# Package 'gauseR'

November 28, 2021

Title Lotka-Volterra Models for Gause's 'Struggle for Existence'

Version 1.1

Description A collection of tools and data for analyzing the Gause microcosm experi-

ments, and for fitting Lotka-Volterra models to time series data. Includes methods for fitting single-species logistic growth, and multi-species interaction models, e.g. of competition, predator/prey relationships, or mutualism. See documentation for individual functions for examples. In general, see the lv\_optim() function for examples of how to fit parameter values in multispecies systems. Note that the general methods applied here, as well as the form of the differential equations that we use, are described in detail in the Quantitative Ecology textbook by Lehman et al., avail-

able at <http://hdl.handle.net/11299/204551>, and in Lina K. Mühlbauer, Maximilienne Schulze, W. Stanley Harpole, and Adam T. Clark. 'gauseR': Simple methods for fitting Lotka-Volterra models describing Gause's 'Struggle for Existence' in the journal Ecology and Evolution.

Imports deSolve, stats, graphics

Suggests knitr, rmarkdown

VignetteBuilder knitr

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gauseR

gauseR: Simple methods for fitting Lotka-Volterra models describing Gause's "Struggle for Existence"

#### Description

A collection of tools and data for analyzing the Gause microcosm experiments, and for fitting Lotka-Volterra models to time series data. Includes methods for fitting single-species logistic growth, and multi-species interaction models, e.g. of competition, predator/prey relationships, or mutualism. See documentation for individual functions for examples. In general, see the lv\_optim() function for examples of how to fit parameter values in multi-species systems.

#### Authors

Adam Clark, Lina Muehlbauer, and Maximilienne Schulze.

# Applications

Note that the general methods applied here, as well as the form of the differential equations that we use, are described in detail in the Quantitative Ecology textbook by Lehman et al., cited below. Using the default functions, species dynamics therefore follow the form:

```
dni/dt = ni * (ri + aii * ni + sum_j(aij * nj))
```

# Source

Clarence Lehman, Shelby Loberg, and Adam T. Clark. (2019). Quantitative Ecology: A New Unified Approach. University of Minnesota Libraries Publishing. University of Minnesota Digital Conservancy, http://hdl.handle.net/11299/204551

Lina K. Muehlbauer, Maximilienne Schulze, W. Stanley Harpole, and Adam T. Clark. "gauseR: Simple methods for fitting Lotka-Volterra models describing Gause's 'Struggle for Existence'." Ecology and Evolution.

# Examples

?lv_interaction_log	<pre># a version of the lv_interaction computed in log abundance space,</pre>
	<pre># which typically works better for optimization</pre>
?lv_optim	<pre># methods for fitting complex n-species models</pre>
?test_goodness_of_fit	<pre># tests goodness of fit of model results, with an R2-like statistic</pre>
?ode_prediction	<pre># generic function for simulating time series,</pre>
	<pre># to be used with other optimizers</pre>

gause\_1931\_AmN\_f01

Growth of population of the flour beetle Tribolium confusum in 16 and 64 grams of flour

# Description

A dataset containing the abundance of Tribolium confusum, grown in monoculture with different quantities of food (flour). Gause's goal was to determine the influence of ecological factors on population growth.

## Usage

gause\_1931\_AmN\_f01

## Format

A data frame with 18 rows and 6 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Species Name of Species: Tribolium confusum

Time Day of experiment

Individuals Number of Individuals

Treatment Treatments: 16 and 64g flour per starting beetle pair

#### Source

Gause (1931) The influence of ecological factors on the size of population. The American Naturalist 65:696, 70-76

4

gause\_1931\_AmN\_f02

The influence of quantity of food on the asymptotic population of Triboliurn confusum.

# Description

A dataset containing the population size of Tribolium confusum at the upper asymptote, as a characteristic for the population beeing in equilibrium, under different quantities of food (flour). Gause's goal was to determine the influence of ecological factors on population growth.

#### Usage

gause\_1931\_AmN\_f02

# Format

A data frame with 4 rows and 5 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Species Name of Species: Tribolium confusum

Individuals Number of Individuals at population equlibrium

Treatment Treatments: 16, 32, 64 and 128g flour per starting beetle pair

# Source

Gause (1931) The influence of ecological factors on the size of population. The American Naturalist 65:696, 70-76

gause\_1931\_AmN\_f03 The influence of temperature on the asymptotic population of Moina macrocopa.

# Description

A dataset containing the population size of Moina macrocopa at the upper asymptote, as a characteristic for the population beeing in equilibrium, under different temperatures. Gause's goal was to determine the influence of ecological factors on population growth.

#### Usage

gause\_1931\_AmN\_f03

A data frame with 3 rows and 5 variables:

**Paper** Paper from which data are drawn

Figure Figure number in paper

Species Name of Species: Tribolium confusum

Individuals Number of Individuals at population equilbrium

Treatment Treatments:20, 25 and 35 degrees Celcius

#### Source

Gause (1931) The influence of ecological factors on the size of population. The American Naturalist 65:696, 70-76

gause\_1932\_QR\_t05 The influence of temperature on the growth of the yeast Saccharomyces cerevisiae

# Description

A dataset containing the abundance of Saccharomyces cerevisiae, grown in different temperatures (5.7 to 41 degrees C). The Volume is measured in "Amount of yeast", which refers to a standardized index, based on the alcohol production per Unit of Yeast.

# Usage

gause\_1932\_QR\_t05

#### Format

A data frame with 102 rows and 7 variables:

Paper Paper from which data are drawn

Table Table number in paper

**Experiment** Number of series: 1,2 or 3

Species Name of Species: Saccharomyces cerevisiae

Time Hour of experiment

Volume "Amount of Yeast" of Species

**Treatment** Treatments: Temperatures (5.7 to 41 degrees C)

# Source

Gause (1932) Ecology of Populations. The Quarterly Review of Biology 7, vol. 1, pp. 27-46

gause\_1934\_book\_app\_t01

Raw data on the abundances and volumes of Saccharomyces cerevisiae and Schizosaccharomyces kephir

# Description

A dataset containing the abundances (number of cells) and volumes of Saccharomyces cerevisiae and Schizosaccharomyces kephir grown in monoculture and in mixture. The data for two experiments with different time periods is reported.

# Usage

gause\_1934\_book\_app\_t01

## Format

A data frame with 60 rows and 10 variables:

Paper Paper from which data are drawn

Table Table number in paper

Experiment Number of experiment: 1 or 2

Species Name of Species: Saccharomyces cerevisiae and Schizosaccharomyces kephir

Time Day of experiment

Volume\_Species Volume of yeast species in the mixture, estimated from counted cells

Squares Number of Squares counted

Individuals\_Square Average Number of cells per Square

Volume\_Total Total volume of yeast

Treatment Treatments: Monoculture and Mixture

# Source

gause\_1934\_book\_app\_t02

Alcohol production of Saccharomyces cerevisiae and Schizosaccharomyces kephir

#### Description

A dataset containing the alcohol production of Saccharomyces cerevisiae and Schizosaccharomyces kephir cultivated under anaerobic and aerobic conditions, in percent and per unit of yeast volume.

## Usage

gause\_1934\_book\_app\_t02

# Format

A data frame with 28 rows and 7 variables:

Paper Paper from which data are drawn

Table Table number in paper

Species Name of Species: Saccharomyces cerevisiae, Schizosaccharomyces kephir

**Time** Hour of experiment

Alcohol Alcohol production in percent

Alcohol.Yeast\_Volume Alcohol production per unit of yeast volume

Treatment Treatments: Anaerobic and Aerobic

#### Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_app\_t03

Raw data of Paramecium caudatum and Paramecium aurelia grown in Monoculture and Mixture

# Description

A dataset containing the mean abundances of Paramecium caudatum and Paramecium aurelia, grown in mixture and monoculture. This dataset contains the mean abundances of three experiments. Note, that for day 20, single values are missing and only the mean is reported.

#### Usage

gause\_1934\_book\_app\_t03

A data frame with 104 rows and 6 variables:

Paper Paper from which data are drawn
Table Table number in paper
Species Name of Species: Paramecium caudatum and Paramecium aurelia
Time Day of experiment
Individuals Number of Individuals
Treatment Treatments: Monoculture and Mixture

## Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_app\_t04

Raw data of Paramecium caudatum and Paramecium aurelia grown on different media

# Description

A dataset containing the abundance of Paramecium caudatum and Paramecium aurelia, grown in monoculture and in mixture on buffered medium with high wild bacteria concentration ("one loop" medium) and low wild bacteria concentration ("half loop" medium). The number of individuals is reported for monoculture and the number of individuals of the species in the mixture.

#### Usage

gause\_1934\_book\_app\_t04

## Format

A data frame with 68 rows and 7 variables:

Paper Paper from which data are drawn

Table Table number in paper

Species Name of Species: Paramecium caudatum and Paramecium aurelia

Time Day of experiment

Individuals Number of Individuals

Individuals\_Mixture Number of Individuals per Species in Mixture

Treatment Treatments: One Loop and Half Loop

# Source

gause\_1934\_book\_app\_t05

Raw data of Stylonychia pustulata in monoculture and mixture with and Paramecium aurelia and P. caudatum

# Description

A dataset containing the abundances of Stylonychia pustulata grown in monoculture and mixture with P. aurelia and P. caudatum and the abundances of P. aurelia and P. caudatum in this mixtures on the medium of Osterhout. This dataset contains the raw data of three different experiments: Stylonychia pustulata grown in monoculture, in mixture with P.aurelia and in mixture wit P.caudatum. The abundances of three to five cultures per experiment and the calculated mean abundances of the experiments are reported. Note, that for day 20 - 25, single values are missing and only the mean is reported.

#### Usage

gause\_1934\_book\_app\_t05

# Format

A data frame with 575 rows and 8 variables:

Paper Paper from which data are drawn

Table Table number in paper

Time Day of experiment

Experiment Number of experiment: 1,2 or 3

Culture Number of culture: 1,2,3,4,5 or Mean

Species Name of Species: Stylonychia pustulata, Paramecium caudatum and Paramecium aurelia

Individuals Number of Individuals

Treatment Treatments: Monoculture and Mixture P.aurelia and Mixture P.caudatum

#### Source

gause\_1934\_book\_f04 Growth of Paramecium caudatum

# Description

A dataset containing the growth in abundance of Paramecium caudatum over six days.

# Usage

gause\_1934\_book\_f04

# Format

A data frame with 8 rows and 5 variables:

Paper Paper from which data are drawnFigure Figure number in paperSpecies Name of Species: Paramecium caudatumTime Day of experimentIndividuals Number of Individuals

#### Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_f09 Growth of Saccharomyces cerevisiae

## Description

A dataset containing the growth in volume of Saccharomyces cerevisiae under anaerobic conditions. The Volume is measured in "Amount of yeast", which refers to a standardized index, based on the alcohol production per Unit of Yeast.

#### Usage

gause\_1934\_book\_f09

#### Format

A data frame with 9 rows and 5 variables:

Paper Paper from which data are drawn
Figure Figure number in paper
Species Name of Species: Saccharomyces cerevisiae
Time Hour of experiment
Volume "Amount of Yeast" of Species

#### Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_f10 Growth of Saccharomyces cerevisiae with medium change

# Description

A dataset containing the abundance of Saccharomyces cerevisiae cultivated in monoculture with a medium change in different time periods under anaerobic conditions. The Number of Individuals is measured in Number of cells per 1/250 mm3.

#### Usage

gause\_1934\_book\_f10

# Format

A data frame with 29 rows and 6 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Species Name of Species: Saccharomyces cerevisiae

Time Hour of experiment

Individuals Number of cells per 1/250 mm3

Treatment Treatments: medium change every 3, 12, 24 hours and control

#### Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_f11 Gause Yeast Data

# Description

A dataset containing yeast volume and alcohol concentration for two replicates of S. cerevisiae grown in monoculture.

#### Usage

gause\_1934\_book\_f11

A data frame with 11 rows and 7 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Species Species name

**Time** Experiment time in hours

Alcohol\_Percent Alcohol concentration, in percent

Yeast\_Volume Yeast volume, listed in papers as 'amount'

Treatment Experiment replicate, under two 'somewhat different' growth medium concentrations

#### Source

Huffaker (1958) Experimental Studies on Predation: Dispersion Factors and Predator-Prey Oscillations. Hilgardia 27: 343-83.

gause\_1934\_book\_f12 Growth of Saccharomyces cerevisiae with additional alcohol

# Description

A dataset containing the growth in volume of Saccharomyces cerevisiae cultivated under anaerobic conditions with added alcohol.Gause measured the effect of alcohol on reaching a saturated population. The saturation is measured in percent of the satured population grown without additional alcohol.

## Usage

gause\_1934\_book\_f12

#### Format

A data frame with 6 rows and 5 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Species Name of Species: Saccharomyces cerevisiae

Alcohol Additional alcohol in percent

Population Percentage of saturated population reached

## Source

gause\_1934\_book\_f13

Growth of Saccharomyces cerevisiae and Schizosaccaromyces kephir in mixed population

## Description

A dataset containing the the growth in volume of Saccharomyces cerevisiae and Schizosaccaromyces kephir cultivated in the mixed population (two series of experiments) under anaerobic conditions. Gause also measured the volume of the mixed population. The Volume is measured in "Amount of yeast", which refers to a standardized index, based on the alcohol production per Unit of Yeast.

#### Usage

gause\_1934\_book\_f13

# Format

A data frame with 47 rows and 5 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Species Name of Species: Saccharomyces cerevisiae, Schizosaccaromyces kephi and Mixed Population

Time Hour of experiment

Volume "Amount of Yeast" of Species

#### Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_f14 Growth of Saccharomyces cerevisiae in mixed population

# Description

A dataset containing the the growth in volume of Saccharomyces cerevisiae cultivated separately and in the mixed population (two series of experiments) with Schizosaccaromyces kephir under anaerobic conditions. The Volume is measured in "Amount of yeast", which refers to a standardized index, based on the alcohol production per Unit of Yeast.

#### Usage

gause\_1934\_book\_f14

A data frame with 32 rows and 6 variables:

**Paper** Paper from which data are drawn

Figure Figure number in paper

Species Name of Species: Saccharomyces cerevisiae

Time Hour of experiment

Volume "Amount of Yeast" of Species

Treatment Treatments: Monoculture and Mixture

# Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_f15 Growth of Schizosaccaromyces kephir in mixed population - anaerobic

## Description

A dataset containing the the growth in volume of Schizosaccaromyces kephir cultivated separately and in the mixed population (two series of experiments) with Saccharomyces cerevisiae under anaerobic conditions. The Volume is measured in "Amount of yeast", which refers to a standardized index, based on the alcohol production per Unit of Yeast.

# Usage

gause\_1934\_book\_f15

# Format

A data frame with 24 rows and 6 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Species Name of Species: Schizosaccaromyces kephir

Time Hour of experiment

Volume "Amount of Yeast" of Species

Treatment Treatments: Monoculture and Mixture

## Source

gause\_1934\_book\_f16

Growth of Schizosaccaromyces kephir and Saccharomyces cerevisiae in mixed population -aerobic

# Description

A dataset containing the the growth in volume of Schizosaccaromyces kephir cultivated separately and in the mixed population with Saccharomyces cerevisiae under aerobic conditions. The Volume is measured in "Amount of yeast", which refers to a standardized index, based on the alcohol production per Unit of Yeast.

#### Usage

gause\_1934\_book\_f16

#### Format

A data frame with 27 rows and 6 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Species Name of Species: Schizosaccaromyces kephir, Saccharomyces cerevisiae and Total\_Yeast

Time Hour of experiment

Volume "Amount of Yeast" of Species

Treatment Treatments: Monoculture and Mixture

#### Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_f18 Growth of Paramecium caudatum and Stylonychia mytilis in Mixture

# Description

A dataset containing the abundance of Paramecium caudatum and Stylonychia mytilis, grown in monoculture and in mixture on buffered medium without wild bacteria.

#### Usage

gause\_1934\_book\_f18

A data frame with 28 rows and 6 variables:

Paper Paper from which data are drawn
Figure Figure number in paper
Species Name of Species: Paramecium caudatum and Stylonychia mytilis
Time Day of experiment
Individuals Number of Individuals

Treatment Treatments: Monoculture and Mixture

#### Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_f19 Growth of Paramecium caudatum and Stylonychia mytilis in Mixture with Wild bacteria

# Description

A dataset containing the abundance of Paramecium caudatum and Stylonychia mytilis, grown in monoculture and in mixture on buffered medium containing wild bacteria.

#### Usage

gause\_1934\_book\_f19

# Format

A data frame with 17 rows and 6 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Species Name of Species: Paramecium caudatum and Stylonychia mytilis

Time Day of experiment

Individuals Number of Individuals

Treatment Treatments: Monoculture and Mixture

# Source

gause\_1934\_book\_f21

# Description

A dataset containing the abundances and the volume of Paramecium caudatum and Paramecium aurelia, to determine the differences in reaching the saturating population regarding Volume and Number of individuals. "Volume" refers to a standardized index, meant to make the abundances of species comparable based on their relative sizes.

# Usage

gause\_1934\_book\_f21

# Format

A data frame with 87 rows and 6 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Species Name of Species: Paramecium caudatum and Paramecium aurelia

Time Day of experiment

Volume "Volume" of Species

Individuals Number of Individuals

## Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_f22 Paramecium competition experiment

#### Description

A dataset containing the abundances of two Paramecium species grown in monoculture and mixture. Note, is for the same experiment as gause\_1934\_science\_f02\_03, except that data were digitized separately, and therefore have small variations. These might be useful for estimating observation error in the data digitization process. "Volume" refers to a standardized index, meant to make the abundances of species comparable based on their relative sizes.

#### Usage

gause\_1934\_book\_f22

A data frame with 72 rows and 4 variables:

Paper Paper from which data are drawn
Figure Figure number in paper
Day Day of experiment
Species1 Name of Species 1
Volume\_Species1 Volume of Paramecium caudatum
Species2 Name of Species 2
Volume\_Species2 Volume of Paramecium aurelia
Treatment Treatments: Pa and Pc moncultures, or mixture

#### Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_f23 Growth of Paramecium caudatum and Paramecium aurelia with wild bacteria

#### Description

A dataset containing the volume of Paramecium caudatum and Paramecium aurelia, grown in monoculture on buffered medium with two different wild bacteria concentrations ("one" and "half loop"). "Volume" refers to a standardized index, meant to make the abundances of species comparable based on their relative sizes.

# Usage

gause\_1934\_book\_f23

#### Format

A data frame with 61 rows and 6 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Species Name of Species: Paramecium caudatum and Paramecium aurelia

**Time** Day of experiment

Volume "Volume" of Species

Treatment Treatments: One Loop and Half Loop

#### Source

gause\_1934\_book\_f24

*Growth of Paramecium caudatum and Paramecium aurelia in Mixture on Half Loop Medium* 

#### Description

A dataset containing the volume of Paramecium caudatum and Paramecium aurelia, grown in monoculture and in mixture on buffered medium with low wild bacteria concentration ("half loop" medium). "Volume" refers to a standardized index, meant to make the abundances of species comparable based on their relative sizes.

## Usage

gause\_1934\_book\_f24

#### Format

A data frame with 56 rows and 6 variables:

**Paper** Paper from which data are drawn

Figure Figure number in paper

Species Name of Species: Paramecium caudatum and Paramecium aurelia

**Time** Day of experiment

Volume "Volume" of Species

Treatment Treatments: Monoculture and Mixture

#### Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_f25 Growth of Paramecium caudatum and Paramecium aurelia in Mixture on One Loop Medium

# Description

A dataset containing the volume of Paramecium caudatum and Paramecium aurelia, grown in monoculture and in mixture on buffered medium with high wild bacteria concentration ("one loop" medium). "Volume" refers to a standardized index, meant to make the abundances of species comparable based on their relative sizes.

# Usage

gause\_1934\_book\_f25

A data frame with 57 rows and 6 variables:

Paper Paper from which data are drawn
Figure Figure number in paper
Species Name of Species: Paramecium caudatum and Paramecium aurelia
Time Day of experiment
Volume "Volume" of Species

Treatment Treatments: Monoculture and Mixture

#### Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_f26 Growth of Stylorzychia pustulata in Monoculture and in Mixture

## Description

A dataset containing the abundances of Stylorzychia pustulata cultivated separately, and in the mixed populations with Paramecium caudatum and Paramecium aurelia and the abundances of P. caudatum and P. aurelia grown in mixture with S. pustulata.

## Usage

gause\_1934\_book\_f26

#### Format

A data frame with 104 rows and 6 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Species Name of Species:Paramecium caudatum, Paramecium aurelia and Stylorzychia pustulata

Time Day of experiment

Individuals Number of Individuals

Treatment Treatments: Monoculture and Mixture(with Species)

#### Source

gause\_1934\_book\_f28

# Description

A dataset containing the abundances of Paramecium caudatum and Didinium nasutum grown in mixture. Didinium was introduced at day two. "Individuals" refers to the number of individuals pro 0.5 c.c..

#### Usage

gause\_1934\_book\_f28

#### Format

A data frame with 12 rows and 6 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Species Name of Species: Paramecium caudatum and Didinium nasutum

Time Day of experiment

Individuals Number of Individuals

Treatment Treatments: Monoculture and Mixture

#### Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_f29 Paramecium/Didinium predator-prey experiment

# Description

A dataset containing the abundances of Paramecium caudatum and Didinium nasutum grown in mixture. Didinium was introduced at different days. "Volume" refers to a standardized index, meant to make the abundances of species comparable based on their relative sizes.

#### Usage

gause\_1934\_book\_f29

A data frame with 62 rows and 7 variables:

Paper Paper from which data are drawn
Figure Figure number in paper
Species Name of Species: Didinium nasutum and Paramecium caudatum
Time Day of experiment
Volume Volume of Didinium
Individuals Number of Individuals
Treatment Treatments: D. nasutum introduced after 0, 24, 36 and 48 hrs

## Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_f30 The elementary interaction between Didinium nasutum and Paramecium caudatum

#### Description

A dataset containing the abundances of Didinium nasutum and Paramecium caudatum grown in mixture on the medium of Osterhout.

# Usage

gause\_1934\_book\_f30

## Format

A data frame with 16 rows and 6 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Time Day of experiment

Species Name of Species: Didinium nasutum and Paramecium caudatum

Individuals Number of Individuals

Treatment Treatment: Osterhout medium

# Source

gause\_1934\_book\_f31

## Description

A dataset containing the abundances of Didinium nasutum and Paramecium caudatum grown in mixture on oat medium with sediment.

#### Usage

gause\_1934\_book\_f31

# Format

A data frame with 12 rows and 6 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

**Time** Day of experiment

Species Name of Species: Didinium nasutum and Paramecium caudatum

Individuals Number of Individuals

Treatment Treatment: Oat medium

# Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_f32 Didinium/Paramecium predator/prey experiment

# Description

A dataset containing the abundances of Paramecium caudatum and Didinium nasutum grown in mixture. Note, is for the same experiment as gause\_1934\_science\_f01, except that data were digitized separately, and therefore have small variations. These might be useful for estimating observation error in the data digitization process.

#### Usage

gause\_1934\_book\_f32

A data frame with 17 rows and 8 variables:

PaperPaper from which data are drawnFigureFigure number in paperDayDay of experimentPreyName of Prey SpeciesIndividuals\_PreyNumber of Prey IndividualsPredatorName of Predator SpeciesIndividuals\_PredatorNumber of Predator IndividualsIndividuals\_PredatorNumber of Predator IndividualsImmigrationIs immigration occurring in this time-step? (yes or no)

#### Source

Gause (1934) The Struggle for Existence. Dover Publications; Reprint edition.

gause\_1934\_book\_f39.1 The interaction between Paramecium bursaria and Schizosaccharomyces pombe

# Description

A dataset containing the abundances of Paramecium bursaria and Schizosaccharomyces pombe grown in mixture.

#### Usage

gause\_1934\_book\_f39.1

# Format

A data frame with 36 rows and 5 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Time Day of experiment

**Species** Name of Species: Paramecium bursaria and Schizosaccharomyces pombe **Individuals** Number of Individuals

#### Source

gause\_1934\_science\_f01

Didinium/Paramecium predator/prey experiment

# Description

A dataset containing the abundances of Paramecium caudatum and Didinium nasutum grown in mixture. Note, is for the same experiment as gause\_1934\_book\_f32, except that data were digitized separately, and therefore have small variations. These might be useful for estimating observation error in the data digitization process.

#### Usage

gause\_1934\_science\_f01

#### Format

A data frame with 17 rows and 3 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Day Day of experiment

Prey Name of Prey Species

Individuals\_Prey Number of Prey Individuals

Predator Name of Predator Species

Individuals\_Predator Number of Predator Individuals

**Immigration** Is immigration occurring in this time-step? (yes or no)

#### Source

Gause (1934) Experimental Analysis of Vito Volterra's Mathematical Theory of the Struggle for Existence. Science 79:16-17.

gause\_1934\_science\_f02\_03

Paramecium competition experiment

#### Description

A dataset containing the abundances of two Paramecium species grown in monoculture and mixture. Note, is for the same experiment as gause\_book\_1934\_f22, except that data were digitized separately, and therefore have small variations. These might be useful for estimating observation error in the data digitization process. "Volume" refers to a standardized index, meant to make the abundances of species comparable based on their relative sizes.

# Usage

gause\_1934\_science\_f02\_03

#### Format

A data frame with 63 rows and 4 variables:

Paper Paper from which data are drawn
Figure Figure number in paper
Day Day of experiment
Species1 Name of Species 1
Volume\_Species1 Volume of Paramecium caudatum
Species2 Name of Species 2
Volume\_Species2 Volume of Paramecium aurelia
Treatment Treatments: Pa and Pc moncultures, or mixture

# Source

Gause (1934) Experimental Analysis of Vito Volterra's Mathematical Theory of the Struggle for Existence. Science 79:16-17.

gause_1936_AnE_f01	Interaction between predators (Cheyletus eruditus) and prey (Aleuiro-
	glyphus agilis) on millet, wheat flour and a mixture of these sub-
	stances.

#### Description

A dataset containing the abundance of Cheyletus eruditus and Aleuiroglyphus agilis, as apredatorprey system under different food conditions for the prey (wheat and millet). Gause's goal was to determine the influence of ecological factors on predator-prey dynamics. The number of individuals is reported as Individuals per 0.2 g prey food.

# Usage

gause\_1936\_AnE\_f01

# Format

A data frame with 34 rows and 6 variables:

Paper Paper from which data are drawn
Figure Figure number in paper
Time Day of experiment
Species Name of Species: Cheyletus eruditus and Aleuiroglyphus agilis
Individuals Number of Individuals per 0.2 g
Treatment Treatments: Wheat, Millet, Wheat+Millet

#### Source

Gause (1936) Further Studies of Interaction between Predators and Prey. Journal of Animal Ecology, vol. 5, pp. 1-18.

gause\_1936\_AnE\_f03.1 Interaction between predators (Cheyletus eruditus) and prey (Aleuroglyphus agilis) with occasional immigration

# Description

A dataset containing the abundance of Cheyletus eruditus and Aleuiroglyphus agilis, as apredatorprey system with an occasional immigration of the prey on the 63rd day. The number of individuals is reported as Individuals per 0.2 g prey food. Wheat flour was used as food.

# Usage

gause\_1936\_AnE\_f03.1

# Format

A data frame with 22 rows and 6 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Time Day of experiment

Species Name of Species: Cheyletus eruditus and Aleuiroglyphus agilis

Individuals Number of Individuals per 0.2 g

Treatment Treatments: Immigration or NA

#### Source

Gause (1936) Further Studies of Interaction between Predators and Prey. Journal of Animal Ecology, vol. 5, pp. 1-18.

gause\_1936\_AnE\_f03.3a Interaction between predators (Cheyletus eruditus) and prey (Aleuroglyphus agilis) with artifical immigration

## Description

A dataset containing the abundance of Cheyletus eruditus and Aleuiroglyphus agilis, as apredatorprey system with an artificial everyday immigration of predator and prey. The number of individuals is reported as Individuals per 0.2 g prey food. Wheat flour was used as food.

#### Usage

gause\_1936\_AnE\_f03.3a

# Format

A data frame with 24 rows and 6 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

**Time** Day of experiment

Species Name of Species: Cheyletus eruditus and Aleuiroglyphus agilis

**Individuals** Number of Individuals per 0.2 g

Treatment Treatment: Immigration

#### Source

Gause (1936) Further Studies of Interaction between Predators and Prey. Journal of Animal Ecology, vol. 5, pp. 1-18.

gause\_1936\_AnE\_f03.3b Interaction between predators (Cheyletus eruditus) and prey (Aleuroglyphus agilis) with artifical immigration

# Description

A dataset containing the abundance of Cheyletus eruditus and Aleuiroglyphus agilis, as apredatorprey system with an artificial immigration of predator and prey every 8th day. The number of individuals is reported as Individuals per 0.2 g prey food. Wheat flour was used as food.

#### Usage

gause\_1936\_AnE\_f03.3b

A data frame with 26 rows and 6 variables:

Paper Paper from which data are drawn
Figure Figure number in paper
Time Day of experiment
Species Name of Species: Cheyletus eruditus and Aleuiroglyphus agilis
Individuals Number of Individuals per 0.2 g
Treatment Treatment: Immigration

# Source

Gause (1936) Further Studies of Interaction between Predators and Prey. Journal of Animal Ecology, vol. 5, pp. 1-18.

gause\_1936\_AnE\_t02 Interaction between predators (Cheyletus eruditus) and prey (Aleuroglyphus agilis)-Raw data

#### Description

A dataset containing the abundance of Cheyletus eruditus and Aleuiroglyphus agilis, as a predatorprey system on semoletta and wheat flour. A fraction at the beginning of each experiment shows the initial relation between predators and prey; e.g. 10/5 means 10 prey and 5 predators (reported in variable "Treatment"). Dataset includes age structured population abundances. The number of individuals is reported as Individuals per 0.2 g prey food. Wheat flour was used as food.

#### Usage

gause\_1936\_AnE\_t02

#### Format

A data frame with 191 rows and 12 variables:

Paper Paper from which data are drawn

 Table
 Table number in paper

Time Day of experiment

Species Name of Species: Cheyletus eruditus and Aleuiroglyphus agilis

Total\_Individuals Number of all Individuals of Species per 0.2 g

Female Number of female individuals per 0.2 g

Male Number of male individuals per 0.2 g

**Imago** Number of adult Individuals per 0.2 g

Hexapod Number of hexapod larvae stage Individuals per 0.2 g

Octapod Number of octapod larvae stage Individuals per 0.2 g

Initial\_Fraction Initial Number of prey/predator

Treatment Treatments: Wheat and Semoletta

#### Source

Gause (1936) Further Studies of Interaction between Predators and Prey. Journal of Animal Ecology, vol. 5, pp. 1-18.

gause\_1936\_AnE\_t03 Interaction between predators (Paramecium bursaria) and prey (Saccharomyces exiguus)

# Description

A dataset containing the abundance of Paramecium bursaria and Saccharomyces exiguus, as a predator-prey system. The primary difference among experimental replicates is the initial abundance of the two species. The number of individuals is reported as Individuals per 0.5 cm3 for Paramecium, and as number of individuals per 0.1 cm3 for Saccharomyces.

#### Usage

gause\_1936\_AnE\_t03

# Format

A data frame with 266 rows and 6 variables:

Paper Paper from which data are drawn

Table Table number in paper

**Experiment** Experimental replicate number

Time Day of experiment

Species Name of Species: Cheyletus eruditus and Aleuiroglyphus agilis

Individuals Number of all Individuals of Species per 0.2 g

#### Source

Gause (1936) Further Studies of Interaction between Predators and Prey. Journal of Animal Ecology, vol. 5, pp. 1-18.

gause\_wrapper

#### Description

Automatically runs routine for finding starting values and optimal parameter values for a Lotka-Volterra interaction system. Using the default functions, species dynamics follow the form  $dni/dt = ni * (ri + aii * ni + sum_j(aij * nj))$  where ri are the elements of vector r, and aij are the elements of matrix A.

# Usage

```
gause_wrapper(
   time,
   species,
   N_starting = NULL,
   r_starting = NULL,
   A_starting = NULL,
   doplot = TRUE,
   keeptimes = FALSE,
   parm_signs = NULL,
   doopt = TRUE,
   ...
)
```

# Arguments

time	Vector of time steps corresponding to observations in species data.frame.
species	A data frame with one column per species to be fitted. Note - column names cannot include white spaces or non-standard special characters.
N_starting	Optional starting values for initial abundances.
r_starting	Optional starting values for species growth rates. If a value is set to zero, it #forces that parameter to zero in the fitting. Values of NA are ignored. Defaults to NULL (no starting values).
A_starting	Optional starting values for species interaction coefficients. If a value is set to zero, it #forces that parameter to zero in the fitting. Values of NA are ignored. Defaults to NULL (no starting values).
doplot	Logical. Should the resulting model be plotted? Defaults to TRUE.
keeptimes	Should predictions be given for the points in the "time" vector, or for a list of 100 evenly spaced time points? Defaults to FALSE.
parm_signs	Optional variable specifying signs for parameters. Defaults to NULL (automatically selected).
doopt	Should optimizer be used (if TRUE), or should the initial linearized estimates by applied (if FALSE)? Defaults to TRUE.
	Optional additional arguments to be passed to ode and optim functions.

# get\_lag

# Value

A list with simulated time series (out), paramter estimates (parameter\_intervals), optimizer output (optout), and raw data used for fitting (rawdata).

## Examples

```
#load competition data
data("gause_1934_science_f02_03")
#subset out data from species grown in mixture
mixturedat<-gause_1934_science_f02_03[gause_1934_science_f02_03$Treatment=="Mixture",]
#extract time and species data
time<-mixturedat$Day
species<-data.frame(mixturedat$Volume_Species1, mixturedat$Volume_Species2)
colnames(species)<-c("P_caudatum", "P_aurelia")
#run wrapper
gause_out<-gause_wrapper(time=time, species=species)</pre>
```

get\_lag

Get timelag

# Description

Calculates time-lagged observations for variable x, separated by treatment.

# Usage

get\_lag(x, time, tau = 1, treatment = NULL, mindt = 0, maxdt = Inf)

# Arguments

x	The time series from which time lagged observations are desired (e.g. popula- tion sizes)
time	The time steps corresponding to each observation
tau	Number of time steps to use between lagged components - defaults to 1
treatment	An optional vector of treatment conditions - time lags will only be computed separately within treatments - defaults to NULL (i.e. no treatments)
mindt	Minimum dt allowed between observations - defaults to 0
maxdt	Maximum dt allowed between observations - defaults to Inf

## Value

Returns a data.frame with 7 columns: x (unlagged time series data); laggedx (lagged time series data); xmid (average of time series and lagged time series values); dt (time lag between x and laggedx); time (time for observation x); laggedtime (time for observation laggedx); treatment (treatment for observation)

# Examples

get\_logistic Logistic Growth

#### Description

 $\label{eq:calculates} Calculates \ logistic \ growth \ for \ population \ based \ on \ formula \ Nt=K*(N0*exp(r*time))/(K+N0*(exp(r*time)-1))/(K+N0*(exp(r*tim$ 

# Usage

get\_logistic(time, N0, r, K)

#### Arguments

time	The time steps corresponding to each observation
NØ	Initial Population Size
r	Growth rate
К	Carying Capacity

# Value

population size N for each time steps as a vector

# Examples

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```
# fit linear model to get dNNdt ~ r + s*N
mod_Pc<-lm(dNNdt_Pc~Volume_Species1, Pcmono)</pre>
rsn_pars<-coef(mod_Pc)</pre>
# transform into logistic growth parameters
logistic_pars<-c(r=unname(rsn_pars["(Intercept)"]),</pre>
     K=unname(-rsn_pars["(Intercept)"]/rsn_pars["Volume_Species1"]))
#fit with nls, using linear model estimates as starting values for parameters
nls_mod<-nls(Volume_Species1~get_logistic(time = Day, N0, r, K),</pre>
             data=Pcmono,
             start=c(N0=unname(Pcmono$Volume_Species1[which.min(Pcmono$Day)]),
             r=unname(logistic_pars["r"]), K=unname(logistic_pars["K"])))
summary(nls_mod)
# plot results
plot(Volume_Species1~Day, Pcmono, type="b", ylab="P. caudatum Volume")
timesq<-seq(0, 30, length=100)</pre>
Ntest<-get_logistic(time = timesq, N0=coef(nls_mod)["N0"], r=coef(nls_mod)["r"],</pre>
       K=coef(nls_mod)["K"])
lines(timesq, Ntest, col="red")
```

dt=lagged\_data\_Pc\$dt)

huffaker\_1963 Huffaker Mite Data

## Description

A dataset containing the abundances mite species from some of the Huffaker experiments.

#### Usage

huffaker\_1963

# Format

A data frame with 168 rows and 6 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

Species Species name

Weeks Experiment week

**Individuals** Number of individuals

Treatment 60 vs. 24-week experiments

#### Source

Huffaker (1958) Experimental Studies on Predation: Dispersion Factors and Predator-Prey Oscillations. Hilgardia 27: 343-83.

lv\_interaction Lotka-Volterra Interactions

## Description

Calculates dn/dt for n species in a Lokta-Volterra system, following the form:  $dn/dt = ni * (ri + aii * ni + sum_j(aij * nj))$  Note that aii coefficients can be positive or negative, although positive coefficients risk having the system run to infinite population sizes, which will crash the function.

#### Usage

lv\_interaction(time, n, parms)

# Arguments

time	The time steps corresponding to each observation - exists to interface with ode function, but should be left blank.
n	A vector of species abundances
parms	A vector of parameters - the first n elements should be the growth rates r1, r2, rn for all n species. The remaining terms should be the elements of the interaction matrix A, listed in the order a11, a12, a1n, a21, a22, a2n, an1, an2, ann.

# Value

vector of growth rates for each species

#### Examples

```
# load data from competition experiment
data(gause_1934_science_f02_03)
# subset data to include just mixtures
```

```
mixturedata<-gause_1934_science_f02_03[gause_1934_science_f02_03$Treatment=="Mixture",]</pre>
```

```
# get time-lagged observations for each species
Pc_lagged<-get_lag(x = mixturedata$Volume_Species1, time = mixturedata$Day)
Pa_lagged<-get_lag(x = mixturedata$Volume_Species2, time = mixturedata$Day)</pre>
```

```
# calculate per-capita growth rates
Pc_dNNdt<-percap_growth(x = Pc_lagged$x, laggedx = Pc_lagged$laggedx, dt = Pc_lagged$dt)
Pa_dNNdt<-percap_growth(x = Pa_lagged$x, laggedx = Pa_lagged$laggedx, dt = Pa_lagged$dt)</pre>
```

# fit linear models to dNNdt, based on average

```
# abundances between current and lagged time steps
Pc_mod_dat<-data.frame(Pc_dNNdt=Pc_dNNdt, Pc=Pc_lagged$laggedx, Pa=Pa_lagged$laggedx)
mod_comp_Pc<-lm(Pc_dNNdt~Pc+Pa, data=Pc_mod_dat)</pre>
Pa_mod_dat<-data.frame(Pa_dNNdt=Pa_dNNdt, Pa=Pa_lagged$laggedx, Pc=Pc_lagged$laggedx)
mod_comp_Pa<-lm(Pa_dNNdt~Pa+Pc, data=Pa_mod_dat)</pre>
# model summaries
summary(mod_comp_Pc)
summary(mod_comp_Pa)
# extract parameters
# note - linear regressions give us dynamics in the form:
# dni/nidt ~ (Intercept) + (n1_slope) * n1 + (n2_slope) n2
# and thus:
# dni/dt = n1*((Intercept) + (n1_slope) * n1 + (n2_slope) n2)
# growth rates
r1 <- unname(coef(mod_comp_Pc)["(Intercept)"])</pre>
r2 <- unname(coef(mod_comp_Pa)["(Intercept)"])</pre>
# self-limitation
a11 <- unname(coef(mod_comp_Pc)["Pc"])</pre>
a22 <- unname(coef(mod_comp_Pa)["Pa"])</pre>
# effect of Pa on Pc
a12 <- unname(coef(mod_comp_Pc)["Pa"])</pre>
# effect of Pc on Pa
a21 <- unname(coef(mod_comp_Pa)["Pc"])</pre>
# run ODE:
# make paramter vector:
parms <- c(r1, r2, a11, a12, a21, a22)
initialN <- c(1, 1)
out <- deSolve::ode(y=initialN, times=1:25, func=lv_interaction, parms=parms)</pre>
matplot(out[,1], out[,-1], type="1",
   xlab="time", ylab="N", col=c("black","red"), lty=c(1,3), lwd=2, ylim=c(0, 150))
legend("topleft", c("Pc", "Pa"), col=c(1,2), lwd=2, lty=c(1,3))
# now, plot in points from data
points(mixturedata$Day, mixturedata$Volume_Species1, col=1)
points(mixturedata$Day, mixturedata$Volume_Species2, col=2)
```

lv\_interaction\_log Lotka-Volterra Interactions in Log Space

#### Description

Calculates dn/dt for n species in a Lokta-Volterra system, in log space, following the form:  $dlog(ni)/dt = (ri + aii * ni + sum_j(aij * nj))$  This form can be helpful for optimization routines where species abundances are close to zero.

# Usage

lv\_interaction\_log(time, n\_log, parms)

# Arguments

time	The time steps corresponding to each observation - exists to interface with ode function, but should be left blank.
n_log	A vector of species abundances, in log space
parms	A vector of parameters - the first n elements should be the growth rates r1, r2, rn for all n species. The remaining terms should be the elements of the interaction matrix A, listed in the order a11, a12, a1n, a21, a22, a2n, an1, an2, ann.

# Value

vector of growth rates for each species in log space

lv\_optim

Optimizer for Lotka-Volterra Interactions

# Description

Identifies optimal parameter values for a Lotka-Volterra interaction system.

# Usage

```
lv_optim(
   pars,
   opt_data,
   parm_signs,
   standardize = TRUE,
   odefun = lv_interaction_log
)
```

# Arguments

pars	A vector of parameter values in log space to be optimized. Must include a logged starting abundance for each species, followed by the logged absolute values of the growth rates, followed by the logged absolute value of the elements of the interaction matrix.
opt_data	Abundance data for optimization. Must include one column labeled 'time' with time steps, and a column for each species abundance.
parm_signs	A vector that provides the desired sign of each parameter (i.e1 or 1). If value is zero, then the term is held at zero (but should be left out of the pars vector).
standardize	A logical, defaulting to TRUE - should error be calculated based on standardized values of outputs? Allows for more equal weighting of observed variabels.
odefun	The function to use to simulate the ODE - defaults to lv_interaction_log

#### lv\_optim

#### Value

squared error between model fits for given parameter values, and observations

# Examples

```
# load data from competition experiment
data(gause_1934_book_f32)
# keep all data - no separate treatments exist for this experiment
predatorpreydata<-gause_1934_book_f32
# get time-lagged observations for each species
prey_lagged < get_lag(x = predatorpreydata$Individuals_Prey, time = predatorpreydata$Day)
predator_lagged<-get_lag(x = predatorpreydata$Individuals_Predator, time = predatorpreydata$Day)</pre>
# calculate per-capita growth rates
prey_dNNdt<-percap_growth(x = prey_lagged$x, laggedx = prey_lagged$laggedx, dt = prey_lagged$dt)</pre>
predator_dNNdt<-percap_growth(x = predator_lagged$x,</pre>
      laggedx = predator_lagged$laggedx, dt = predator_lagged$dt)
# fit linear models to dNNdt, based on average
# abundances between current and lagged time steps
prey_mod_dat<-data.frame(prey_dNNdt=prey_dNNdt, prey=prey_lagged$laggedx,</pre>
      predator=predator_lagged$laggedx)
mod_prey<-lm(prey_dNNdt~prey+predator, data=prey_mod_dat)</pre>
predator_mod_dat<-data.frame(predator_dNNdt=predator_dNNdt,</pre>
      predator=predator_lagged$laggedx, prey=prey_lagged$laggedx)
mod_predator<-lm(predator_dNNdt~predator+prey, data=predator_mod_dat)</pre>
# model summaries
summary(mod_prey)
summary(mod_predator)
# extract parameters
# growth rates
r1 <- unname(coef(mod_prey)["(Intercept)"])</pre>
r2 <- unname(coef(mod_predator)["(Intercept)"])</pre>
# self-limitation
a11 <- unname(coef(mod_prey)["prey"])</pre>
a22 <- unname(coef(mod_predator)["predator"])</pre>
# effect of Pa on Pc
a12 <- unname(coef(mod_prey)["predator"])</pre>
# effect of Pc on Pa
a21 <- unname(coef(mod_predator)["prey"])</pre>
# run ODE:
# make parameter vector:
parms <- c(r1, r2, a11, a12, a21, a22)
```

```
initialN <- c(4, 0.1)
out <- deSolve::ode(y=initialN, times=seq(1, 17, length=100), func=lv_interaction, parms=parms)
matplot(out[,1], out[,-1], type="l",
        xlab="time", ylab="N", col=c("black","red"), lty=c(1,3), lwd=2, ylim=c(0, 60))
legend("topright", c("Pc", "Dn"), col=c(1,2), lwd=2, lty=c(1,3))
# now, plot in points from data
points(predatorpreydata$Day, predatorpreydata$Individuals_Predator , col=2)
points(predatorpreydata$Day, predatorpreydata$Individuals_Prey, col=1)
# uh-oh - This is a bad fit. This suggests that our linear model
# approximation isn't very good. Instead, we should try optimizing
# directly using the ode solver
# Re-run using an optimizer
# Data for the optimizer:
# Must have a column with time steps labeled 'time', and
# columns for each species in the community.
opt_data<-data.frame(time=predatorpreydata$Day, Prey=predatorpreydata$Individuals_Prey,
    Predator=predatorpreydata$Individuals_Predator)
# Save the signs of the parameters -
# optimizer works in log space, so these
# must be specified separately
parm_signs<-sign(parms)</pre>
# parameter vector for optimizer -
# must be a vector with, first, the
# starting abundances in log space,
# and second, the parameter values,
# again in log space
pars<-c(log(initialN), log(abs(parms)))</pre>
# run optimizer
optout<-optim(par = pars, fn = lv_optim, hessian = TRUE,
             opt_data=opt_data, parm_signs=parm_signs)
# extract parameter vector:
parms <- exp(optout$par[-c(1:2)])*parm_signs</pre>
initialN <- exp(optout$par[1:2])</pre>
out <- deSolve::ode(y=initialN, times=seq(1, 17, length=100), func=lv_interaction, parms=parms)
matplot(out[,1], out[,-1], type="1",
        xlab="time", ylab="N", col=c("black","red"), lty=c(1,3), lwd=2, ylim=c(0, 60))
legend("topright", c("Pc", "Dn"), col=c(1,2), lwd=2, lty=c(1,3))
# now, plot in points from data
points(predatorpreydata$Day, predatorpreydata$Individuals_Predator , col=2)
points(predatorpreydata$Day, predatorpreydata$Individuals_Prey, col=1)
# get rough estimate of confidence intervals
fisher_info<-solve(-optout$hessian)</pre>
optout$par_sd<-sqrt(abs(diag(fisher_info)))</pre>
```

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rownames(parameter\_intervals)<-c("prey", "predator", "r1", "r2", "a11", "a12", "a21", "a22")
parameter\_intervals</pre>

mclaren\_1994\_f03 Wolf, Moose, and Fir dynamics from Isle Royale

# Description

A dataset containing the abundances of wolves, moose, and fir trees from the Isle Royale study of McLaren et al.

# Usage

mclaren\_1994\_f03

# Format

A data frame with 140 rows and 7 variables:

Paper Paper from which data are drawn

Figure Figure number in paper

year Year of measurements

Species Species name

width Width of tree rings

individuals Number of wolf or moose individuals

AET..mm. AET water availablility index

#### Source

McLaren & Peterson (1994) Wolves, Moose, and Tree Rings on Isle Royale. Science 266:1555-1558.

ode\_prediction

#### Description

Takes in paramter values in the form returned by the gause\_wrapper function, and calculates expected abundances for all n species, returned as a single vector. This function is potentially useful in combination with other optimizer software, e.g. as might be used for hypothesis testing.

#### Usage

ode\_prediction(pars\_full, time, N)

# Arguments

pars_full	Initial Population Size
time	A vector of times. Must be repeated once per species.
Ν	Number of species. Can be either a number, or a vector the same length as time.

# Value

a stacked vector with predicted abundances for all species

# Examples

#load competition data
data("gause\_1934\_science\_f02\_03")

```
#subset out data from species grown in mixture
mixturedat<-gause_1934_science_f02_03[gause_1934_science_f02_03$Treatment=="Mixture",]</pre>
```

```
#extract time and species data
time<-mixturedat$Day
species<-data.frame(mixturedat$Volume_Species1, mixturedat$Volume_Species2)
colnames(species)<-c("P_caudatum", "P_aurelia")</pre>
```

```
#run wrapper
gause_out<-gause_wrapper(time=time, species=species)</pre>
```

yest<-ode\_prediction(pars\_full, time=fittigdata\$time, N=fittigdata\$N)</pre>

#### percap\_growth

percap\_growth Per-capita growth rate

# Description

Calculates per-capita growth rate, using log ratios following the formula dN/Ndt = log(N(t)/N0)/dt.

# Usage

```
percap_growth(x, laggedx, dt)
```

# Arguments

х	Abundance
laggedx	Lagged abundance
dt	Time lag between observations

# Value

Per-capita growth rate

# Examples

test\_goodness\_of\_fit Test goodness of fit

#### Description

Tests goodness of fit for predictions vs. observations. This statistic can be though of in the same way as a classic "R2", except that it measures scatter around the 1-1 line, rather than around a fitted regresson line of observed vs. predicted values. Value close to 1 indicate a that predictions match observations closely. Values at or below zero indicate that predictions do not match observations any better than the grand mean taken across all observations.

## Usage

```
test_goodness_of_fit(observed, predicted, bycolumn = FALSE, droptimecol = TRUE)
```

#### Arguments

observed	A vector or matrix of observed values.
predicted	A vector or matrix of predicted values.
bycolumn	If TRUE, then separate values are calculated for each column in observed and predicted.
droptimecol	If TRUE, will automatically remove the column labeled "time" in the predicted variable. This is useful for dealing with the default output of the gause_wrapper function. Defaults to FALSE.

# Examples

#load competition data
data("gause\_1934\_science\_f02\_03")

#subset out data from species grown in mixture mixturedat<-gause\_1934\_science\_f02\_03[gause\_1934\_science\_f02\_03\$Treatment=="Mixture",]</pre>

```
#extract time and species data
time<-mixturedat$Day
species<-data.frame(mixturedat$Volume_Species1, mixturedat$Volume_Species2)
colnames(species)<-c("P_caudatum", "P_aurelia")</pre>
```

```
#run wrapper
#note - keeptimes=TRUE is needed, so that predicted time steps match
#observed time steps
gause_out<-gause_wrapper(time=time, species=species, keeptimes = TRUE)</pre>
```

```
#calculate goodness of fit
test_goodness_of_fit(observed=species, predicted=gause_out)
```

```
# > 0.9 for both time series - these are good fits!
```

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