# Package 'chemdeg'

June 25, 2024

experimenter in modeling chemical degradation kinetic data. The selection of the appropriate degradation model and parameter estimation is carried out automatically as far as possible and is driven by a rigorous statistical interpretation of the results. The package integrates already available goodness-of-fit statistics for nonlinear models. In addition it allows data fitting with the nonlinear first-order multi-target (FOMT) model. License GPL (>= 3)URL https://github.com/migliomatte/chemdeg, https://migliomatte.github.io/chemdeg/ BugReports https://github.com/migliomatte/chemdeg/issues **Depends** R (>= 2.10) Imports graphics, MASS, methods, stats Suggests knitr, rmarkdown, testthat (>= 3.0.0), tibble VignetteBuilder knitr Config/testthat/edition 3 **Encoding** UTF-8 LazyData true RoxygenNote 7.3.1 NeedsCompilation no Author Matteo Migliorini [aut, cre, cph],

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**Date/Publication** 2024-06-25 08:30:02 UTC

**Repository** CRAN

**Description** A collection of functions that have been developed to assist

Title Analysis of Chemical Degradation Kinetic Data

Type Package

Version 0.1.4

2 AICC

# **Contents**

ICC	2
niquad_red	3
et_order	4
OMT	
omtdata	7
OMTm	7
gen	8
oodness_of_fit	9
in_regr	10
rd1	11
ar_est_FOMT	
hase_space	
lot_ord	13
esults	
rfa	14
	16
	1

AICC

Index

Akaike Information Criterion With Correction

## **Description**

The function calculates the Akaike Information Criterion with correction for small samples size.

## Usage

AICC(fit)

# Arguments

fit

a 'nls'-object

## **Details**

When the sample size is small, there is a substantial probability that AIC (see stats::AIC() for more details) will select models that have too many parameters, i.e. that AIC will overfit. AICc is AIC with a correction for small sample sizes.

The AICc is computed as follows:

$$AICc = AIC + \frac{2k(k+1)}{n-k-1}$$

where n denotes the sample size and k denotes the number of parameters. Thus, AICc is essentially AIC with an extra penalty term for the number of parameters. Note that as  $n \to \infty$ , the extra penalty term converges to 0, and thus AICc converges to AIC.

chiquad\_red 3

## Value

Returns the AICc value

#### See Also

stats::AIC() for uncorrected AIC, stats::BIC(), stats::sigma() ,chiquad\_red() for other
goodness of fit indicators. goodness\_of\_fit()

## **Examples**

```
t <- seq(0, 10, 1)
y <- 1 / (0.5 * exp(t) + 1) + stats::rnorm(length(t), 0, 0.05)
fit <- nls(y ~ 1 / (k * exp(t) + 1),
  data = list(t = t, y = y),
  start = list(k = 0.2)
)
AICC(fit)</pre>
```

chiquad\_red

Reduced chi-squared

## **Description**

Function that returns the reduced chi-squared ( $\chi^2_{red}=\chi^2/df$ , where df are the degrees of freedom) value for a non-linear regression model (nls object). Reduced-chi squared is a goodness-of-fit measure. Values close to 1 indicates a good fit, while values >> 1 indicate poor fit and values < 1 indicate over-fitting. The function is calculated only with non-linear regression weighted on experimental error.

# Usage

```
chiquad_red(fit)
```

## **Arguments**

fit

nls object with weighted fit

## Value

Returns the reduced chi-squared value

## References

Philip R. Bevington, D. Keith Robinson, J. Morris Blair, A. John Mallinckrodt, Susan McKay (1993). *Data Reduction and Error Analysis for the Physical Sciences* 

det\_order

## See Also

```
stats::dchisq() for chi-squared distribution; stats::AIC(), stats::BIC(), stats::sigma()
(for RMSE), AICC() for other goodness-of-fit indicators. goodness_of_fit()
```

## **Examples**

```
x <- c(1, 2, 3, 4, 5)
y <- c(1.2, 3.9, 8.6, 17.4, 26)
er <- c(0.5, 0.8, 0.5, 1.9, 1.2)
fit1 <- nls(y ~ k * x^2,
    data = list(x = x, y = y),
    start = list(k = 1),
    weights = 1 / er^2
)
chiquad_red(fit1)

fit2 <- nls(y ~ k * x^3,
    data = list(x = x, y = y),
    start = list(k = 1),
    weights = 1 / er^2
)
chiquad_red(fit2)</pre>
```

det\_order

Determining reaction order and kinetic formula

## **Description**

The functions seeks to determine the reaction order and kinetic rate constant for chemical models that best fit degradation kinetic data. The input of the function is a data-frame organized as follows:

- 1. first columns, time data;
- 2. second columns, concentration data;
- 3. third column (optional, but highly recommended), experimental error

## Usage

```
det_order(dframe)
```

## **Arguments**

dframe

a data-frame with 2 or 3 columns, containing time, concentrations, and (optional) error data.

FOMT 5

#### Value

A ord\_res object containing in a list the following information:

- 1. the phase space coordinates of transformed data;
- 2. the linear regression performed in the phase space;
- 3. a boolean variable indicating if the estimate of the degradation rate constant is statistically significant;
- 4. non-linear regression performed using a n^th^-order kinetic model (if n=0 the regression is linear);
- 5. the data-frame given as the input;
- 6. the estimated reaction order.

#### See Also

results() to print the results or goodness\_of\_fit() to visualize the major goodness-of-fit measures; plot\_ord() to plot the regressions in both the phase and conventional spaces; kin\_regr() to extract the best kinetic model that explain the data and phase\_space() to extract the linear regression in the phase space.

## **Examples**

```
t <- c(0, 4, 8, 12, 16, 20)
conc <- c(1, 0.51, 0.24, 0.12, 0.07, 0.02)
err <- c(0.02, 0.05, 0.04, 0.04, 0.03, 0.02)
dframe <- data.frame(t, conc)
res <- det_order(dframe)

class(res)

dframe2 <- data.frame(t, conc, err)
res2 <- det_order(dframe2)

res2[[5]] == dframe2</pre>
```

FOMT

First-Order Multi-Target model regression

#### **Description**

The function performs a non-linear regression using the first-order multi-target model. The model equation is:

$$\frac{S}{S_0} = 1 - (1 - e^{-kt})^m$$

where  $S/S_0$  is the fraction of surviving molecules, k is the average number of hits per time unit, m is the number of hits required to degrade the molecule, and t is time.

6 FOMT

## Usage

```
FOMT(dtframe)
```

## **Arguments**

dtframe

A data-frame containing 2 or 3 columns: time, normalized concentration and error (optional), respectively

#### **Details**

The FOMT model has been proposed as an alternative to the Weibull equation that is commonly used when the time-dependent behavior of the data significantly deviates from that predicted by standard chemical models.

#### Value

Returns the results of the regression as a nls object.

#### See Also

```
FOMTm(), par_est_FOMT()
```

```
t <- c(0, 4, 8, 12, 16, 20)
conc <- c(1, 0.98, 0.99, 0.67, 0.12, 0.03)
err <- c(0.02, 0.05, 0.04, 0.04, 0.03, 0.02)
dframe <- data.frame(t, conc, err)</pre>
FOMT <- FOMT(dframe)</pre>
plot(dframe[[1]], dframe[[2]])
arrows(dframe[[1]], dframe[[2]] + dframe[[3]],
  dframe[[1]], dframe[[2]] - dframe[[3]],
  length = 0
newt \leftarrow seq(0, 21, by = 0.1)
lines(newt, predict(FOMT, newdata = list(t = newt)))
dframe1 <- data.frame(t, conc)</pre>
FOMT1 <- FOMT(dframe1)</pre>
plot(dframe1[[1]], dframe1[[2]])
lines(newt, predict(FOMT1, newdata = list(t = newt)))
summary(FOMT)
summary(FOMT1)
```

fomtdata 7

fomtdata

Total CQA with ascorbic acid (FOMT model)

# Description

Degradation data of 1.2 mM 5-caffeoylquinic acid (5-CQA) in the presence of 1.2 mM of ascorbic acid at 37°C. The data refer to total CQA concentration.

## Usage

fomtdata

#### **Format**

A data frame with 8 rows and 2 columns:

time\_h Time in hours

tCQA\_AA Normalized concentration of total CQA measured at each time point

#### **Source**

Yusaku N. and Kuniyo I. (2013) Degradation Kinetics of Chlorogenic Acid at Various pH Values and Effects of Ascorbic Acid and Epigallocatechin Gallate on Its Stability under Alkaline Conditions, Journal of Agricultural and Food Chemistry, doi:10.1021/jf304105w, fig 2D, solid diamonds

FOMTm

First-Order Multi-Target model

## **Description**

Call the function to return the formula of the Single-Hit Multi-Target model (FOMT):

$$1 - (1 - e^{-kt})^n$$

# Usage

FOMTm(t, k, m)

#### **Arguments**

t time

k average number of hits per time unit

m minimum number of hits required to degrade the molecule

8 f\_gen

## Value

Returns calculated values using the formula of FOMT model It can be used inside the function nls as the RHS of the formula.

## See Also

```
FOMT(), par_est_FOMT(), stats::nls()
```

## **Examples**

```
t <- seq(0, 100, by = 1)
k <- 0.1
n <- 200
y <- FOMTm(t, k, n)
plot(t, y, type = "l")</pre>
```

f\_gen

Formula of an n-order model.

## **Description**

Given the reaction order n , the function returns the equation corresponding to that particular n^th^order kinetic model. For  $n \neq 1$ :

$$y(t) = ((n-1)kt + y_0^{1-n})^{\frac{1}{n-1}}$$
$$y(t) = y_0 e^{-kt}$$

Usage

f\_gen(n)

for n = 1:

# Arguments

n

reaction order

## Value

A formula object containing the equation of the selected  $n^{h}$  order kinetic model.

goodness\_of\_fit 9

goodness\_of\_fit

Goodness-of-fit, non-linear regression

## **Description**

Function that returns the following goodness-of-fit statistics for non-linear regression: AIC, AICc, BIC, RMSE and reduced Chi-squared.

## Usage

```
goodness_of_fit(fit)
```

## Arguments

fit

a nls, lm or ord\_res object

#### **Details**

The function returns the values of AIC, AICC, BIC, RMSE and reduced chi-squared ( $\chi^2_{red}$ ) for nls objects. If a linear model object is passed, the function returns its summary.

Given an ord\_res object (output of the function det\_order()), the function returns one of the results above depending on the model chosen to explain the data.

Because the chiquad\_red() function returns the value only with weighted data, the  $\chi^2_{red}$  will be returned only with weighted regressions.

#### Value

It returns a table with the values of AIC, AICc, BIC, RSME and reduced Chi squared. Single goodness-of-fit measures can be obtained as follows:

- 1. call standard R functions stats::AIC(), stats::BIC(), stats::sigma() for AIC, BIC and RMSE, respectively;
- 2. call chemdeg functions AICC() and chiquad\_red() for AICc and reduced chi-squared, respectively.

## See Also

```
stats::AIC(), AICC(), stats::BIC(), stats::sigma(), chiquad_red()
```

```
x <- c(1, 2, 3, 4, 5)
y <- c(1.2, 3.9, 8.6, 17.4, 26)
er <- c(0.5, 0.8, 0.5, 1.9, 1.2)
fit1 <- nls(y ~ k * x^2,
    data = list(x = x, y = y), start = list(k = 1),
    weights = 1 / er^2
)
goodness_of_fit(fit1)</pre>
```

10 kin\_regr

kin\_regr

Degradation kinetics

## **Description**

Returns from an ord\_res object either the linear or the non-linear regression of the degradation kinetics data.

# Usage

```
kin_regr(x)
```

## **Arguments**

Х

an ord\_res object

#### **Details**

After the analysis in the phase space for the determination of the reaction order, det\_order() performs either a linear or a non-linear regression of the kinetic data, depending on whether the reaction order is n=0 or n>0, respectively. To access the regression object call kin\_degr.

## Value

Returns either an nls or lm object based on the regression performed by the function det\_order().

## See Also

```
det_order(), phase_space(), results(), stats::lm()
```

```
t <- c(0, 4, 8, 12, 16, 20)
conc <- c(1, 0.51, 0.24, 0.12, 0.07, 0.02)
dframe <- data.frame(t, conc)
res <- det_order(dframe)
kin_regr(res)</pre>
```

ord1 11

ord1

First order kinetic data

## **Description**

Synthetic data from a first-order kinetic model with k=0.7

## Usage

ord1

## **Format**

A data frame with 6 rows and 3 columns:

t time

concentration simulated concentration data at each time point std.error simulated experimental error

par\_est\_FOMT

First-Order Multi-Target parameter starting values

# Description

par\_est\_FOMT estimates the starting values of the parameters of the first-order multi-target model from a data-set.

# Usage

```
par_est_FOMT(x, y = NULL)
```

## **Arguments**

у

x A data-frame with time and concentration in the first and second columns,respectively. Alternatively, it could be an array of time and y an array of concentrations.

Optional, an array of concentrations. To be inserted only if x is an array.

## Value

The function returns an array with the suggested initial values of parameters.

## See Also

```
FOMT(), FOMTm()
```

phase\_space

## **Examples**

```
t <- seq(0, 30, by = 6)
k <- 0.3
n <- 40
set.seed(100)
y <- FOMTm(t, k, n) * (1 + rnorm(length(t), 0, 0.05))

nlsFOMT <- nls(y ~ FOMTm(t, k, n),
    data = list(y = y, t = t),
    start = par_est_FOMT(t, y)
)
summary(nlsFOMT)</pre>
```

phase\_space

Phase space, linearized model

## **Description**

Given an ord\_res object, this function returns the linearized model that best fits the data in the phase space. ord\_res object can be obtained using the function det\_order().

## Usage

```
phase_space(x)
```

## **Arguments**

Х

an ord\_res object

## Value

Returns a 1m class object.

## See Also

```
det_order(), kin_regr(), results(), stats::lm()
```

```
t <- c(0, 4, 8, 12, 16, 20)
conc <- c(1, 0.51, 0.24, 0.12, 0.07, 0.02)
dframe <- data.frame(t, conc)
res <- det_order(dframe)

phase_space(res)</pre>
```

plot\_ord 13

plot\_ord

Plot of phase space and kinetic curve

## **Description**

The function plots the results obtained from a det\_order() function. Two plots are shown: one representing the transformed data in the phase space and the other the kinetic data in the conventional space along with their regression curves.

## Usage

```
plot_ord(ord_res)
```

## **Arguments**

```
ord_res an 'ord_res' object
```

#### Value

Two plots. The first representing the transformed data in the phase space and the other the kinetic data in the conventional space along with their regression curves. Black line represent the best regression curve, whereas green lines show the fits with the reaction order chosen.

# Examples

```
t <- c(0, 4, 8, 12, 16, 20)
conc <- c(1, 0.51, 0.24, 0.12, 0.07, 0.02)
dframe <- data.frame(t, conc)
res <- det_order(dframe)

plot_ord(res)</pre>
```

results

Summary of 'ord\_res' object

## **Description**

Returns the results of the analyses performed by det\_order() function.

## Usage

```
results(object)
```

## **Arguments**

```
object
```

an 'ord\_res' object

14 urfa

#### **Details**

The function prints:

- 1. the linear regression performed in the phase space, together with the estimated *n* value and its 95% confidence interval
- 2. a brief conclusion on the results obtained in the phase space stating which reaction order should be preferred
- 3. the (non-)linear regression performed with parameters associated statistics. If a non-linear regression has been performed, the most common goodness-of-fit measures calculated with goodness\_of\_fit() are printed

#### Value

It prints a summary of the analysis in the phase space, the reaction order, and the regression results.

## See Also

```
det_order(), kin_regr(), phase_space()
```

## **Examples**

```
t <- c(0, 4, 8, 12, 16, 20)
conc <- c(1, 0.51, 0.24, 0.12, 0.07, 0.02)
err <- c(0.02, 0.05, 0.04, 0.04, 0.03, 0.02)
dframe <- data.frame(t, conc, err)
res <- det_order(dframe)
results(res)
```

urfa

*Urfa pepper ascorbic acid degradation (2-nd order)* 

## **Description**

Data describing the degradation kinetics of ascorbic acid during dehydration of Urfa peppers. The peppers were treated with hot air at 55, 65 and 75 °C.

#### Usage

urfa

#### Format

A data frame with 8 rows and 4 columns:

**time\_min** time in minutes

AA\_55 normalized concentration of ascorbic acid of Urfa peppers dehydrated at 55°C

AA\_65 normalized concentration of ascorbic acid of Urfa peppers dehydrated at 65°C

AA\_75 normalized concentration of ascorbic acid of Urfa peppers dehydrated at 75°C

# Source

Ş. Dağhan, A. Yildirim, F. Mehmet Yilmaz, H. Vardin and M. Karaaslan (2018) *The effect of temperature and method of drying on isot (Urfa pepper) and its Vitamin C degradation kinetics*, Italian Journal of Food Science, doi:10.14674/IJFS1070, fig 5, hot-air

# **Index**

```
* datasets
                                                  stats::BIC(), 3, 4, 9
    fomtdata, 7
                                                  stats::dchisq(), 4
    ord1, 11
                                                  stats::lm(), 10, 12
    urfa, 14
                                                  stats::nls(),8
                                                  stats::sigma(), 3, 4, 9
AICC, 2
                                                  summary, 9
AICC(), 4, 9
                                                  urfa, 14
chiquad_red, 3
chiquad_red(), 3, 9
det_order, 4
det_order(), 9, 10, 12-14
f_gen, 8
FOMT, 5
FOMT(), 8, 11
fomtdata, 7
FOMTm, 7
FOMTm(), 6, 11
goodness_of_fit, 9
goodness_of_fit(), 3-5, 14
kin_regr, 10
kin_regr(), 5, 12, 14
nls, 6, 8
ord1, 11
par_est_FOMT, 11
par_est_FOMT(), 6, 8
phase_space, 12
phase_space(), 5, 10, 14
plot_ord, 13
plot\_ord(), 5
results, 13
results(), 5, 10, 12
stats::AIC(), 2-4, 9
```