

Package ‘growthrates’

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Type Package

Title Estimate Growth Rates from Experimental Data

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Description A collection of methods to determine growth rates from experimental data, in particular from batch experiments and plate reader trials.

Depends lattice, deSolve

Imports stats, graphics, methods, parallel, utils, FME

License GPL (>= 2)

URL <https://github.com/tpetzoldt/growthrates>

VignetteBuilder knitr

Suggests knitr

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Collate 'aaa_growthmodel-class.R' 'aab_growthmodel-constructor.R'
'aac_classes.R' 'aad_set_generics.R' 'all_easylinear.R'
'all_growthmodels.R' 'all_splines.R' 'bactgrowth.R' 'cost.R'
'extract.R' 'fit_easylinear.R' 'fit_growthmodel.R'
'fit_spline.R' 'grow_baranyi.R' 'grow_exponential.R'
'grow_genlogistic.R' 'grow_gompertz.R' 'grow_huang.R'
'grow_logistic.R' 'grow_richards.R' 'grow_twostep.R'
'growthrates-package.R' 'lm_window.R' 'methods.R'
'multisplit.R' 'names.R' 'parse_formula.R'
'parse_formula_nonlin.R' 'plot.R' 'utilities.R'

NeedsCompilation yes

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growthrates-package *Estimate Growth Rates from Experimental Data*

Description

A collection of methods to determine growth rates from experimental data, in particular from batch experiments and plate reader trials.

Details

Package: growthrates
 Type: Package
 Version: 0.6.5
 Date: 2016-03-12
 License: GPL (>= 2)
 LazyLoad: yes

The package contains basically three methods:

- fit a linear regression to a subset of data with the steepest log-linear increase (a method, similar to Hall et al., 2013),
- fit parametric nonlinear models to the complete data set, where the model functions can be given either in closed form or as numerically solved (system of) differential equation(s),
- use maximum of the 1st derivative of a smoothing spline with log-transformed y-values (similar to Kahm et al., 2010).

The package can fit data sets of single experiments or complete series containing multiple data sets. Included are functions for extracting estimates and for plotting. The package supports growth models given as numerically solved differential equations. Multi-core computation is used to speed up fitting of parametric models.

Author(s)

Thomas Petzoldt

References

Hall, B. G., Acar, H. and Barlow, M. 2013. Growth Rates Made Easy. *Mol. Biol. Evol.* 31: 232-238 doi:10.1093/molbev/mst197

Kahm, M., Hasenbrink, G., Lichtenberg-Frate, H., Ludwig, J., Kschischo, M. 2010. *grofit*: Fitting Biological Growth Curves with R. *Journal of Statistical Software*, 33(7), 1-21. URL <http://www.jstatsoft.org/v33/i07/>

Soetaert, K. and Petzoldt, T. 2010. Inverse Modelling, Sensitivity and Monte Carlo Analysis in R Using Package FME. *Journal of Statistical Software*, 33(3), 1-28. URL <http://www.jstatsoft.org/v33/i03/>

Soetaert, K., Petzoldt, T. Setzer, R. W. 2010. Solving Differential Equations in R: Package deSolve. *Journal of Statistical Software*, 33(9), 1-25. URL <http://www.jstatsoft.org/v33/i09/>

See Also

package [grofit](#) for a related package from Kahm et al. (2010) that implements a different approach.

Examples

```
data(bactgrowth)
splitted.data <- multisplit(bactgrowth, c("strain", "conc", "replicate"))

## get table from single experiment
dat <- splitted.data[["D:0:1"]]

fit1 <- fit_spline(dat$time, dat$value)
plot(fit1, log="y")
plot(fit1)
```

```

## derive start parameters from spline fit
p <- coef(fit1)

## subset of first 10 data
first10 <- dat[1:10, ]
fit2 <- fit_growthmodel(grow_exponential, p=p, time=first10$time, y=first10$value)

## use parameters from spline fit and take K from the data maximum
p <- c(coef(fit1), K = max(dat$value))
fit3 <- fit_growthmodel(grow_logistic, p=p, time=dat$time, y=dat$value, transform="log")

plot(fit1)
lines(fit2, col="green")
lines(fit3, col="red")

```

all_easylinear

Easy Growth Rates Fit to data Frame

Description

Determine maximum growth rates from log-linear part of the growth curve for a series of experiments.

Usage

```
all_easylinear(...)
```

```
## S3 method for class 'formula'
```

```
all_easylinear(formula, data, h = 5, quota = 0.95,
  subset = NULL, ...)
```

```
## S3 method for class 'data.frame'
```

```
all_easylinear(data, grouping, time = "time",
  y = "value", h = 5, quota = 0.95, ...)
```

Arguments

formula	model formula specifying dependent, independent and grouping variables in the form: dependent ~ independent group1 + group2 +
data	data frame of observational data.
h	width of the window (number of data).
quota	part of window fits considered for the overall linear fit (relative to max. growth rate).
subset	a specification of the rows to be used: defaults to all rows.
grouping	model formula or character vector of criteria defining subsets in the data frame.
time	character vectors with name independent variable.

y character vector with name of dependent variable
... reserved for future extensions.

Value

object with parameters of all fits.

References

Hall, B. G., H. Acar and M. Barlow 2013. Growth Rates Made Easy. Mol. Biol. Evol. 31: 232-238
doi:10.1093/molbev/mst197

See Also

Other fitting functions: [all_growthmodels](#), [all_splines](#), [fit_easylinear](#), [fit_growthmodel](#),
[fit_spline](#)

Examples

```
library("growthrates")
L <- all_easylinear(value ~ time | strain + conc + replicate, data=bactgrowth)
summary(L)
coef(L)
rsquared(L)

results <- results(L)

library(lattice)
xyplot(mumax ~ conc|strain, data=results)
```

all_growthmodels

Fit Nonlinear Growth Models to Data Frame

Description

Determine maximum growth rates by nonlinear fits for a series of experiments.

Usage

```
all_growthmodels(...)  
  
## S3 method for class 'formula'  
all_growthmodels(formula, data, p, lower = -Inf,  
  upper = Inf, which = names(p), FUN = NULL, method = "Marq",  
  transform = c("none", "log"), ..., subset = NULL,  
  ncores = detectCores(logical = FALSE))
```

```
## S3 method for class 'function'
all_growthmodels(FUN, p, data, grouping = NULL,
  time = "time", y = "value", lower = -Inf, upper = Inf,
  which = names(p), method = "Marq", transform = c("none", "log"), ...,
  ncores = detectCores(logical = FALSE))
```

Arguments

formula	model formula specifying dependent, independent and grouping variables in the form: dependent ~ independent group1 + group2 +
data	data frame of observational data.
p	named vector of start parameters and initial values of the growth model.
lower	lower bound of the parameter vector.
upper	upper bound of the parameter vector.
which	vector of parameter names that are to be fitted.
FUN	function of growth model to be fitted.
method	character vector specifying the optimization algorithm.
transform	fit model to non-transformed or log-transformed data.
subset	a specification of the rows to be used: defaults to all rows.
ncores	number of CPU cores used for parallel computation. The number of real cores is detected automatically by default, but for debugging purposes it could be wise to set ncores = 1. Usage of logical (hyperthreading) cores does not speed up computation.
grouping	vector of grouping variables defining subsets in the data frame.
time	character vector with name of independent variable.
y	character vector with name of dependent variable.
...	additional parameters passed to the optimizer.

Value

object containing the parameters of all fits.

See Also

Other fitting functions: [all_easylinear](#), [all_splines](#), [fit_easylinear](#), [fit_growthmodel](#), [fit_spline](#)

Examples

```
data(bactgrowth)
splitted.data <- multisplit(value ~ time | strain + conc + replicate,
  data = bactgrowth)

## show which experiments are in splitted.data
```

```
names(splitted.data)

## get table from single experiment
dat <- splitted.data[["D:0:1"]]

fit0 <- fit_spline(dat$time, dat$value)

fit1 <- all_splines(value ~ time | strain + conc + replicate,
                  data = bactgrowth, spar = 0.5)

## these examples require some CPU power and may take a bit longer

## initial parameters
p <- c(coef(fit0), K = max(dat$value))

## avoid negative parameters
lower = c(y0 = 0, mumax = 0, K = 0)

## fit all models
fit2 <- all_growthmodels(value ~ time | strain + conc + replicate,
                        data = bactgrowth, FUN=grow_logistic,
                        p = p, lower = lower, ncores = 2)

results1 <- results(fit1)
results2 <- results(fit2)
plot(results1$mumax, results2$mumax, xlab="smooth splines", ylab="logistic")

## experimental: nonlinear model as part of the formula

fit3 <- all_growthmodels(
  value ~ grow_logistic(time, parms) | strain + conc + replicate,
  data = bactgrowth, p = p, lower = lower, ncores = 2)

## this allows also to fit to the 'global' data set or any subsets
fit4 <- all_growthmodels(
  value ~ grow_logistic(time, parms),
  data = bactgrowth, p = p, lower = lower, ncores = 1)
plot(fit4)

fit5 <- all_growthmodels(
  value ~ grow_logistic(time, parms) | strain + conc,
  data = bactgrowth, p = p, lower = lower, ncores = 2)
plot(fit5)
```

Description

Determine maximum growth rates from log-linear part of the growth curve for a series of experiments by using smoothing splines.

Usage

```
all_splines(...)

## S3 method for class 'formula'
all_splines(formula, data = NULL, optgrid = 50,
             subset = NULL, ...)

## S3 method for class 'data.frame'
all_splines(data, grouping = NULL, time = "time",
             y = "value", optgrid = 50, ...)
```

Arguments

formula	model formula specifying dependent, independent and grouping variables in the form: dependent ~ independent group1 + group2 +
data	data frame of observational data.
optgrid	number of steps on the x-axis used for searching the maximum of the first derivative of the spline. The default should work in most cases, as long as the data are equally spaced. A smaller number may lead to non-detectable speed-up, but has the risk that the search is trapped in a local minimum.
subset	a specification of the rows to be used: defaults to all rows.
grouping	vector of grouping variables defining subsets in the data frame.
time	character vectors with name independent variable.
y	character vector with name of dependent variable.
...	other parameters passed to smooth.spline , see details.

Details

The method was inspired by an algorithm of Kahm et al. (2010), with different settings and assumptions. In the moment, spline fitting is always done with log-transformed data, assuming exponential growth at the time point of the maximum of its first derivative.

All the hard work is done by function [smooth.spline](#) from package **stats**, that is highly user configurable. Normally, smoothness is automatically determined via cross-validation. This works well in many cases, whereas manual adjustment is required otherwise, e.g. by setting `spar` to a fixed value $[0, 1]$ that also disables cross-validation. A typical case where cross validation does not work is, if time dependent measurements are taken as pseudoreplicates from the same experimental unit.

Value

object with parameters of the fit.

References

Kahm, M., Hasenbrink, G., Lichtenberg-Frate, H., Ludwig, J., Kschischo, M. 2010. grofit: Fitting Biological Growth Curves with R. Journal of Statistical Software, 33(7), 1-21. URL <http://www.jstatsoft.org/v33/i07/>

See Also

Other fitting functions: [all_easylinear](#), [all_growthmodels](#), [fit_easylinear](#), [fit_growthmodel](#), [fit_spline](#)

Examples

```
data(bactgrowth)
L <- all_splines(value ~ time | strain + conc + replicate,
                data = bactgrowth, spar = 0.5)

par(mfrow=c(4, 3))
plot(L)
results <- results(L)
xyplot(mumax ~ log(conc + 1)|strain, data=results)

## fit splines at lower grouping levels
L2 <- all_splines(value ~ time | conc + strain,
                 data = bactgrowth, spar = 0.5)
plot(L2)

## total data set without any grouping
L3 <- all_splines(value ~ time,
                 data = bactgrowth, spar = 0.5)
par(mfrow=c(1, 1))
plot(L3)
```

bactgrowth

Plate Reader Data of Bacterial Growth

Description

Example data set from growth experiments with different concentrations of antibiotics.

Format

Data frame with the following columns:

strain identifier of the bacterial strain, D=donor, R=recipient, T=transconjugant.

replicate replicate of the trial.

conc concentration of the antibiotics (Tetracycline).

time time in hours.

value bacteria concentration measured as optical density.

Details

This rather 'difficult' data set was intentionally selected to make model fitting by the package more challenging.

Source

Claudia Seiler, TU Dresden, Institute of Hydrobiology.

Examples

```
## plot data and determine growth rates
data(bactgrowth)

library(lattice)
xyplot(value ~ time | strain + as.factor(conc),
       data = bactgrowth, groups = replicate)
```

fit_easylinear

Fit Exponential Growth Model with a Heuristic Linear Method

Description

Determine maximum growth rates from the log-linear part of a growth curve using a heuristic approach similar to the “growth rates made easy”-method of Hall et al. (2013).

Usage

```
fit_easylinear(time, y, h = 5, quota = 0.95)
```

Arguments

time	vector of independent variable.
y	vector of dependent variable (concentration of organisms).
h	width of the window (number of data).
quota	part of window fits considered for the overall linear fit (relative to max. growth rate)

Details

The algorithm works as follows:

1. Fit linear regressions to all subsets of h consecutive data points. If for example $h = 5$, fit a linear regression to points 1 ... 5, 2 ... 6, 3 ... 7 and so on. The method seeks the highest rate of exponential growth, so the dependent variable is of course log-transformed.

2. Find the subset with the highest slope

$$b_{max}$$

and include also the data points of adjacent subsets that have a slope of at least

$$quota \cdot b_{max}$$

, e.g. all data sets that have at least 95% of the maximum slope.

3. Fit a new linear model to the extended data window identified in step 2.

Value

object with parameters of the fit.

References

Hall, B. G., H. Acar and M. Barlow 2013. Growth Rates Made Easy. *Mol. Biol. Evol.* 31: 232-238
doi:10.1093/molbev/mst197

See Also

Other fitting functions: [all_easylinear](#), [all_growthmodels](#), [all_splines](#), [fit_growthmodel](#), [fit_spline](#)

Examples

```
data(bactgrowth)

splitted.data <- multisplit(bactgrowth, c("strain", "conc", "replicate"))
dat <- splitted.data[[1]]

plot(value ~ time, data=dat)
fit <- fit_easylinear(dat$time, dat$value)

plot(fit)
plot(fit, log="y")
plot(fit, which="diagnostics")

fitx <- fit_easylinear(dat$time, dat$value, h=8, quota=0.95)

plot(fit, log="y")
lines(fitx, pch="+", col="blue")

plot(fit)
lines(fitx, pch="+", col="blue")
```

fit_growthmodel *Fit Nonlinear Parametric Growth Model*

Description

Determine maximum growth rates by fitting nonlinear models.

Usage

```
fit_growthmodel(FUN, p, time, y, lower = -Inf, upper = Inf,  
  which = names(p), method = "Marq", transform = c("none", "log"),  
  control = NULL, ...)
```

Arguments

FUN	function of growth model to be fitted.
p	named vector of start parameters and initial values of the growth model.
time	vector of independent variable.
y	vector of dependent variable (concentration of organisms).
lower	lower bound of the parameter vector (optional).
upper	upper bound of the parameter vector (optional).
which	vector of parameter names that are to be fitted.
method	character vector specifying the optimization algorithm.
transform	fit model to non-transformed or log-transformed data.
control	A list of control parameters for the optimizers. See Details.
...	additional parameters passed to the optimizer.

Details

This function calls `modFit` from package **FME**. Syntax of control parameters and available options may differ, depending on the optimizer used, except `control=list(trace=...)` that switches tracing on and off for all methods and is either `TRUE`, or `FALSE`, or an integer value like 0, 1, 2, 3, depending on the optimizer.

Value

object with parameters of the fit.

See Also

[modFit](#) about constrained fitting of models to data

Other fitting functions: [all_easylinear](#), [all_growthmodels](#), [all_splines](#), [fit_easylinear](#), [fit_spline](#)

Examples

```

data(bactgrowth)
splitted.data <- multisplit(bactgrowth, c("strain", "conc", "replicate"))

## get one element either by index or by name
dat <- splitted.data[[1]]
dat <- splitted.data[["D:0:1"]]

p <- c(y0 = 0.01, mumax = 0.2, K = 0.1)

## unconstrained fitting
fit1 <- fit_growthmodel(FUN = grow_logistic, p = p, dat$time, dat$value)
coef(fit1)
summary(fit1)

## optional box-constraints
lower <- c(y0 = 1e-6, mumax = 0, K = 0)
upper <- c(y0 = 0.05, mumax = 5, K = 0.5)
fit1 <- fit_growthmodel(
  FUN = grow_logistic, p = p, dat$time, dat$value,
  lower = lower, upper = upper)

plot(fit1, log="y")

```

fit_spline

*Fit Exponential Growth Model with Smoothing Spline***Description**

Determine maximum growth rates from the first derivative of a smoothing spline.

Usage

```
fit_spline(time, y, optgrid = length(time), ...)
```

Arguments

time	vector of independent variable.
y	vector of dependent variable (concentration of organisms).
optgrid	number of steps on the x-axis used for the optimum search . algorithm. The default should work in most cases, as long as the data are equally spaced. A smaller number may lead to non-detectable speed-up, but has the risk that the search gets trapped in a local minimum.
...	other parameters passed to smooth.spline , see details.

Details

The method was inspired by an algorithm of Kahm et al. (2010), with different settings and assumptions. In the moment, spline fitting is always done with log-transformed data, assuming exponential growth at the time point of the maximum of the first derivative of the spline fit.

All the hard work is done by function `smooth.spline` from package `stats`, that is highly user configurable. Normally, smoothness is automatically determined via cross-validation. This works well in many cases, whereas manual adjustment is required otherwise, e.g. by setting `spar` to a fixed value `[0, 1]` that also disables cross-validation.

Value

object with parameters of the fit

References

Kahm, M., Hasenbrink, G., Lichtenberg-Frate, H., Ludwig, J., Kschischo, M. 2010. `grofit`: Fitting Biological Growth Curves with R. *Journal of Statistical Software*, 33(7), 1-21. URL <http://www.jstatsoft.org/v33/i07/>

See Also

Other fitting functions: [all_easylinear](#), [all_growthmodels](#), [all_splines](#), [fit_easylinear](#), [fit_growthmodel](#)

Examples

```
data(bactgrowth)
splitted.data <- multisplit(bactgrowth, c("strain", "conc", "replicate"))

dat <- splitted.data[[2]]
time <- dat$time
y <- dat$value

## automatic smoothing with cv
res <- fit_spline(time, y)

plot(res, log="y")
plot(res)
coef(res)

## a more difficult data set
dat <- splitted.data[[56]]
time <- dat$time
y <- dat$value

## default parameters
res <- fit_spline(time, y)
plot(res, log="y")

## small optgrid, trapped in local minimum
```

```

res <- fit_spline(time, y, optgrid=5)
plot(res, log="y")

## manually selected smoothing parameter
res <- fit_spline(time, y, spar=.5)
plot(res, log="y")
plot(res, ylim=c(0.005, 0.03))

```

function_growthmodel-class

Union Class of Growth Model or Function

Description

This class union comprises parametric model functions from class `growthmodel` and ordinary functions to describe time-dependent growth of organisms.

See Also

the constructor function [growthmodel](#) how to create instances of class `growthmodel`.

growthmodel

Create a User-defined Parametric Growth Model

Description

This constructor method allows to create user-defined functions that can be used as parametric models describing time-dependent growth of organisms.

Usage

```
growthmodel(x, pnames = NULL)
```

Arguments

x	a function with arguments <code>times</code> and <code>parms</code> , and returning a matrix with three columns <code>time</code> , <code>y</code> and <code>log_y</code> .
pnames	character vector with the names of the model parameters.

Details

Package **growthrates** has a plug-in architecture allowing user-defined growth models of the following form:

```

identifier <- function(time, parms) {
  ... content of function here ...
  return(as.matrix(data.frame(time=time, y=y, log_y=log(y))))
}

```

where `time` is a numeric vector and `parms` a named, non-nested list of model parameters. The constructor function `growthmodel` is used to attach the names of the parameters as an optional attribute.

See Also

Other growth models: [grow_baranyi](#), [grow_exponential](#), [grow_gompertz](#), [grow_huang](#), [grow_logistic](#), [grow_richards](#), [ode_genlogistic](#), [ode_twostep](#)

Examples

```

test <- function(time, parms) {
  with(as.list(parms), {
    y <- (K * y0) / (y0 + (K - y0) * exp(-mumax * time)) + y_shift
    return(as.matrix(data.frame(time=time, y=y, log_y=log(y))))
  })
}

mygrowthmodel <- growthmodel(test, c("y0", "mumax", "K", "y_shift"))

```

growthmodel-class *Class of Growth Model Functions*

Description

This class is used for the parametric `grow_...` functions of the package and can also be used for user-defined functions to describe time-dependent growth of organisms.

See Also

the constructor function [growthmodel](#) how to create instances of this class.

growthrates_fit-class *S4 Classes of Package* **growthrates**

Description

growthrates_fit: top-level class representing a growthrates fit with any method.
 nonlinear_fit: single nonlinear growthrates fit with package FME.
 easylinear_fit: single fit from the “growthrates made easy”-method.
 spline_fit: single fit with (optionally cross-validated) smoothing splines.
 multiple_fits: top-level class representing multiple fits with any method.
 multiple_easylinear_fits: class representing multiple fits with the “growthrates made easy”-method.
 multiple_nonlinear_fits: class representing multiple nonlinear fits.
 multiple_smooth.spline_fits: class representing multiple smooth.spline fits.

Slots

FUN model function used.
 fit results of the model fit.
 obs observation data used for model fitting.
 rsquared coefficient of determination.
 par parameters of the fit.
 ndx index values of the time points used (for easylinear_fit).
 xy x and y values at the maximum of the spline.

grow_baranyi *The Baranyi and Roberts Growth Model*

Description

The growth model of Baranyi and Roberts (1995) written as analytical solution of the system of differential equations.

Usage

```
grow_baranyi(time, parms)
```

Arguments

time	vector of time steps (independent variable).
parms	named parameter vector of the Baranyi growth model with: <ul style="list-style-type: none"> • y_0 initial value of abundance, • μ_{max} maximum growth rate (1/time), • K carrying capacity (max. abundance), • h_0 parameter specifying the initial physiological state of organisms (e.g. cells) and in consequence the lag phase ($h_0 = \text{max growth rate} * \text{lag phase}$).

Details

The version of the equation used in this package has the following form:

$$A = \text{time} + 1/\mu_{max} * \log(\exp(-\mu_{max} * \text{time}) + \exp(-h_0) - \exp(-\mu_{max} * \text{time} - h_0))$$

$$\log(y) = \log(y_0) + \mu_{max} * A - \log(1 + (\exp(\mu_{max} * A) - 1)/\exp(\log(K) - \log(y_0)))$$

Value

vector of dependent variable (y) and its log-transformed values (\log_y).

References

Baranyi, J. and Roberts, T. A. (1994). A dynamic approach to predicting bacterial growth in food. *International Journal of Food Microbiology*, 23, 277-294.

Baranyi, J. and Roberts, T.A. (1995). Mathematics of predictive microbiology. *International Journal of Food Microbiology*, 26, 199-218.

See Also

Other growth models: [grow_exponential](#), [grow_gompertz](#), [grow_huang](#), [grow_logistic](#), [grow_richards](#), [growthmodel](#), [ode_genlogistic](#), [ode_twostep](#)

Examples

```
time <- seq(0, 30, length=200)
y <- grow_baranyi(time, c(y0=0.01, mumax=.5, K=0.1, h0=5))[, "y"]
plot(time, y, type="l")
plot(time, y, type="l", log="y")
```

grow_exponential *Exponential Growth Model*

Description

Unlimited exponential growth model.

Usage

```
grow_exponential(time, parms)
```

Arguments

time vector of time steps (independent variable).
parms named parameter vector of the exponential growth model with:

- y_0 initial abundance (e.g. concentration of bacterial cells).
- m_{\max} maximum growth rate (1/time).

Details

The equation used is:

$$y = y_0 * \exp(m_{\max} * time)$$

Value

vector of dependent variable (y) and its log-transformed values (\log_y).

See Also

Other growth models: [grow_baranyi](#), [grow_gompertz](#), [grow_huang](#), [grow_logistic](#), [grow_richards](#), [growthmodel](#), [ode_genlogistic](#), [ode_twostep](#)

Examples

```
time <- seq(0, 30, length=200)
y <- grow_exponential(time, c(y0=1, mumax=0.5))["y"]
plot(time, y, type="l")
```

 grow_gompertz

Growth Model According to Gompertz

Description

Gompertz growth model written as analytical solution of the differential equation system.

Usage

```
grow_gompertz(time, parms)
```

Arguments

time	vector of time steps (independent variable).
parms	named parameter vector of the Gompertz growth model with: <ul style="list-style-type: none"> • y_0 initial value of abundance, • m_{max} maximum growth rate (1/time), • K maximum abundance (carrying capacity).

Details

The equation used here is:

$$y = K * \exp(\log(y_0/K) * \exp(-m_{max} * time))$$

Value

vector of dependent variable (y)

References

Tsoularis, A. (2001) Analysis of Logistic Growth Models. Res. Lett. Inf. Math. Sci, (2001) 2, 23-46.

See Also

Other growth models: [grow_baranyi](#), [grow_exponential](#), [grow_huang](#), [grow_logistic](#), [grow_richards](#), [growthmodel](#), [ode_genlogistic](#), [ode_twostep](#)

Examples

```
time <- seq(0, 30, length=200)
y <- grow_gompertz(time, c(y0=1, mumax=.2, K=10))[, "y"]
plot(time, y, type="l", ylim=c(0, 20))
```

grow_huang

Growth Model According to Huang

Description

Huang's growth model written as analytical solution of the differential equations.

Usage

```
grow_huang(time, parms)
```

Arguments

time	vector of time steps (independent variable).
parms	named parameter vector of Huang's growth model with: <ul style="list-style-type: none"> • y_0 initial value of abundance, • m_{max} maximum growth rate (1/time), • K carrying capacity (max. total concentration of cells), • α shape parameter determining the curvature, • λ parameter determining the lag time.

Details

The version of the equation used in this package has the following form:

$$B = time + 1/\alpha * \log((1 + \exp(-\alpha * (time - \lambda)))/(1 + \exp(\alpha * \lambda)))$$

$$\log(y) = \log(y_0) + \log(K) - \log(y_0 + (K - y_0) * \exp(-m_{max} * B))$$

In contrast to the original publication, all parameters related to population abundance (y , y_0 , K) are given as untransformed values. They are not log-transformed.

In general, using log-transformed parameters would indeed be a good idea to avoid the need of constrained optimization, but tests showed that box-constrained optimization worked reasonably well. Therefore, handling of optionally log-transformed parameters was removed from the package to avoid confusion. If you want to discuss this, please let me know.

Value

vector of dependent variable (y) and its log-transformed values (\log_y).

References

- Huang, Lihan (2008) Growth kinetics of *Listeria monocytogenes* in broth and beef frankfurters - determination of lag phase duration and exponential growth rate under isothermal conditions. *Journal of Food Science* 73(5), E235 – E242. doi:10.1111/j.1750-3841.2008.00785.x
- Huang, Lihan (2011) A new mechanistic growth model for simultaneous determination of lag phase duration and exponential growth rate and a new Belehradek-type model for evaluating the effect of temperature on growth rate. *Food Microbiology* 28, 770 – 776. doi:10.1016/j.fm.2010.05.019
- Huang, Lihan (2013) Introduction to USDA Integrated Pathogen Modeling Program (IPMP). Residue Chemistry and Predictive Microbiology Research Unit. USDA Agricultural Research Service.

See Also

Other growth models: [grow_baranyi](#), [grow_exponential](#), [grow_gompertz](#), [grow_logistic](#), [grow_richards](#), [growthmodel](#), [ode_genlogistic](#), [ode_twostep](#)

Examples

```
time <- seq(0, 30, length=200)
y <- grow_huang(time, c(y0=0.01, mumax=.1, K=0.1, alpha=1.5, lambda=3))[, "y"]
plot(time, y, type="l")
plot(time, y, type="l", log="y")
```

grow_logistic

Logistic Growth Model

Description

Classical logistic growth model written as analytical solution of the differential equation.

Usage

```
grow_logistic(time, parms)
```

Arguments

- | | |
|-------|---|
| time | vector of time steps (independent variable) |
| parms | named parameter vector of the logistic growth model with: <ul style="list-style-type: none"> • y0 initial value of population measure • mumax intrinsic growth rate (1/time) • K carrying capacity (max. total concentration of cells) |

Details

The equation used is:

$$y = (K * y0) / (y0 + (K - y0) * \exp(-mumax * time))$$

Value

vector of dependent variable (y) and its log-transformed values (log_y).

See Also

Other growth models: [grow_baranyi](#), [grow_exponential](#), [grow_gompertz](#), [grow_huang](#), [grow_richards](#), [growthmodel](#), [ode_genlogistic](#), [ode_twostep](#)

Examples

```
time <- seq(0, 30, length=200)
y <- grow_logistic(time, c(y0=1, mumax=0.5, K=10))[, "y"]
plot(time, y, type="l")
```

grow_richards

Growth Model According to Richards

Description

Richards growth model written as analytical solution of the differential equation.

Usage

```
grow_richards(time, parms)
```

Arguments

time	vector of time steps (independent variable).
parms	named parameter vector of the Richards growth model with: <ul style="list-style-type: none"> • y0 initial value of abundance, • mumax maximum growth rate (note different interpretation compared to exponential growth), • K carrying capacity (max. total concentration of cells), • beta shape parameter determining the curvature.

Details

The equation used is:

$$y = K * (1 - \exp(-beta * mumax * time) * (1 - (y0/K)^{-beta}))^{1/beta}$$

The naming of parameters used here follows the convention of Tsoularis (2001), but uses mumax for growtrate and y for abundance to make them consistent to other growth functions.

Value

vector of dependent variable (y) and its log-transformed values (log_y).

References

Richards, F. J. (1959) A Flexible Growth Function for Empirical Use. *Journal of Experimental Botany* 10 (2): 290–300.

Tsoularis, A. (2001) Analysis of Logistic Growth Models. *Res. Lett. Inf. Math. Sci.* (2001) 2, 23–46.

See Also

Other growth models: [grow_baranyi](#), [grow_exponential](#), [grow_gompertz](#), [grow_huang](#), [grow_logistic](#), [growthmodel](#), [ode_genlogistic](#), [ode_twostep](#)

Examples

```
time <- seq(0, 30, length=200)
y <- grow_richards(time, c(y0=1, mumax=.5, K=10, beta=2))[, "y"]
plot(time, y, type="l")
y <- grow_richards(time, c(y0=1, mumax=.5, K=10, beta=100))[, "y"]
lines(time, y, col="red")
y <- grow_richards(time, c(y0=1, mumax=.5, K=10, beta=.2))[, "y"]
lines(time, y, col="blue")
```

names.growthmodel *Get Names Attributes of Growth Models*

Description

Methods to get the parameter names of a growth model or to get or set identifiers of [multiple_fits](#) objects.

Usage

```
## S3 method for class 'growthmodel'
names(x)

## S4 method for signature 'multiple_fits'
names(x)

## S4 replacement method for signature 'multiple_fits'
names(x) <- value
```

Arguments

- x either a function being a parametric growth model of package **growthmodels** or an object with multiple fits.
- value a character vector of up to the same length as x, or NULL

Value

character vector of the parameter names

Methods

Method for class `growthmodel`: returns information about valid parameter names if a `pnames` attribute exists, else NULL. NULL.

Method for class `multiple_fits`: can be applied to objects returned by `all_growthmodels`, `all_splines` or `all_easylinear` respectively. This can be useful for selecting subsets, e.g. for plotting, see example below.

See Also

[multiple_fits](#), [all_growthmodels](#), [all_splines](#), [all_easylinear](#)

Examples

```
## growthmodel-method
names(grow_baranyi)

## multiple_fits-method
L <- all_splines(value ~ time | strain + conc + replicate,
  data = bactgrowth)

names(L)

## plot only the 'R' strain
par(mfrow=c(4, 6))
plot(L[grep("R:", names(L))])
```

ode_genlogistic

Generalized Logistic Growth Model

Description

Generalized logistic growth model solved as differential equation.

Usage

```
ode_genlogistic(time, y, parms, ...)
```

```
grow_genlogistic(time, parms, ...)
```

Arguments

time	vector of simulation time steps
y	named vector with initial value of the system (e.g. cell concentration)
parms	parameters of the generalized logistic growth model <ul style="list-style-type: none"> • mumax maximum growth rate (1/time) • K carrying capacity (max. abundance) • alpha, beta, gamma parameters determining the shape of growth. Setting all values to one returns the ordinary logistic function.
...	additional parameters passed to the ode-function.

Details

The model is given as its first derivative:

$$dy/dt = mumax * y^{alpha} * (1 - (y/K)^{beta})^{gamma}$$

that is then numerically integrated ('simulated') according to time (t).

The generalized logistic according to Tsoularis (2001) is a flexible model that covers exponential and logistic growth, Richards, Gompertz, von Bertalanffy, and some more as special cases.

The differential equation is solved numerically, where function `ode_genlogistic` is the differential equation, and `grow_genlogistic` runs a numerical simulation over time.

The default version `grow_genlogistic` is run directly as compiled code, whereas the R versions `ode_logistic` is provided for testing by the user.

Value

For `ode_genlogistic`: matrix containing the simulation outputs. The return value of has also class `deSolve`.

For `grow_genlogistic`: vector of dependent variable (y) and its log-transformed values (`log_y`).

- time time of the simulation
- y abundance of organisms
- log_y natural log of abundance

References

Tsoularis, A. (2001) Analysis of Logistic Growth Models. Res. Lett. Inf. Math. Sci, (2001) 2, 23-46.

See Also

Other growth models: [grow_baranyi](#), [grow_exponential](#), [grow_gompertz](#), [grow_huang](#), [grow_logistic](#), [grow_richards](#), [growthmodel](#), [ode_twostep](#)

Examples

```
time <- seq(0, 30, length=200)
parms <- c(mumax=0.5, K=10, alpha=1, beta=1, gamma=1)
y0 <- c(y=.1)
out <- ode(y0, time, ode_genlogistic, parms)
plot(out)

out2 <- ode(y0, time, ode_genlogistic, parms = c(mumax=0.2, K=10, alpha=2, beta=1, gamma=1))
out3 <- ode(y0, time, ode_genlogistic, parms = c(mumax=0.2, K=10, alpha=1, beta=2, gamma=1))
out4 <- ode(y0, time, ode_genlogistic, parms = c(mumax=0.2, K=10, alpha=1, beta=1, gamma=2))
out5 <- ode(y0, time, ode_genlogistic, parms = c(mumax=0.2, K=10, alpha=.5, beta=1, gamma=1))
out6 <- ode(y0, time, ode_genlogistic, parms = c(mumax=0.2, K=10, alpha=1, beta=.5, gamma=1))
out7 <- ode(y0, time, ode_genlogistic, parms = c(mumax=0.3, K=10, alpha=1, beta=1, gamma=.5))
plot(out, out2, out3, out4, out5, out6, out7)

## growth with lag (cf. log_y)
plot(ode(y0, time, ode_genlogistic, parms = c(mumax=1, K=10, alpha=2, beta=.8, gamma=5)))
```

ode_twostep

*Twostep Growth Model***Description**

System of two differential equations describing bacterial growth as two-step process of activation (or adaptation) and growth.

Usage

```
ode_twostep(time, y, parms, ...)
```

```
grow_twostep(time, parms, ...)
```

Arguments

time	actual time (for the ode) resp. vector of simulation time steps.
y	named vector with state of the system (y _i , y _a : abundance of inactive and active organisms, e.g. concentration of inactive resp. active cells).
parms	parameters of the two-step growth model: <ul style="list-style-type: none"> • y_i, y_a initial abundance of active and inactive organisms,

- kw activation (“wakeup”) constant (1/time),
 - mumax maximum growth rate (1/time),
 - K carrying capacity (max. abundance).
- ... placeholder for additional parameters (for user-extended versions of this function)

Details

The model is given as a system of two differential equations:

$$dy_i/dt = -kw * y_i$$

$$dy_a/dt = kw * y_i + mumax * (1 - (y_i + y_a)/K) * y_a$$

that are then numerically integrated (‘simulated’) according to time (t). The model assumes that the population consists of active (y_a) and inactive (y_i) cells so that the observed abundance is ($y = y_i + y_a$). Adapting inactive cells change to the active state with a first order ‘wakeup’ rate (kw).

Function `ode_twostep` is the system of differential equations, whereas `grow_twostep` runs a numerical simulation over time.

A similar two-compartment model, but without the logistic term, was discussed by Baranyi (1998).

Value

For `ode_twostep`: matrix containing the simulation outputs. The return value of has also class `deSolve`.

For `grow_twostep`: vector of dependent variable (y) and its log-transformed values (`log_y`):

- time time of the simulation
- yi concentration of inactive cells
- ya concentration of active cells
- y total cell concentration
- log_y natural log of total cell concentration

References

Baranyi, J. (1998). Comparison of stochastic and deterministic concepts of bacterial lag. *J. heor. Biol.* 192, 403–408.

See Also

Other growth models: [grow_baranyi](#), [grow_exponential](#), [grow_gompertz](#), [grow_huang](#), [grow_logistic](#), [grow_richards](#), [growthmodel](#), [ode_genlogistic](#)

Examples

```

time <- seq(0, 30, length=200)
parms <- c(kw = 0.1, mumax=0.2, K=0.1)
y0 <- c(yi=0.01, ya=0.0)
out <- ode(y0, time, ode_twostep, parms)

plot(out)

o <- grow_twostep(0:100, c(yi=0.01, ya=0.0, kw = 0.1, mumax=0.2, K=0.1))
plot(o)

```

plot

Plot Model Fits

Description

Methods to plot growth model fits together with the data and, alternatively, plot diagnostics

Usage

```

## S4 method for signature 'nonlinear_fit,missing'
plot(x, y, log = "", which = c("fit",
  "diagnostics"), ...)

## S4 method for signature 'nonlinear_fit'
lines(x, ...)

## S4 method for signature 'easylinear_fit,missing'
plot(x, y, log = "", which = c("fit",
  "diagnostics"), ...)

## S4 method for signature 'smooth.spline_fit,missing'
plot(x, y, ...)

## S4 method for signature 'easylinear_fit'
lines(x, ...)

## S4 method for signature 'multiple_fits,missing'
plot(x, y, ...)

```

Arguments

x an object returned by a model fitting function of package **growthrates**, that can contain one or multiple fits.

y (ignored) for compatibility with the default plot method.

log a character string which contains "y" if the y axis is to be logarithmic.
 which either "fit" (default) or "diagnostics".
 ... other arguments passed to the plotting methods, see [plot.default](#) and [par](#).

Details

The plot methods detect automatically which type of plot is appropriate, depending on the class of x and can plot either one single model fit or a complete series (multiple fits). In the latter case it may be wise to redirect the graphics to an external file (e.g. a pdf) and / or to use something like `par(mfrow=c(3,3))`.

The lines-method is currently only available for single fits.

If you need more control, you can of course also write own plotting functions.

See Also

[plot.default](#), [par](#), [fit_growthmodel](#), [fit_easylinear](#), [all_growthmodels](#), [all_easylinear](#)

Examples

```
data(bactgrowth)
splitted.data <- multisplit(bactgrowth, c("strain", "conc", "replicate"))

## get table from single experiment
dat <- splitted.data[["D:0:1"]]

fit1 <- fit_spline(dat$time, dat$value)
plot(fit1, log="y")
plot(fit1)

## derive start parameters from spline fit
p <- coef(fit1)

## subset of first 10 data
first10 <- dat[1:10, ]
fit2 <- fit_growthmodel(grow_exponential, p=p, time=first10$time, y=first10$value)

p <- c(coef(fit1), K = max(dat$value))
fit3 <- fit_growthmodel(grow_logistic, p=p, time=dat$time, y=dat$value, transform="log")

plot(fit1)
lines(fit2, col="green")
lines(fit3, col="red")

all.fits <- allSplines(value ~ time | strain + conc + replicate, data = bactgrowth)
par(mfrow=c(3,3))
plot(all.fits)

## it is also possible to plot a single fit or a subset of the fits
par(mfrow=c(1,1))
```

```

plot(all.fits[["D:0:1"]])
par(mfrow=c(2,2))
plot(all.fits[1:4])

## plot only the 'R' strain
par(mfrow=c(4, 6))
plot(all.fits[grep("R:", names(all.fits))])

```

rsquared, growthrates_fit-method

*Accessor Methods of Package **growthrates**.*

Description

Functions to access the results of fitted growthrate objects: summary, coeff, rsquared, deviance, residuals, df.residual, obs, results.

Usage

```

## S4 method for signature 'growthrates_fit'
rsquared(object, ...)

## S4 method for signature 'growthrates_fit'
obs(object, ...)

## S4 method for signature 'growthrates_fit'
coef(object, ...)

## S4 method for signature 'easylinear_fit'
coef(object, ...)

## S4 method for signature 'smooth.spline_fit'
coef(object, ...)

## S4 method for signature 'growthrates_fit'
deviance(object, ...)

## S4 method for signature 'growthrates_fit'
summary(object, ...)

## S4 method for signature 'nonlinear_fit'
summary(object, cov = TRUE, ...)

## S4 method for signature 'growthrates_fit'
residuals(object, ...)

```

```

## S4 method for signature 'growthrates_fit'
df.residual(object, ...)

## S4 method for signature 'smooth.spline_fit'
summary(object, cov = TRUE, ...)

## S4 method for signature 'smooth.spline_fit'
df.residual(object, ...)

## S4 method for signature 'smooth.spline_fit'
deviance(object, ...)

## S4 method for signature 'multiple_fits'
coef(object, ...)

## S4 method for signature 'multiple_fits'
rsquared(object, ...)

## S4 method for signature 'multiple_fits'
deviance(object, ...)

## S4 method for signature 'multiple_fits'
results(object, ...)

## S4 method for signature 'multiple_easylinear_fits'
results(object, ...)

## S4 method for signature 'multiple_fits'
summary(object, ...)

## S4 method for signature 'multiple_fits'
residuals(object, ...)

```

Arguments

object	name of a 'growthrate' object.
cov	boolean if the covariance matrix should be printed.
...	other arguments passed to the methods.

Examples

```

data(bactgrowth)
splitted.data <- multisplit(bactgrowth, c("strain", "conc", "replicate"))

## get table from single experiment
dat <- splitted.data[[10]]

fit1 <- fit_spline(dat$time, dat$value, spar=0.5)

```

```

coef(fit1)
summary(fit1)

## derive start parameters from spline fit
p <- c(coef(fit1), K = max(dat$value))
fit2 <- fit_growthmodel(grow_logistic, p=p, time=dat$time, y=dat$value, transform="log")
coef(fit2)
rsquared(fit2)
deviance(fit2)

summary(fit2)

plot(residuals(fit2) ~ obs(fit2)[,2])

```

[,multiple_fits,ANY,missing-method

Extract or Replace Parts of a 'multiple_fits' Object

Description

Operators to access parts of 'multiple_fits' objects

Usage

```
## S4 method for signature 'multiple_fits,ANY,missing'
x[i, j, ..., drop = TRUE]
```

```
## S4 method for signature 'multiple_fits,ANY,missing'
x[[i, j, ...]]
```

Arguments

x	object of class multiple_fits
i	numeric or character index
j	NULL (for compatibility with other uses of [or [[)
drop	If TRUE the result is coerced to the lowest possible dimension
...	optional arguments passed to [

Examples

```

data(bactgrowth)
L <- all_splines(value ~ time | strain + conc + replicate, data=bactgrowth)

coef(L[[1]])

```

```
plot(L[["R:0:2"]])  
  
par(mfrow=c(2, 2))  
plot(L[1:4])
```

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