", "nloptr_CRS2_LM",
                "nloptr_ISRES", "nloptr_DIRECT_L_RAND", "DEoptimR", "GenSA"),
    lambda_cov_form = c("none", "global"),
    effect_cov_form = c("none", "global"),
    response_cov_form = c("none", "global"),
    initial_values \(=1\) ist (lambda \(=1\), effect \(=1\), response \(=1\), lambda_cov \(=0\), effect_cov
            = 0, response_cov = 0),
        lower_bounds = NULL,
        upper_bounds = NULL,
        fixed_terms = NULL,
        bootstrap_samples = 0
    )
```


## Arguments

either a list of dataframes or a single dataframe. if 'data' is a list, each element is a dataframe with the following columns:

- fitness: fitness metric for each observation
- neighbours: named columns giving the number of neighbours of each column the names of the list elements are taken to be the names of the focal species.
If 'data' is a dataframe, it also needs a 'focal' column. Regardless of the data structure, all focal species need to have the same number of observations (i.e. same number of rows), and the set of neighbour species needs to be the same as the set of focal species, so that the neighbours columns correspond to the names of the list elements or, if 'data' is a dataframe, to the values of the 'focal' column. Future versions will relax this requirement.
model_family family of model to use. Available families are BH (Beverton-Holt), LV (LotkaVolterra), RK (Ricker), and LW (Law-Watkinson). Users may also define their own families and models (see vignette 4).
covariates a data structure equivalent to 'data', in which each column are the values of a covariate.
optimization_method
numerical optimization method.
lambda_cov_form
form of the covariate effects on lambda. Either "none" (no covariate effects) or "global" (one estimate per covariate).
effect_cov_form
form of the covariate effects on competitive effects. Either "none" (no covariate effects) or "global" (one estimate per covariate)
response_cov_form
form of the covariate effects on competitive responses. Either "none" (no covariate effects) or "global" (one estimate per covariate)
initial_values list with components "lambda","effect","response", and optionally "lambda_cov", "effect_cov", "response_cov", specifying the initial values for numerical optimization. Single values are allowed.
lower_bounds optional list with single values for "lambda", "effect","response", and optionally "lambda_cov", "effect_cov", "response_cov".
upper_bounds optional list with single values for "lambda", "effect","response", and optionally "lambda_cov", "effect_cov", "response_cov".
fixed_terms optional list specifying which model parameters are fixed.
bootstrap_samples
number of bootstrap samples for error calculation. Defaults to 0 , i.e. no error is calculated.


## Value

an object of class 'cxr_er_fit' which is a list with the following components:

- model_name: string with the name of the fitness model
- model: model function
- data: data supplied
- taxa: names of the taxa fitted
- covariates: covariate data supplied
- optimization_method: optimization method used
- initial_values: list with initial values
- fixed_terms: list with fixed terms
- lambda: fitted values for lambdas, or NULL if fixed
- effect: fitted values for competitive effects, or NULL if fixed
- response: fitted values for competitive responses, or NULL if fixed
- lambda_cov: fitted values for effect of covariates on lambdas, or NULL if fixed
- effect_cov: fitted values for effect of covariates on competitive effects, or NULL if fixed
- response_cov: fitted values for effect of covariates on competitive responses, or NULL if fixed
- lambda_standard_error: standard errors for lambdas, if calculated
- effect_standard_error: standard errors for competitive effects, if calculated
- response_standard_error: standard errors for competitive responses, if calculated
- lambda_cov_standard_error: standard errors for effect of covariates on lambdas, if calculated
- effect_cov_standard_error: standard errors for effect of covariates on competitive effects, if calculated
- response_cov_standard_error: standard errors for effect of covariates on competitive responses, if calculated
- log_likelihood: log-likelihood of the fits


## Examples

```
# fit three species at once
data("neigh_list")
# these species all have >250 observations
example_sp <- c("BEMA","LEMA","HOMA")
sp.pos <- which(names(neigh_list) %in% example_sp)
data <- neigh_list[sp.pos]
n.obs <- 250
# keep only fitness and neighbours columns
for(i in 1:length(data)){
    data[[i]] <- data[[i]][1:n.obs,c(2,sp.pos+2)]#2:length(data[[i]])]
}
# covariates: salinity
data("salinity_list")
salinity <- salinity_list[example_sp]
# keep only salinity column
for(i in 1:length(salinity)){
    salinity[[i]] <- salinity[[i]][1:n.obs,2:length(salinity[[i]])]
}
initial_values = list(lambda = 1,
                effect = 1,
                response = 1
```

```
    # lambda_cov = 0,
    # effect_cov = 0,
    # response_cov = 0
)
lower_bounds = list(lambda = 0,
    effect = 0,
    response = 0
    # lambda_cov = 0,
    # effect_cov = 0,
    # response_cov = 0
)
upper_bounds = list(lambda = 100,
    effect = 10,
    response = 10
    # lambda_cov = 0,
    # effect_cov = 0,
    # response_cov = 0
)
er_3sp <- cxr_er_fit(data = data,
            model_family = "BH",
            # fit without covariates,
            # as it may be very computationally expensive
            # covariates = salinity,
            optimization_method = "bobyqa",
            lambda_cov_form = "none",
            effect_cov_form = "none",
            response_cov_form = "none",
            initial_values = initial_values,
            lower_bounds = lower_bounds,
            upper_bounds = upper_bounds,
            # syntaxis for fixed values
            # fixed_terms = list("response"),
            bootstrap_samples = 3)
# brief summary
summary(er_3sp)
```

cxr_generate_test_data

## Description

Model fitness responses to neighbours and covariates using a Beverton-Holt functional form. This function is fairly restricted and under development, but can be used to generate simple test data to run the main functions of cxr.

## Usage

```
cxr_generate_test_data(
    focal_sp = 1,
    neigh_sp = 1 ,
    covariates = 0,
    observations = 10,
    alpha_form = c("pairwise", "none", "global"),
    lambda_cov_form = c("none", "global"),
    alpha_cov_form = c("none", "global", "pairwise"),
    focal_lambda = NULL,
    min_lambda \(=0\),
    max_lambda = 10,
    alpha = NULL,
    min_alpha = 0,
    max_alpha = 1,
    alpha_cov = NULL,
    min_alpha_cov = -1,
    max_alpha_cov = 1,
    lambda_cov = NULL,
    min_lambda_cov = -1,
    max_lambda_cov = 1 ,
    min_cov \(=0\),
    max_cov \(=1\)
)
```


## Arguments

| focal_sp <br> neigh_sp | number of focal species, defaults to 1. <br> number of neighbour species, defaults to 1. |
| :--- | :--- |
| covariates | number of covariates, defaults to 0. |
| observations | number of observations, defaults to 10. |
| alpha_form | what form does the alpha parameter take? one of "none" (no alpha in the model), <br> "global" (a single alpha for all pairwise interactions), or "pairwise" (one alpha <br> value for every interaction). |
| lambda_cov_form |  |
| form of the covariate effects on lambda. Either "none" (no covariate effects) or |  |
| "global" (one estimate per covariate). |  |


| alpha | optional interaction matrix, neigh_sp x neigh_sp |
| :--- | :--- |
| min_alpha | if no focal_alpha is provided, alphas are taken from a uniform distribution with <br> min_alpha and max_alpha as minimum and maximum values. |
| max_alpha | if no focal_alpha is provided, alphas are taken from a uniform distribution with <br> min_alpha and max_alpha as minimum and maximum values. |
| alpha_cov | _-_Under development-_ |
| min_alpha_cov | if no focal_alpha_cov is provided, alpha_covs are taken from a uniform dis- <br> tribution with min_alpha_cov and max_alpha_cov as minimum and maximum <br> values. |
| max_alpha_cov | if no focal_alpha_cov is provided, alpha_covs are taken from a uniform distri- <br> bution with min_alpha and max_alpha as minimum and maximum values. |
| lambda_cov | optional matrix of neigh_sp x covariates giving the effect of each covariate over <br> the fecundity (lambda) of each species. |
| min_lambda_covif no focal_lambda_cov is provided, lambda_covs are taken from a uniform dis- <br> tribution with min_lambda_cov and max_lambda_cov as minimum and maxi- <br> mum values. |  |
| max_lambda_covif no focal_lambda_cov is provided, lambda_covs are taken from a uniform dis- <br> tribution with min_lambda and max_lambda as minimum and maximum values. |  |
| max_cov minimum value for covariates |  |

## Value

list with two components: 'observations' is a list with as many components as focal species. Each component of 'observations' is a dataframe with stochastic number of neighbours and associated fitness. The second component, 'covariates', is again a list with one component per focal species. Each component of 'covariates' is a dataframe with the values of each covariate for each associated observation.

## Examples

```
example_obs <- cxr_generate_test_data(focal_sp = 2,
    neigh_sp = 2,
    alpha_form = "pairwise",
    lambda_cov_form = "global",
    alpha_cov_form = "global",
    covariates = 1)
```

```
cxr_pm_bootstrap Standard error estimates for model parameters
```


## Description

Computes bootstrap standard errors for a given population dynamics model. This function is provided for completeness, but error calculation is integrated in the function cxr_pm_fit.

## Usage

```
cxr_pm_bootstrap(
        fitness_model,
        optimization_method,
        data,
        focal_column,
        covariates,
        init_par,
        lower_bounds,
        upper_bounds,
        fixed_parameters,
        bootstrap_samples
    )
```


## Arguments

fitness_model function returning a single value to minimize, given a set of parameters and a fitness metric
optimization_method
numerical optimization method
data dataframe with observations in rows and two sets of columns:

- fitness: fitness metric for the focal individual
- neighbours: columns with user-defined names with number of neighbours for each group
focal_column optional integer value giving the position, or name, of the column with neighbours from the same species as the focal one. This is necessary if "alpha_intra" is specified.
covariates optional matrix with observations in rows and covariates in columns. Each cell is the value of a covariate in a given observation.
init_par $\quad 1 \mathrm{~d}$ vector of initial parameters
lower_bounds 1 d vector of lower bounds
upper_bounds 1 d vector of upper bounds
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","alpha_intra","alpha_inter" and "alpha_cov".
bootstrap_samples
how many bootstrap samples to compute.


## Value

1d vector, the standard error of each parameter in init_par

```
cxr_pm_fit General optimization for population models
```


## Description

Estimates parameters of user-specified population dynamics models.

## Usage

```
    cxr_pm_fit(
```

        data,
        focal_column = NULL,
        model_family,
        covariates \(=\) NULL ,
    optimization_method = c("Nelder-Mead", "BFGS", "CG", "ucminf", "L-BFGS-B", "nlm",
            "nlminb", "Rcgmin", "Rvmmin", "spg", "bobyqa", "nmkb", "hjkb", "nloptr_CRS2_LM",
            "nloptr_ISRES", "nloptr_DIRECT_L_RAND", "DEoptimR", "GenSA"),
        alpha_form = c("none", "global", "pairwise"),
        lambda_cov_form = c("none", "global"),
        alpha_cov_form = c("none", "global", "pairwise"),
    initial_values \(=\) list \(\left(\operatorname{lambda}=0\right.\), alpha_intra \(=0\), alpha_inter \(=0, ~ l a m b d a \_c o v=0\),
                alpha_cov = 0),
    lower_bounds = NULL,
    upper_bounds \(=\) NULL,
        fixed_terms \(=\) NULL,
        bootstrap_samples \(=0\)
    )
    
## Arguments

data dataframe with observations in rows and two sets of columns:

- fitness: fitness metric for the focal individual
- neighbours: numeric columns with user-defined names, giving number of neighbours for each group
focal_column optional integer or character giving the column with neighbours from the same species as the focal one. This field is necessary if "alpha_intra" is specified in initial_values, lower_bounds, upper_bounds, or fixed_terms.
model_family family of model to use. Available families are BH (Beverton-Holt), LV (LotkaVolterra), RK (Ricker), and LW (Law-Watkinson). Users may also define their own families and models (see vignette 4).
covariates optional named matrix or dataframe with observations (rows) of any number of environmental covariates (columns).

```
    optimization_method
                            numerical optimization method.
    alpha_form what form does the alpha parameter take? one of "none" (no alpha in the model),
        "global" (a single alpha for all pairwise interactions), or "pairwise" (one alpha
        value for every interaction).
    lambda_cov_form
    form of the covariate effects on lambda. Either "none" (no covariate effects) or
        "global" (one estimate per covariate).
    alpha_cov_form form of the covariate effects on alpha. One of "none" (no covariate effects),
        "global" (one estimate per covariate on every alpha), or "pairwise" (one estimate
        per covariate and pairwise alpha)
    initial_values list with components "lambda","alpha_intra","alpha_inter","lambda_cov", "al-
        pha_cov", specifying the initial values for numerical optimization. Single values
        are allowed.
    lower_bounds optional list with single values for "lambda","alpha_intra","alpha_inter","lambda_cov",
        "alpha_cov".
    upper_bounds optional list with single values for "lambda","alpha_intra","alpha_inter","lambda_cov",
        "alpha_cov".
    fixed_terms optional list of numeric vectors specifying the value of fixed model parameters,
    among "lambda","alpha_intra","alpha_inter","lambda_cov", and "alpha_cov".
    bootstrap_samples
        number of bootstrap samples for error calculation. Defaults to 0, i.e. no error is
        calculated.
```


## Value

an object of class 'cxr_pm_fit' which is a list with the following components:

- model_name: string with the name of the fitness model
- model: model function
- data: data supplied
- focal_ID: name/ID of the focal taxa, if provided in 'focal_column'
- covariates: covariate data supplied
- optimization_method: optimization method used
- initial_values: list with initial values
- fixed_terms: list with fixed terms
- lambda: fitted value for lambda, or NULL if fixed
- alpha_intra: fitted value for intraspecific alpha, or NULL if fixed
- alpha_inter: fitted value for interspecific alpha, or NULL if fixed
- lambda_cov: fitted value(s) for lambda_cov, or NULL if fixed.
- alpha_cov: fitted value(s) for alpha_cov, or NULL if fixed. These are structured as a list with one element for each covariate.
- lambda_standard_error: standard error for lambda, if computed
- alpha_intra_standard_error: standard error for intraspecific alpha, if computed
- alpha_inter_standard_error: standard error for interspecific alpha, if computed
- lambda_cov_standard_error: standard error for lambda_cov, if computed
- alpha_cov_standard_error: standard error for alpha_cov, if computed
- log_likelihood: log-likelihood of the fit


## Examples

```
data("neigh_list")
my.sp <- "BEMA"
# data for a single species, keep only fitness and neighbours columns
sp_data <- neigh_list[[my.sp]][2:ncol(neigh_list[[1]])]
sp_fit <- cxr_pm_fit(data = sp_data,
                focal_column = my.sp,
                optimization_method = "bobyqa",
                        model_family = "BH",
                        alpha_form = "pairwise",
                        lambda_cov_form = "none",
                        alpha_cov_form = "none",
        initial_values = list(lambda = 1,alpha_intra = 0.1,alpha_inter = 0.1),
            lower_bounds = list(lambda = 0,alpha_intra = 0,alpha_inter = 0),
            upper_bounds = list(lambda = 100,alpha_intra = 1,alpha_inter = 1),
            bootstrap_samples = 3)
    summary(sp_fit)
```

    cxr_pm_multifit Multi-species parameter optimization
    
## Description

This function is a wrapper for estimating parameters for several focal species, instead of making separate calls to cxr_pm_fit.

## Usage

cxr_pm_multifit( data,
model_family = c("BH"),
focal_column = NULL,
covariates = NULL,
optimization_method = c("BFGS", "CG", "Nelder-Mead", "ucminf", "L-BFGS-B", "nlm",
"nlminb", "Rcgmin", "Rvmmin", "spg", "bobyqa", "nmkb", "hjkb", "nloptr_CRS2_LM",
"nloptr_ISRES", "nloptr_DIRECT_L_RAND", "DEoptimR", "GenSA"),
alpha_form = c("none", "global", "pairwise"),

```
    lambda_cov_form = c("none", "global"),
    alpha_cov_form = c("none", "global", "pairwise"),
    initial_values = NULL,
    lower_bounds = NULL,
    upper_bounds = NULL,
    fixed_terms = NULL,
    bootstrap_samples = 0
)
```


## Arguments

| data | named list in which each component is a dataframe with a fitness column and a <br> number of columns representing neighbours <br> family of model to use. Available families are BH (Beverton-Holt), LV (Lotka- <br> Volterra), RK (Ricker), and LW (Law-Watkinson). Users may also define their <br> own families and models (see vignette 4). |
| :--- | :--- |
| model_family |  |
| character vector with the same length as data, giving the names of the columns |  |
| representing intraspecific observations for each species, or numeric vector giv- |  |
| ing the position of such columns. |  |
| optional named list in which each component is a dataframe with values of each |  |
| covariate for each observation. The ith component of covariates are the co- |  |
| variate values that correspond to the ith component of data, so they must have |  |
| the same number of observations. |  |

## Value

an object of class 'cxr_pm_multifit' which is a list with the following components:

- model_name: string with the name of the fitness model
- model: model function
- data: data supplied
- taxa: names of the taxa fitted
- covariates: covariate data supplied
- optimization_method: optimization method used
- initial_values: list with initial values
- fixed_terms: list with fixed terms
- lambda: fitted values for lambda, or NULL if fixed
- alpha_intra: fitted values for alpha_intra, or NULL if fixed
- alpha_inter: fitted values for alpha_inter, or NULL if fixed
- lambda_cov: fitted values for lambda_cov, or NULL if fixed
- alpha_cov: fitted values for alpha_cov, or NULL if fixed
- lambda_standard_error: standard errors for lambda, if computed
- alpha_standard_error: standard errors for alpha, if computed
- lambda_cov_standard_error: standard errors for lambda_cov, if computed
- alpha_cov_standard_error: standard errors for alpha_cov, if computed
- $\log _{\text {_likelihood: }}$ log-likelihoods of the fits


## Examples

```
# fit three species at once
data("neigh_list")
data <- neigh_list[1:3]
# keep only fitness and neighbours columns
for(i in 1:length(data)){
    data[[i]] <- data[[i]][,2:length(data[[i]])]
}
# be explicit about the focal species
focal.sp <- names(data)
# covariates: salinity
data("salinity_list")
salinity <- salinity_list[1:3]
# keep only salinity column
for(i in 1:length(salinity)){
    salinity[[i]] <- data.frame(salinity = salinity[[i]][,2:length(salinity[[i]])])
}
    fit_3sp <- cxr_pm_multifit(data = data,
                                    optimization_method = "bobyqa",
                                    model_family = "BH",
                                    focal_column = focal.sp,
```

```
covariates = salinity,
alpha_form = "pairwise",
lambda_cov_form = "global",
alpha_cov_form = "global",
initial_values = list(lambda = 1,
                                    alpha_intra = 0.1,
                                    alpha_inter = 0.1,
                                    lambda_cov = 0.1,
                                    alpha_cov = 0.1),
lower_bounds = list(lambda = 0.01,
    alpha_intra = 0,
    alpha_inter = 0,
    lambda_cov = 0,
    alpha_cov = 0),
upper_bounds = list(lambda = 100,
        alpha_intra = 1,
        alpha_inter = 1,
        lambda_cov = 1,
        alpha_cov = 1),
bootstrap_samples = 3)
```

\# brief summary
summary (fit_3sp)
\# interaction matrix
fit_3sp\$alpha_matrix
densities_to_df Converts a densities list to a tidy dataframe

## Description

Converts a densities list to a tidy dataframe

## Usage

densities_to_df(densities)

## Arguments

densities list, species (optionally x year) with each element holding a sites x stages matrix. This function assumes three life stages.

## Value

dataframe with columns species-stage-site(-year)-density
fill_demography_matrix

## Description

Fill for a given species, across all sites.

## Usage

fill_demography_matrix(focal.sp, vpm, transition_matrices)

## Arguments

focal.sp integer, focal species.
vpm data structure holding all vector-permutation matrices; see 'vec_permutation_matrices'.
If not in an appropriate format, it is likely to fail without warning.
transition_matrices
nested list species $x$ sites, in which each element holds a $3 \times 3$ transition matrix.
If not in that format, it is likely to fail without warning.

Value
vec-permutation demography matrix for a given species across sites.
fill_dispersal_matrix Fill the vec-permutation dispersal matrix

## Description

Fill for a given species, all sites

## Usage

```
fill_dispersal_matrix(
    focal.sp,
    num.sites,
    param,
    vpm,
    env = NULL,
    current.densities
)
```


## Arguments

```
    focal.sp integer, focal species
    num.sites integer, how many sites
    param param nested list,see 'build_param' function
    vpm data structure holding all vector-permutation matrices; see 'vec_permutation_matrices'
    env optional numeric, environmental forcing for a given timestep
    current.densities
                            list of length sp, each element is a matrix sites*stages
```


## Value

dispersal matrix, stages*sites

```
fill_transition_matrix
```

Fill a transition matrix

## Description

Calculates the elements of a site-specific transition matrix for a given sp. Note that here, and through all functions, we fix three life stages. Also note that 'param' and 'env' must match, as for the 'vital_rate' function.

## Usage

fill_transition_matrix(focal.sp, site, param, env = NULL, current.densities)

## Arguments

focal.sp integer, species
site integer, site
param param structure (see 'build_param' function)
env optional numeric, environmental forcing for a given timestep
current.densities
list of length sp , each element is a matrix site*stages

## Value

$3 \times 3$ transition matrix

## Description

Fitness ratio among two or more species

## Usage

```
fitness_ratio(
    effect_response_fit = NULL,
    fitness_sp1 = NULL,
    fitness_sp2 = NULL
    )
```


## Arguments

effect_response_fit cxr_er_fit object
fitness_sp1 numeric value representing the fitness (a.k.a. competitive ability) of the first taxa
fitness_sp2 numeric value representing the fitness (a.k.a. competitive ability) of the second taxa

## Value

either a matrix with fitness ratios for all pairs of fitted species, or a single numeric value. The matrix elements represent the ratios of species in columns over species in rows, and conversely, the numeric value represents the ratio of sp 1 over sp 2 .

## Examples

fitness_ratio(fitness_sp1 = 0.6, fitness_sp2 = 0.3)

```
generate_vital_rate_coefs
```

Generate coefficients for obtaining vital rates

## Description

Any vital rate is a function of several parameters, potentially including interactions or environmental effects. This function generates the coefficients for these parameters, so that users do not have to introduce them all manually in a 'param' list. Coefficients can be generated from a random sampling of a normal distribution with specified mean and standard deviation, or they can be retrieved from a model object that accepts a 'tidy' function from the broom/broom.mixed packages. This is because coefficients for vital rates can be understood as coefficients from statistical regressions.

## Usage

```
    generate_vital_rate_coefs(
        param,
        sp = NULL,
        sites = NULL,
        vital.rate = NULL,
        vr.coef = NULL,
        mean.coef = NULL,
        sd.coef = NULL,
        glm.object = NULL,
        glm.coef.equivalence = NULL
    )
```


## Arguments

param the original list with the structure of species, sites, vital rates to calculate, and parameters affecting them. See the function 'build_param'
sp number or character of the species to calculate coefficients for. If empty, all species are assumed.
sites number or character of the sites to calculate coefficients for. If empty, all sites are assumed.
vital.rate character giving the vital rate to calculate coefficients for. If empty, all vital rates are assumed.
vr.coef character giving a specific coefficient to calculate. If empty, all coefficients are assumed.
mean. coef optional numeric value, mean for sampling coefficient values
sd.coef optional numeric value, standard deviation for sampling coefficient values
glm.object optional model object/coef table
glm.coef.equivalence
if a glm table is provided and its names differ from the 'param' data structure, you can include a named list in which names are the names from 'param' and its elements are the equivalent names from the glm table

## Details

In the current version, we assume that the model coefficients come from a logistic regression with binomial family. Otherwise, the function will probably not fail, but the coefficients will not be interpretable and the results in terms of obtaining the actual vital rates from these will be meaningless.
Also note that you need to take care manually of the signs of the coefficients, if entered through mean/sd pairs.

## Value

the updated parameter list

## Description

A table with coefficients from a GLM to serve as an example for importing into the data structure of the metapopulation model.

## Usage

```
data(glm_example_coefs)
```


## Format

A named numerical matrix of 8 rows and 4 columns

```
LV_er_lambdacov_global_effectcov_global_responsecov_global
    Effect response Lotka-Volterra model with covariate effects on
    lambda, effect, and response
```


## Description

Note that, as e and r are not pair-specific, all species parameters are fit in the same function.

```
Usage
    LV_er_lambdacov_global_effectcov_global_responsecov_global(
        par,
        fitness,
        target,
        density,
        covariates,
        fixed_parameters
    )
```


## Arguments

par 1d vector with initial parameters in the order: lambda,lambda_cov,effect,effect_cov,response,response_co
fitness $\quad 1 d$ vector with fitness observations
target matrix with species in rows, observations in columns. Value is 1 if a species is focal for a given observation, 0 otherwise.
density matrix with species in rows, observations in columns. Value is density of each sp as neighbour for each observation.
covariates numeric dataframe or matrix with observations in rows and covariates in columns.
Each cell is the value of a covariate in a given observation
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","lambda_cov","effect","eff
"response","response_cov".

## Value

log-likelihood value

```
LV_er_lambdacov_none_effectcov_none_responsecov_none
    Effect response Lotka-Volterra model without covariate effects
```


## Description

Note that, as e and r are not pair-specific, all species parameters are fit in the same function.

```
Usage
    LV_er_lambdacov_none_effectcov_none_responsecov_none(
        par,
        fitness,
        target,
        density,
        covariates,
        fixed_parameters
    )
```


## Arguments

par 1d vector with initial parameters in the order: lambda,effect,response,sigma.
fitness $\quad 1 d$ vector with fitness observations.
target matrix with species in rows, observations in columns. Value is 1 if a species is focal for a given observation, 0 otherwise.
density matrix with species in rows, observations in columns. Value is density of each sp as neighbour for each observation.
covariates included for compatibility, not used in this model.
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","effect","response".

## Value

log-likelihood value

```
LV_pm_alpha_global_lambdacov_none_alphacov_none
    Lotka-Volterra model with a global alpha and no covariate effects
```


## Description

Lotka-Volterra model with a global alpha and no covariate effects

## Usage

```
LV_pm_alpha_global_lambdacov_none_alphacov_none(
    par,
    fitness,
    neigh_intra_matrix = NULL,
    neigh_inter_matrix,
    covariates,
    fixed_parameters
)
```


## Arguments

par 1d vector of initial parameters: lambda, alpha, and sigma.
fitness $\quad 1 d$ vector of fitness observations, in log scale.
neigh_intra_matrix
included for compatibility, not used in this model.
neigh_inter_matrix
matrix of arbitrary columns, number of neighbours for each observation. As in this model there is a single alpha argument, do not distinguish neighbour identity
covariates included for compatibility, not used in this model.
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","alpha_inter".

## Value

log-likelihood value

LV_pm_alpha_none_lambdacov_none_alphacov_none
Lotka-Volterra model with no alphas and no covariate effects

## Description

This model, in all families, is simply given by lambda.

## Usage

```
LV_pm_alpha_none_lambdacov_none_alphacov_none(
    par,
    fitness,
    neigh_intra_matrix = NULL,
    neigh_inter_matrix,
    covariates,
    fixed_parameters
)
```


## Arguments

par $\quad$ 1d vector of initial parameters: lambda and sigma
fitness

| neigh_intra_matrix |
| :--- |
| included for compatibility, not used in this model. |


| neigh_inter_matrix |
| ---: | :--- |


| included for compatibility, not used in this model. |
| :--- | :--- |


| covariates | included for compatibility, not used in this model |
| :--- | :--- |
| fixed_parameters |  | included for compatibility, not used in this model

## Value

log-likelihood value

```
LV_pm_alpha_pairwise_lambdacov_global_alphacov_global
    Lotka-Volterra model with pairwise alphas and global covariate ef-
    fects on lambda and alpha
```


## Description

Lotka-Volterra model with pairwise alphas and global covariate effects on lambda and alpha

## Usage

```
LV_pm_alpha_pairwise_lambdacov_global_alphacov_global(
    par,
    fitness,
    neigh_intra_matrix = NULL,
    neigh_inter_matrix,
    covariates,
    fixed_parameters
)
```


## Arguments

par
1d vector of initial parameters: lambda, lambda_cov, alpha, alpha_cov, and sigma
fitness $\quad 1 d$ vector of fitness observations, in $\log$ scale
neigh_intra_matrix optional matrix of one column, number of intraspecific neighbours for each observation
neigh_inter_matrix matrix of arbitrary columns, number of interspecific neighbours for each observation
covariates optional matrix with observations in rows and covariates in columns. Each cell is the value of a covariate in a given observation
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","alpha_intra","alpha_inter"

## Value

log-likelihood value

```
LV_pm_alpha_pairwise_lambdacov_global_alphacov_pairwise
    Lotka-Volterra model with pairwise alphas, covariate effects on
    lambda, and pairwise covariate effects on alpha
```


## Description

Lotka-Volterra model with pairwise alphas, covariate effects on lambda, and pairwise covariate effects on alpha

## Usage

```
LV_pm_alpha_pairwise_lambdacov_global_alphacov_pairwise(
    par,
    fitness,
    neigh_intra_matrix = NULL,
    neigh_inter_matrix,
    covariates,
    fixed_parameters
)
```


## Arguments

par
1d vector of initial parameters: lambda, lambda_cov, alpha, alpha_cov, and sigma
fitness 1d vector of fitness observations, in log scale
neigh_intra_matrix
optional matrix of one column, number of intraspecific neighbours for each ob-
servation
neigh_inter_matrix
matrix of arbitrary columns, number of interspecific neighbours for each obser-
vation
covariates optional matrix with observations in rows and covariates in columns. Each cell
is the value of a covariate in a given observation
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","alpha_intra","alpha_inter"

## Value

log-likelihood value

```
LV_pm_alpha_pairwise_lambdacov_none_alphacov_none
    Lotka-Volterra model with pairwise alphas and no covariate effects
```


## Description

Lotka-Volterra model with pairwise alphas and no covariate effects

## Usage

```
LV_pm_alpha_pairwise_lambdacov_none_alphacov_none(
    par,
    fitness,
    neigh_intra_matrix = NULL,
    neigh_inter_matrix,
    covariates,
    fixed_parameters
    )
```


## Arguments

par 1d vector of initial parameters: 'lambda', 'alpha_intra' (optional), 'alpha_inter', and 'sigma'
fitness $\quad 1 d$ vector of fitness observations, in log scale
neigh_intra_matrix
optional matrix of one column, number of intraspecific neighbours for each observation
neigh_inter_matrix
matrix of arbitrary columns, number of interspecific neighbours for each observation
covariates included for compatibility, not used in this model
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","alpha_intra","alpha_inter"

## Value

log-likelihood value

LV_project_alpha_global_lambdacov_none_alphacov_none
Lotka-Volterra model for projecting abundances, with a global alpha and no covariate effects

## Description

Lotka-Volterra model for projecting abundances, with a global alpha and no covariate effects

## Usage

LV_project_alpha_global_lambdacov_none_alphacov_none( lambda, alpha_intra, alpha_inter, lambda_cov, alpha_cov, abundance, covariates
)

## Arguments

lambda numeric lambda value.
alpha_intra included for compatibility, not used in this model.
alpha_inter single numeric value.
lambda_cov included for compatibility, not used in this model.
alpha_cov included for compatibility, not used in this model.
abundance named numeric vector of abundances in the previous timestep.
covariates included for compatibility, not used in this model.

## Value

numeric abundance projected one timestep

```
LV_project_alpha_none_lambdacov_none_alphacov_none
    Model for projecting abundances, with no alpha and no covariate ef-
    fects
```


## Description

Model for projecting abundances, with no alpha and no covariate effects

## Usage

LV_project_alpha_none_lambdacov_none_alphacov_none( lambda, alpha_intra, alpha_inter, lambda_cov, alpha_cov, abundance, covariates
)

## Arguments

| lambda | numeric lambda value. |
| :--- | :--- |
| alpha_intra | included for compatibility, not used in this model. |
| alpha_inter | included for compatibility, not used in this model. |
| lambda_cov | included for compatibility, not used in this model. |
| alpha_cov | included for compatibility, not used in this model. |
| abundance | named numeric vector of abundances in the previous timestep. |
| covariates | included for compatibility, not used in this model. |

## Value

numeric abundance projected one timestep

LV_project_alpha_pairwise_lambdacov_global_alphacov_global
Lotka-Volterra model for projecting abundances, with specific alpha values and global covariate effects on alpha and lambda

## Description

Lotka-Volterra model for projecting abundances, with specific alpha values and global covariate effects on alpha and lambda

## Usage

```
LV_project_alpha_pairwise_lambdacov_global_alphacov_global(
        lambda,
        alpha_intra,
        alpha_inter,
        lambda_cov,
        alpha_cov,
        abundance,
        covariates
    )
```


## Arguments

| lambda | numeric lambda value. |
| :--- | :--- |
| alpha_intra | single numeric value. |
| alpha_inter | numeric vector with interspecific alpha values. |
| lambda_cov | numeric vector with effects of covariates over lambda. |
| alpha_cov | named list of numeric values with effects of each covariate over alpha. |
| abundance | named numeric vector of abundances in the previous timestep. <br> covariates |
|  | matrix with observations in rows and covariates in columns. Each cell is the <br> value of a covariate in a given observation. |

## Value

numeric abundance projected one timestep

LV_project_alpha_pairwise_lambdacov_global_alphacov_pairwise
Lotka-Volterra model for projecting abundances, with specific alpha values and global covariate effects on alpha and lambda

## Description

Lotka-Volterra model for projecting abundances, with specific alpha values and global covariate effects on alpha and lambda

```
Usage
    LV_project_alpha_pairwise_lambdacov_global_alphacov_pairwise(
        lambda,
    alpha_intra,
    alpha_inter,
    lambda_cov,
    alpha_cov,
    abundance,
    covariates
)
```


## Arguments

lambda named numeric lambda value.
alpha_intra
alpha_inter
lambda_cov
alpha_cov named list of named numeric vectors with effects of each covariate over alpha values.
abundance named numeric vector of abundances in the previous timestep.
covariates matrix with observations in rows and covariates in named columns. Each cell is the value of a covariate in a given observation.

## Value

numeric abundance projected one timestep

```
LV_project_alpha_pairwise_lambdacov_none_alphacov_none
    Lotka-Volterra model for projecting abundances, with specific alpha
    values and no covariate effects
```


## Description

Lotka-Volterra model for projecting abundances, with specific alpha values and no covariate effects

## Usage

```
LV_project_alpha_pairwise_lambdacov_none_alphacov_none(
    lambda,
    alpha_intra,
    alpha_inter,
    lambda_cov,
    alpha_cov,
    abundance,
    covariates
    )
```


## Arguments

lambda numeric lambda value.
alpha_intra included for compatibility, not used in this model.
alpha_inter single numeric value.
lambda_cov included for compatibility, not used in this model.
alpha_cov included for compatibility, not used in this model.
abundance named numeric vector of abundances in the previous timestep.
covariates included for compatibility, not used in this model.

## Value

numeric abundance projected one timestep

```
LW_er_lambdacov_global_effectcov_global_responsecov_global
    Effect response Law-Watkinson model with covariate effects on
    lambda, effect, and response
```


## Description

Note that, as e and $r$ are not pair-specific, all species parameters are fit in the same function.

## Usage

```
    LW_er_lambdacov_global_effectcov_global_responsecov_global(
        par,
        fitness,
        target,
        density,
        covariates,
        fixed_parameters
    )
```


## Arguments

par 1d vector with initial parameters in the order: lambda,lambda_cov,effect,effect_cov,response,response_co
fitness $\quad 1 d$ vector with fitness observations
target matrix with species in rows, observations in columns. Value is 1 if a species is focal for a given observation, 0 otherwise.
density matrix with species in rows, observations in columns. Value is density of each sp as neighbour for each observation.
covariates numeric dataframe or matrix with observations in rows and covariates in columns. Each cell is the value of a covariate in a given observation
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","lambda_cov","effect","eff "response","response_cov".

## Value

log-likelihood value

```
LW_er_lambdacov_none_effectcov_none_responsecov_none
                        Effect response Law-Watkinson model without covariate effects
```


## Description

Note that, as e and $r$ are not pair-specific, all species parameters are fit in the same function.

## Usage

```
LW_er_lambdacov_none_effectcov_none_responsecov_none(
        par,
        fitness,
        target,
        density,
        covariates,
        fixed_parameters
    )
```


## Arguments

par 1d vector with initial parameters in the order: lambda,effect,response,sigma.
fitness $\quad 1 d$ vector with fitness observations.
target matrix with species in rows, observations in columns. Value is 1 if a species is focal for a given observation, 0 otherwise.
density matrix with species in rows, observations in columns. Value is density of each sp as neighbour for each observation.
covariates included for compatibility, not used in this model.
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","effect","response".

## Value

log-likelihood value

LW_pm_alpha_global_lambdacov_none_alphacov_none
Law-Watkinson model with a global alpha and no covariate effects

## Description

Law-Watkinson model with a global alpha and no covariate effects

## Usage

```
LW_pm_alpha_global_lambdacov_none_alphacov_none(
    par,
    fitness,
    neigh_intra_matrix = NULL,
    neigh_inter_matrix,
    covariates,
    fixed_parameters
    )
```


## Arguments

par 1d vector of initial parameters: lambda, alpha, and sigma.
fitness $\quad 1 d$ vector of fitness observations, in $\log$ scale.
neigh_intra_matrix included for compatibility, not used in this model.
neigh_inter_matrix
matrix of arbitrary columns, number of neighbours for each observation. As in this model there is a single alpha argument, do not distinguish neighbour identity
covariates included for compatibility, not used in this model.
fixed_parameters optional list specifying values of fixed parameters, with components "lambda","alpha_inter".

## Value

log-likelihood value

LW_pm_alpha_none_lambdacov_none_alphacov_none
Law-Watkinson model with no alphas and no covariate effects

## Description

This model, in all families, is simply given by lambda.

## Usage

LW_pm_alpha_none_lambdacov_none_alphacov_none(
par,
fitness,
neigh_intra_matrix = NULL,
neigh_inter_matrix,
covariates,
fixed_parameters
)

## Arguments

| par | 1 d vector of initial parameters: lambda and sigma |
| :---: | :---: |
| fitness | 1 d vector of fitness observations, in log scale |
| neigh_intra_matrix |  |
|  | included for compatibility, not used in this model. |
| neigh_inter_matrix |  |
|  | included for compatibility, not used in this model. |
| covariates | included for compatibility, not used in this model |
| fixed_parameters |  |
|  | included for compatibility, not used in this model |

## Value

log-likelihood value

```
LW_pm_alpha_pairwise_lambdacov_global_alphacov_global
                    Law-Watkinson model with pairwise alphas and global covariate ef-
                    fects on lambda and alpha
```


## Description

Law-Watkinson model with pairwise alphas and global covariate effects on lambda and alpha

## Usage

LW_pm_alpha_pairwise_lambdacov_global_alphacov_global(
par,
fitness,
neigh_intra_matrix = NULL,
neigh_inter_matrix,
covariates,
fixed_parameters
)

## Arguments

par 1d vector of initial parameters: lambda, lambda_cov, alpha, alpha_cov, and sigma
fitness 1d vector of fitness observations, in log scale
neigh_intra_matrix
optional matrix of one column, number of intraspecific neighbours for each observation

```
neigh_inter_matrix
```

                    matrix of arbitrary columns, number of interspecific neighbours for each obser-
                        vation
    covariates optional matrix with observations in rows and covariates in columns. Each cell
is the value of a covariate in a given observation
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","alpha_intra","alpha_inter"

## Value

log-likelihood value

```
LW_pm_alpha_pairwise_lambdacov_global_alphacov_pairwise
    Law-Watkinson model with pairwise alphas, covariate effects on
    lambda, and pairwise covariate effects on alpha
```


## Description

Law-Watkinson model with pairwise alphas, covariate effects on lambda, and pairwise covariate effects on alpha

## Usage

LW_pm_alpha_pairwise_lambdacov_global_alphacov_pairwise(
par,
fitness,
neigh_intra_matrix = NULL,
neigh_inter_matrix, covariates, fixed_parameters
)

## Arguments

par 1d vector of initial parameters: lambda, lambda_cov, alpha, alpha_cov, and sigma
fitness $\quad 1 \mathrm{~d}$ vector of fitness observations, in log scale
neigh_intra_matrix
optional matrix of one column, number of intraspecific neighbours for each observation
neigh_inter_matrix
matrix of arbitrary columns, number of interspecific neighbours for each observation
covariates optional matrix with observations in rows and covariates in columns. Each cell is the value of a covariate in a given observation
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","alpha_intra","alpha_inter"

## Value

log-likelihood value

```
LW_pm_alpha_pairwise_lambdacov_none_alphacov_none
```

                                    Law-Watkinson model with pairwise alphas and no covariate effects
    
## Description

Law-Watkinson model with pairwise alphas and no covariate effects

## Usage

LW_pm_alpha_pairwise_lambdacov_none_alphacov_none( par, fitness,
neigh_intra_matrix = NULL,
neigh_inter_matrix, covariates, fixed_parameters
)

## Arguments

par 1d vector of initial parameters: 'lambda', 'alpha_intra' (optional), 'alpha_inter', and 'sigma'
fitness $\quad 1 d$ vector of fitness observations, in log scale
neigh_intra_matrix
optional matrix of one column, number of intraspecific neighbours for each observation
neigh_inter_matrix
matrix of arbitrary columns, number of interspecific neighbours for each observation
covariates included for compatibility, not used in this model
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","alpha_intra","alpha_inter"

## Value

log-likelihood value

LW_project_alpha_global_lambdacov_none_alphacov_none
Law-Watkinson model for projecting abundances, with a global alpha and no covariate effects

## Description

Law-Watkinson model for projecting abundances, with a global alpha and no covariate effects

## Usage

```
LW_project_alpha_global_lambdacov_none_alphacov_none(
        lambda,
        alpha_intra,
        alpha_inter,
        lambda_cov,
        alpha_cov,
        abundance,
        covariates
    )
```


## Arguments

| lambda | numeric lambda value. |
| :--- | :--- |
| alpha_intra | included for compatibility, not used in this model. |
| alpha_inter | single numeric value. |
| lambda_cov | included for compatibility, not used in this model. |
| alpha_cov | included for compatibility, not used in this model. |
| abundance | named numeric vector of abundances in the previous timestep. |
| covariates | included for compatibility, not used in this model. |

## Value

numeric abundance projected one timestep

```
LW_project_alpha_none_lambdacov_none_alphacov_none
                Model for projecting abundances, with no alpha and no covariate ef-
                    fects
```


## Description

Model for projecting abundances, with no alpha and no covariate effects

## Usage

```
LW_project_alpha_none_lambdacov_none_alphacov_none(
        lambda,
        alpha_intra,
        alpha_inter,
        lambda_cov,
        alpha_cov,
        abundance,
        covariates
    )
```


## Arguments

lambda numeric lambda value.
alpha_intra included for compatibility, not used in this model.
alpha_inter included for compatibility, not used in this model.
lambda_cov included for compatibility, not used in this model.
alpha_cov included for compatibility, not used in this model.
abundance named numeric vector of abundances in the previous timestep.
covariates included for compatibility, not used in this model.

## Value

numeric abundance projected one timestep

LW_project_alpha_pairwise_lambdacov_global_alphacov_global
Law-Watkinson model for projecting abundances, with specific alpha values and global covariate effects on alpha and lambda

## Description

Law-Watkinson model for projecting abundances, with specific alpha values and global covariate effects on alpha and lambda

```
Usage
    LW_project_alpha_pairwise_lambdacov_global_alphacov_global(
    lambda,
    alpha_intra,
    alpha_inter,
    lambda_cov,
    alpha_cov,
    abundance,
    covariates
)
```


## Arguments

| lambda | numeric lambda value. |
| :--- | :--- |
| alpha_intra | single numeric value. |
| alpha_inter | numeric vector with interspecific alpha values. |
| lambda_cov | numeric vector with effects of covariates over lambda. |
| alpha_cov | named list of numeric values with effects of each covariate over alpha. |
| abundance | named numeric vector of abundances in the previous timestep. |
| covariates | matrix with observations in rows and covariates in columns. Each cell is the <br> value of a covariate in a given observation. |

## Value

numeric abundance projected one timestep

```
LW_project_alpha_pairwise_lambdacov_global_alphacov_pairwise
    Law-Watkinson model for projecting abundances, with specific alpha
    values and global covariate effects on alpha and lambda
```


## Description

Law-Watkinson model for projecting abundances, with specific alpha values and global covariate effects on alpha and lambda

## Usage

LW_project_alpha_pairwise_lambdacov_global_alphacov_pairwise( lambda, alpha_intra, alpha_inter, lambda_cov, alpha_cov, abundance, covariates
)

## Arguments

lambda named numeric lambda value.
alpha_intra single numeric value.
alpha_inter numeric vector with interspecific alpha values.
lambda_cov numeric vector with effects of covariates over lambda.
alpha_cov named list of named numeric vectors with effects of each covariate over alpha values.

| abundance | named numeric vector of abundances in the previous timestep. |
| :--- | :--- |
| covariates | matrix with observations in rows and covariates in named columns. Each cell is <br> the value of a covariate in a given observation. |

Value
numeric abundance projected one timestep

LW_project_alpha_pairwise_lambdacov_none_alphacov_none
Law-Watkinson model for projecting abundances, with specific alpha values and no covariate effects

## Description

Law-Watkinson model for projecting abundances, with specific alpha values and no covariate effects

## Usage

LW_project_alpha_pairwise_lambdacov_none_alphacov_none( lambda, alpha_intra, alpha_inter, lambda_cov, alpha_cov, abundance, covariates
)

## Arguments

lambda numeric lambda value.
alpha_intra included for compatibility, not used in this model.
alpha_inter single numeric value.
lambda_cov included for compatibility, not used in this model.
alpha_cov included for compatibility, not used in this model.
abundance named numeric vector of abundances in the previous timestep.
covariates included for compatibility, not used in this model.

## Value

numeric abundance projected one timestep

```
metapopulation_example_param
    Metapopulation dynamics coefficients
```


## Description

A nested list containing vital rate coefficients for projecting metapopulation dynamics. The first level of the list has 3 elements, one for each species modelled. The second level of the list has 2 elements, one for each site modelled. For each combination species-site, there is a data.frame of eight rows - one per each vital rate, and eight columns - one per coefficient, that correspond to the coefficients of a GLM. These are named as alpha,beta1, etc, in the data.frame, and correspond to the intercept, environmental effect, effects of each of the three species' density, and environment:density interactions

## Usage

data(metapopulation_example_param)

## Format

A nested list with $3 \times 2$ elements, each of which a dataframe of 8 rows and 8 numeric columns
neigh_list neighbours and fitness observations

## Description

A dataset containing fitness and neighbours for plant individuals of 17 species. The dataset is a named list with 16 elements, each of which is a dataframe with the following columns:

- obs_ID: unique identifier for each observation
- fitness: number of viable seeds of the focal individual
- 17 columns indicating the number of neighbours from each plant sp . in a radius of 7.5 cm from the focal individual


## Usage <br> data(neigh_list)

## Format

A list with 17 elements, each of which a dataframe of variable number of rows and 18 columns

## Note

For details, see Lanuza et al. 2018 Ecology Letters.
niche_overlap Niche overlap between two species

## Description

quoting Godoy et al. (2014): reflects the average degree to which species limit individuals of their own species relative to competitors. Low niche overlap causes species to have greater per capita growth rates when rare than when common. If species limit individuals of their own species and their competitors equally, then niche overlap is 1 , and coexistence is not possible unless species are otherwise identical. At the other extreme, if species have no interspecific effects, then niche overlap is 0 .

## Usage

```
    niche_overlap(
        cxr_multifit = NULL,
        cxr_sp1 = NULL,
        cxr_sp2 = NULL,
        pair_matrix = NULL
    )
```


## Arguments

cxr_multifit cxr_pm_multifit object, with parameters for a series of species.
cxr_sp1 cxr_pm_fit object giving the parameters from the first species.
cxr_sp2 cxr_pm_fit object giving the parameters from the second species.
pair_matrix $2 \times 2$ matrix with intra and interspecific interaction coefficients between the two species.

## Details

Niche overlap has a common functional form, in the context of Modern Coexistence Theory (MCT), for a series of models, including those specified in table A1 of Hart et al. (2018) Journal of Ecology 106, 1902-1909. Other model families may not adhere to the general definition.
Furthermore, the MCT definition only accounts for competitive interactions (i.e. positive alpha coefficients in these models). An alternative definition is given in Saavedra et al. (2017) Ecological Monographs 87,470-486. In this 'structural approach', positive interactions are allowed. Incidentally, both approaches yield qualitatively similar, but not equivalent, results for purely competitive matrices.
In all cases, these definitions only apply to models whose feasible equilibrium point can be described by a linear equation (see Saavedra et al. 2017, Hart et al. 2018 for details).
This function calculates niche overlap among two or more taxa, using both the MCT and the structural formulation. The function, as in avg_fitness_diff and competitive_ability, accepts three different parameterizations:

- A cxr_pm_multifit object, from which niche overlap will be computed across all species pairs.
- two cxr_pm_fit objects, one for each species.
- explicit lambda and alpha values, as well as the model family from which these parameters were obtained.

If negative interactions are present, the MCT niche overlap will be NA. The cxr objects may be calculated with user-defined model families. If this is the case, or if simply a $2 \times 2$ matrix is provided, the niche overlap metrics will be calculated and a warning will be raised.

## Value

either a dataframe with as many rows as species, or a single named numeric vector, containing niche overlap values for the MCT (modern coexistence theory) and SA (structural approach) formulations.

## Examples

niche_overlap(pair_matrix $=\operatorname{matrix}(c(0.33,0.12,0.2,0.4)$, nrow $=2))$

```
RK_er_lambdacov_global_effectcov_global_responsecov_global
                                    Effect response Beverton-Holt model with covariate effects on lambda,
                                    effect, and response
```


## Description

Note that, as e and $r$ are not pair-specific, all species parameters are fit in the same function.

## Usage

```
    RK_er_lambdacov_global_effectcov_global_responsecov_global(
        par,
        fitness,
        target,
        density,
        covariates,
        fixed_parameters
    )
```


## Arguments

par 1d vector with initial parameters in the order: lambda,lambda_cov,effect,effect_cov,response,response_co fitness $\quad 1 d$ vector with fitness observations
target matrix with species in rows, observations in columns. Value is 1 if a species is focal for a given observation, 0 otherwise.
density matrix with species in rows, observations in columns. Value is density of each sp as neighbour for each observation.

```
    covariates numeric dataframe or matrix with observations in rows and covariates in columns.
    Each cell is the value of a covariate in a given observation
    fixed_parameters
```

                            optional list specifying values of fixed parameters, with components "lambda","lambda_cov","effect","eff
                            "response","response_cov".
    
## Value

log-likelihood value

```
RK_er_lambdacov_none_effectcov_none_responsecov_none
    Effect response Ricker model without covariate effects
```


## Description

Note that, as e and r are not pair-specific, all species parameters are fit in the same function.

```
Usage
    RK_er_lambdacov_none_effectcov_none_responsecov_none(
        par,
        fitness,
        target,
        density,
        covariates,
        fixed_parameters
    )
```


## Arguments

par 1d vector with initial parameters in the order: lambda,effect,response,sigma.
fitness $\quad 1 \mathrm{~d}$ vector with fitness observations.
target matrix with species in rows, observations in columns. Value is 1 if a species is focal for a given observation, 0 otherwise.
density matrix with species in rows, observations in columns. Value is density of each sp as neighbour for each observation.
covariates included for compatibility, not used in this model.
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","effect","response".

## Value

log-likelihood value

```
RK_pm_alpha_global_lambdacov_none_alphacov_none
    Ricker model with a global alpha and no covariate effects
```


## Description

Ricker model with a global alpha and no covariate effects

## Usage

```
RK_pm_alpha_global_lambdacov_none_alphacov_none(
        par,
        fitness,
        neigh_intra_matrix = NULL,
        neigh_inter_matrix,
        covariates,
        fixed_parameters
    )
```


## Arguments

par 1d vector of initial parameters: lambda, alpha, and sigma.
fitness $\quad 1 d$ vector of fitness observations, in log scale.
neigh_intra_matrix
included for compatibility, not used in this model.
neigh_inter_matrix
matrix of arbitrary columns, number of neighbours for each observation. As in this model there is a single alpha argument, do not distinguish neighbour identity
covariates included for compatibility, not used in this model.
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","alpha_inter".

## Value

log-likelihood value

```
RK_pm_alpha_none_lambdacov_none_alphacov_none
    Ricker model with no alphas and no covariate effects
```


## Description

This model, in all families, is simply given by lambda.

## Usage

```
RK_pm_alpha_none_lambdacov_none_alphacov_none(
    par,
    fitness,
    neigh_intra_matrix = NULL,
    neigh_inter_matrix,
    covariates,
    fixed_parameters
)
```


## Arguments

| par | 1d vector of initial parameters: lambda and sigma |
| :---: | :---: |
| fitness | 1d vector of fitness observations, in log scale |
| neigh_intra_matrix |  |
|  | included for compatibility, not used in this model. |
| neigh_inter_matrix |  |
|  | included for compatibility, not used in this model. |
| covariates i | included for compatibility, not used in this model |
| fixed_parameters |  | included for compatibility, not used in this model

## Value

log-likelihood value

```
RK_pm_alpha_pairwise_lambdacov_global_alphacov_global
    Ricker model with pairwise alphas and global covariate effects on
    lambda and alpha
```


## Description

Ricker model with pairwise alphas and global covariate effects on lambda and alpha

## Usage

```
RK_pm_alpha_pairwise_lambdacov_global_alphacov_global(
        par,
        fitness,
        neigh_intra_matrix = NULL,
        neigh_inter_matrix,
        covariates,
        fixed_parameters
    )
```


## Arguments

par
1d vector of initial parameters: lambda, lambda_cov, alpha, alpha_cov, and sigma
fitness $\quad 1 d$ vector of fitness observations, in log scale
neigh_intra_matrix optional matrix of one column, number of intraspecific neighbours for each observation
neigh_inter_matrix matrix of arbitrary columns, number of interspecific neighbours for each observation
covariates optional matrix with observations in rows and covariates in columns. Each cell is the value of a covariate in a given observation
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","alpha_intra","alpha_inter"

## Value

log-likelihood value

```
RK_pm_alpha_pairwise_lambdacov_global_alphacov_pairwise
    Ricker model with pairwise alphas, covariate effects on lambda, and
    pairwise covariate effects on alpha
```


## Description

Ricker model with pairwise alphas, covariate effects on lambda, and pairwise covariate effects on alpha

## Usage

```
RK_pm_alpha_pairwise_lambdacov_global_alphacov_pairwise(
    par,
    fitness,
    neigh_intra_matrix = NULL,
    neigh_inter_matrix,
    covariates,
    fixed_parameters
)
```


## Arguments

par
1d vector of initial parameters: lambda, lambda_cov, alpha, alpha_cov, and sigma
fitness 1d vector of fitness observations, in log scale
neigh_intra_matrix
optional matrix of one column, number of intraspecific neighbours for each ob-
servation
neigh_inter_matrix
matrix of arbitrary columns, number of interspecific neighbours for each obser-
vation
covariates optional matrix with observations in rows and covariates in columns. Each cell
is the value of a covariate in a given observation
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","alpha_intra","alpha_inter"

## Value

log-likelihood value

```
RK_pm_alpha_pairwise_lambdacov_none_alphacov_none
    Ricker model with pairwise alphas and no covariate effects
```


## Description

Ricker model with pairwise alphas and no covariate effects

## Usage

```
RK_pm_alpha_pairwise_lambdacov_none_alphacov_none(
    par,
    fitness,
    neigh_intra_matrix = NULL,
    neigh_inter_matrix,
    covariates,
    fixed_parameters
    )
```


## Arguments

par 1d vector of initial parameters: 'lambda', 'alpha_intra' (optional), 'alpha_inter', and 'sigma'
fitness $\quad 1 d$ vector of fitness observations, in log scale
neigh_intra_matrix
optional matrix of one column, number of intraspecific neighbours for each observation
neigh_inter_matrix
matrix of arbitrary columns, number of interspecific neighbours for each observation
covariates included for compatibility, not used in this model
fixed_parameters
optional list specifying values of fixed parameters, with components "lambda","alpha_intra","alpha_inter"

## Value

log-likelihood value

```
RK_project_alpha_global_lambdacov_none_alphacov_none
    Ricker model for projecting abundances, with a global alpha and no
    covariate effects
```


## Description

Ricker model for projecting abundances, with a global alpha and no covariate effects

## Usage

```
RK_project_alpha_global_lambdacov_none_alphacov_none(
        lambda,
        alpha_intra,
        alpha_inter,
        lambda_cov,
        alpha_cov,
        abundance,
        covariates
    )
```


## Arguments

lambda numeric lambda value.
alpha_intra included for compatibility, not used in this model.
alpha_inter single numeric value.
lambda_cov included for compatibility, not used in this model.
alpha_cov included for compatibility, not used in this model.
abundance named numeric vector of abundances in the previous timestep.
covariates included for compatibility, not used in this model.

## Value

numeric abundance projected one timestep

```
RK_project_alpha_none_lambdacov_none_alphacov_none
    Model for projecting abundances, with no alpha and no covariate ef-
    fects
```


## Description

Model for projecting abundances, with no alpha and no covariate effects

## Usage

RK_project_alpha_none_lambdacov_none_alphacov_none( lambda, alpha_intra, alpha_inter, lambda_cov, alpha_cov, abundance, covariates
)

## Arguments

| lambda | numeric lambda value. |
| :--- | :--- |
| alpha_intra | included for compatibility, not used in this model. |
| alpha_inter | included for compatibility, not used in this model. |
| lambda_cov | included for compatibility, not used in this model. |
| alpha_cov | included for compatibility, not used in this model. |
| abundance | named numeric vector of abundances in the previous timestep. |
| covariates | included for compatibility, not used in this model. |

## Value

numeric abundance projected one timestep

```
RK_project_alpha_pairwise_lambdacov_global_alphacov_global
    Ricker model for projecting abundances, with specific alpha values
    and global covariate effects on alpha and lambda
```


## Description

Ricker model for projecting abundances, with specific alpha values and global covariate effects on alpha and lambda

## Usage

```
RK_project_alpha_pairwise_lambdacov_global_alphacov_global(
    lambda,
    alpha_intra,
    alpha_inter,
    lambda_cov,
    alpha_cov,
    abundance,
    covariates
)
```


## Arguments

| lambda | numeric lambda value. |
| :--- | :--- |
| alpha_intra | single numeric value. |
| alpha_inter | numeric vector with interspecific alpha values. |
| lambda_cov | numeric vector with effects of covariates over lambda. |
| alpha_cov | named list of numeric values with effects of each covariate over alpha. |
| abundance | named numeric vector of abundances in the previous timestep. |
| covariates | matrix with observations in rows and covariates in columns. Each cell is the <br> value of a covariate in a given observation. |

## Value

numeric abundance projected one timestep

RK_project_alpha_pairwise_lambdacov_global_alphacov_pairwise Ricker model for projecting abundances, with specific alpha values and global covariate effects on alpha and lambda

## Description

Ricker model for projecting abundances, with specific alpha values and global covariate effects on alpha and lambda

## Usage

RK_project_alpha_pairwise_lambdacov_global_alphacov_pairwise(
lambda,
alpha_intra,
alpha_inter,
lambda_cov,
alpha_cov,
abundance,
covariates
)

## Arguments

| lambda | named numeric lambda value. |
| :--- | :--- |
| alpha_intra | single numeric value. <br> alpha_inter <br> lambda_cov <br> alpha_cov |
| numeric vector with interspecific alpha values. <br> numeric vector with effects of covariates over lambda. <br> named list of named numeric vectors with effects of each covariate over alpha <br> values. |  |
| abundance | named numeric vector of abundances in the previous timestep. <br> covariates |
| matrix with observations in rows and covariates in named columns. Each cell is <br> the value of a covariate in a given observation. |  |

## Value

numeric abundance projected one timestep

```
RK_project_alpha_pairwise_lambdacov_none_alphacov_none
    Ricker model for projecting abundances, with specific alpha values
    and no covariate effects
```


## Description

Ricker model for projecting abundances, with specific alpha values and no covariate effects

## Usage

```
RK_project_alpha_pairwise_lambdacov_none_alphacov_none(
    lambda,
    alpha_intra,
    alpha_inter,
    lambda_cov,
    alpha_cov,
    abundance,
    covariates
    )
```


## Arguments

lambda numeric lambda value.
alpha_intra included for compatibility, not used in this model.
alpha_inter single numeric value.
lambda_cov included for compatibility, not used in this model.
alpha_cov included for compatibility, not used in this model.
abundance named numeric vector of abundances in the previous timestep.
covariates included for compatibility, not used in this model.

## Value

numeric abundance projected one timestep
salinity_list Salinity measurements

## Description

A list containing salinity values associated to the data from 'neigh_list'. The list has 17 elements, one for each focal species considered. Each element of the list is a dataframe with 2 columns:

- obs_ID: unique identifier of each observation
- salinity: salinity measurement for that observation, in accumulated microsiemens/m2

```
Usage
data(salinity_list)
```


## Format

A list with 17 elements, each of which a dataframe of variable number of rows and 2 numeric columns

## Note

For details, see Lanuza et al. 2018 Ecology Letters.

```
spatial_sampling spatial arrangement of the observations
```


## Description

A dataset giving the spatial arrangement of observations. The dataset is a list of 16 elements following the structure of 'neigh_list'. Each list component is a dataframe with columns:

## Usage

data(spatial_sampling)

## Format

A list with 16 elements, each of which a dataframe of variable number of rows and 18 columns

## Details

- obs_ID: unique identifier for each observation
- plot: one of 9 plots of $8.5 \times 8.5 \mathrm{~m}$
- subplot: one of 36 subplots of 1 x 1 m within each plot


## Note

For details, see Lanuza et al. 2018 Ecology Letters.

```
species_fitness Fitness of a species
```


## Description

Calculates the fitness of a species sensu Godoy et al. (2014).Note that its definition is modelspecific, i.e. it depends on the model family from which interaction coefficients were estimated. The function given here assumes a community of $n$-species, so that species fitness is calculated according to a general competitive response (r) substituting the 2 -sp denominator terms of table A1 of Hart et al. 2018. This competitive response can be calculated for a series of species with the function 'cxr_er_fit'.

## Usage

species_fitness(
effect_response_fit = NULL,
lambda = NULL,
competitive_response = NULL,
model_family = NULL
)

## Arguments

effect_response_fit
cxr_er_fit object with valid lambda and response terms.
lambda per capita fecundity of the species in the absence of competition.
competitive_response
parameter reflecting the species' sensitivity to competition.
model_family model family for which to calculate species fitness.

## Details

Thus, the function accepts two sets of parameters. First, a 'cxr_er_fit' object returned from that function. In this case, species fitness will be calculated for all focal taxa included in the 'cxr_er_fit' object.
Otherwise, users may enter a specification of the model to use, as well as lambda and competitive response parameters of a single species.

If no model family is provided, or a model family for which there is no associated ' XX _species_fitness' function, the function resorts to the standard Lotka-Volterra formulation (Hart et al. 2018). Overall, we strongly suggest that you use the standard formulation ONLY if you are completely confident that the model from which you obtained your parameters is consistent with it. Otherwise, you should include your own formulation of species fitness (see vignette 4).

Value
single numeric value/vector, species fitness of one or several taxa
species_rates Species germination and survival rates

## Description

A dataset containing germination and survival rates for 17 plant species. It includes columns with the scientific names and their associated codes.

## Usage <br> data(species_rates)

## Format

A data frame with 17 rows and 4 variables

## Details

- species: binomial name
- code: four-letter code used in other datasets
- germination: germination rate
- seed.survival: annual survival of ungerminated seed in the soil


## Note

For details, see Lanuza et al. 2018 Ecology Letters.
summary.cxr_er_fit CXR summary method for effect response model fits

## Description

CXR summary method for effect response model fits

## Usage

\#\# S3 method for class 'cxr_er_fit'
summary (object, ...)

## Arguments

$$
\begin{array}{ll}
\text { object } & \text { a cxr_er_fit object, from the function with the same name } \\
\ldots & \text { other arguments, not used }
\end{array}
$$

## Value

console output
summary.cxr_pm_fit CXR summary method for population model fits

## Description

CXR summary method for population model fits

## Usage

\#\# S3 method for class 'cxr_pm_fit'
summary (object, ...)

## Arguments

object a cxr_pm_fit object, from the function with the same name

## Value

console output

```
summary.cxr_pm_multifit
CXR summary method for multispecies fits
```


## Description

CXR summary method for multispecies fits

## Usage

\#\# S3 method for class 'cxr_pm_multifit'
summary(object, ...)

## Arguments

object a cxr_pm_multifit object, from the function with the same name ... other arguments, not used

## Value

console output

```
vec_permutation_matrices
                    Generate templates for dispersal, demography, and permutation ma-
                    trices
```


## Description

this follows the vec-permutation approach as defined in: Hunter and Caswell 2005, doi:10.1016/j.ecolmodel.2005.05.002, Ozgul et al. 2009, doi: 10.1086/597225

## Usage

vec_permutation_matrices(num.sp, num.sites, num.stages)

## Arguments

num. sp integer, number of species
num.sites integer, number of sites
num. stages integer, number of stages

## Value

nested list, of the form 'list[[type]][[sp]]', where 'type' is demography, dispersal, or permutation.

## Examples

```
# number of demographic stages - this should be always fixed to 3 for
# compatibility with other functions
num.stages <- 3
num.sp <- 4
num.sites <- 5
vpm <- vec_permutation_matrices(num.sp,num.sites,num.stages)
```

vital_rate Vital rate calculation

## Description

Calculates vital rates from their effect sizes and terms. This is equivalent to predicting from a binomial glm with given coefficients. In this version, the user needs to ensure that 'param' and 'env' match, i.e. that if the 'param' list is defined with environmental forcing, it is passed here, and viceversa. In future versions I may implement checks for that here, but for now, be aware that it will fail.

## Usage

vital_rate(vr, sp, site, param, env = NULL, densities)

## Arguments

vr
sp
site
param
env
densities
integer or char, vital rate to obtain, from the ones defined in 'param'. So far, valid names are "Sj","Sn","Sr","Rn","Rr","D","Ds,"O".
integer or char, species
intger or char, site
param nested list (see 'build_param')
optional numeric, environmental forcing
densities of all sp in the site, including individuals from all three life stages

## Value

numeric value

## Index

```
* datasets competitive_ability,18
    abundance,4 cxr_er_bootstrap,20
    glm_example_coefs, 38
    metapopulation_example_param,58
    neigh_list,58
    salinity_list,70
    spatial_sampling,70
    species_rates,72
abundance, 4
abundance_projection,4
avg_fitness_diff,5
cxr_er_fit,21
cxr_generate_test_data, 24
cxr_pm_bootstrap,27
cxr_pm_fit,28
cxr_pm_multifit,30
densities_to_df, 33
fill_demography_matrix, 34
fill_dispersal_matrix,34
fill_transition_matrix, 35
BH_er_lambdacov_global_effectcov_global_respofisercossgi&mbid, 36
    7
BH_er_lambdacov_none_effectcov_none_responsecgen@mate_vital_rate_coefs, 36
    8 glm_example_coefs,38
BH_pm_alpha_global_lambdacov_none_alphacov_none,
    9 LV_er_lambdacov_global_effectcov_global_responsecov_global
BH_pm_alpha_none_lambdacov_none_alphacov_none,
    38
    9 LV_er_lambdacov_none_effectcov_none_responsecov_none,
BH_pm_alpha_pairwise_lambdacov_global_alphacov_global39
    10 LV_pm_alpha_global_lambdacov_none_alphacov_none,
BH_pm_alpha_pairwise_lambdacov_global_alphacov_pairwi4@,
    11 LV_pm_alpha_none_lambdacov_none_alphacov_none,
BH_pm_alpha_pairwise_lambdacov_none_alphacov_none, 40
    1 2 ~ L V \_ p m \_ a l p h a \_ p a i r w i s e \_ l a m b d a c o v \_ g l o b a l \_ a l p h a c o v \_ g l o b a l , ,
BH_project_alpha_global_lambdacov_none_alphacov_none, 41
    13 LV_pm_alpha_pairwise_lambdacov_global_alphacov_pairwise,
BH_project_alpha_none_lambdacov_none_alphacov_none, 42
    14 LV_pm_alpha_pairwise_lambdacov_none_alphacov_none,
BH_project_alpha_pairwise_lambdacov_global_alphacov_g4dbal,
    14 LV_project_alpha_global_lambdacov_none_alphacov_none,
BH_project_alpha_pairwise_lambdacov_global_alphacov_p44 rwise,
    15 LV_project_alpha_none_lambdacov_none_alphacov_none,
BH_project_alpha_pairwise_lambdacov_none_alphacov_none5
    16
build_param,17
calculate_densities,18
LV_project_alpha_pairwise_lambdacov_global_alphacov_global
LV_project_alpha_pairwise_lambdacov_global_alphacov_pairwi
    4 6
```



```
    4 7
                            6 7
LW_er_lambdacov_global_effectcov_global_respof||efoodjgdobalpha_pairwise_lambdacov_global_alphacov_global
    4 867
LW_er_lambdacov_none_effectcov_none_responsecßK_mo@\dot{éct_alpha_pairwise_lambdacov_global_alphacov_pairwi}
    4 9
                                    6 8
LW_pm_alpha_global_lambdacov_none_alphacov_nomik,project_alpha_pairwise_lambdacov_none_alphacov_none,
    4 9
                            6 9
LW_pm_alpha_none_lambdacov_none_alphacov_none,
    50
        salinity_list,70
LW_pm_alpha_pairwise_lambdacov_global_alphacospagib&lag,ampling, 70
    51 species_fitness,71
LW_pm_alpha_pairwise_lambdacov_global_alphacosppgies{irates, 72
    5 2 ~ s u m m a r y . c x r \_ e r \_ f i t , 7 3
LW_pm_alpha_pairwise_lambdacov_none_alphacov_sumpmary.cxr_pm_fit,73
    5 3 ~ s u m m a r y . c x r \_ p m \_ m u l t i f i t , 7 4
LW_project_alpha_global_lambdacov_none_alphacov_none,
    54
                                    ve\overline{c}_permutation_matrices, 74
LW_project_alpha_none_lambdacov_none_alphacovital_rate, 75
    54
LW_project_alpha_pairwise_lambdacov_global_alphacov_global,
    5 5
LW_project_alpha_pairwise_lambdacov_global_alphacov_pairwise,
    56
LW_project_alpha_pairwise_lambdacov_none_alphacov_none,
    5 7
```

metapopulation_example_param, 58
neigh_list, 58
niche_overlap, 59
RK_er_lambdacov_global_effectcov_global_responsecov_global,
60
RK_er_lambdacov_none_effectcov_none_responsecov_none,
61
RK_pm_alpha_global_lambdacov_none_alphacov_none,
62
RK_pm_alpha_none_lambdacov_none_alphacov_none,
62
RK_pm_alpha_pairwise_lambdacov_global_alphacov_global,
63
RK_pm_alpha_pairwise_lambdacov_global_alphacov_pairwise,
64
RK_pm_alpha_pairwise_lambdacov_none_alphacov_none,
65
RK_project_alpha_global_lambdacov_none_alphacov_none,
66

