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AICc

Second-order Akaike Information Criterion

## Description

Calculates second-order Akaike information criterion for one or several fitted model objects (AIC for small samples).

# Usage

```
AICc (object, ..., k = 2)
```

### **Arguments**

object	a fitted model object
	optionally more fitted model objects
k	the "penalty" per parameter to be used; the default $k = 2$ is the classical AIC

#### Value

If just one object is provided, returns a numeric value with the corresponding AICc; if more than one object are provided, returns a data frame with rows corresponding to the objects and columns representing the number of parameters in the model (df), AICc and the AIC.

# Author(s)

Kamil Bartoń

### References

Burnham, K. P. and Anderson, D. R (2002) *Model selection and multimodel inference: a practical information-theoretic approach*. 2nd ed.

### See Also

Akaike's An Information Criterion: AIC

2 Cement

Cement

Cement hardening data

# Description

Cement hardening data from Woods et al (1939).

### Usage

data (Cement)

#### **Format**

cement is a data frame with 5 variables. x1-x4 are four predictor variables expressed as a percentage of weight.

- x1 calcium aluminate
- x2 tricalcium silicate
- x3 tetracalcium alumino ferrite
- x4 dicalcium silicate
- y calories of heat evolved per gram of cement after 180 days of hardening

# Author(s)

Kamil Bartoń

#### Source

Woods H., Steinour H.H., Starke H.R. (1932) Effect of composition of Portland cement on heat evolved during hardening. *Industrial & Engineering Chemistry* 24, 1207-1214

### References

Burnham, K. P. and Anderson, D. R (2002) *Model selection and multimodel inference: a practical information-theoretic approach.* 2nd ed.

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|--|--|

# Description

runs models with all possible combinations of the explanatory variables in the supplied model.

### Usage

### **Arguments**

global.model	a fitted 'global' model object. Currently, it can be a lm, glm, gam, lme, lmer, sarlm or spautolm.
beta	logical should standardized coefficients be returned rather than normal ones?
eval	whether to evaluate and rank the models. If set to FALSE only a list of all possible model formulas is returned $$
rank	custom rank function (information criterion) to use instead AICc, e.g. ${\tt QAIC}$ or ${\tt BIC},$ See 'Details'
fixed	optional, either a single sided formula or a character vector giving names of terms to be included in all models
m.max	maximum number of terms to be included in single model, defaults to the number of terms in ${\tt global.model}$
	optional arguments for the rank function

### **Details**

rank is found by a call to match. fun and typically is specified as a function or a symbol (e.g. a backquoted name) or a character string specifying a function to be searched for from the environment of the call to lapply.

Function rank must be able to accept model as a first argument and must always return a scalar.

### Value

dredge returns a data.frame of class model.selection with models' coefficients, k, deviance/RSS, r-squared, AIC, AICc, etc. This depends on a type of model. Models are ordered according to AICc (lowest on top), or by rank function if specified. It has also a formulas attribute - a list containing all the model formulas. ...

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#### Note

Use the lmer from **lme4** library rather than the old lmer from **Matrix** package. Complex expressions (like  $\log(y)$ ) cannot be used as the response variable, the response must be specified as simple variable name. They may be used on the right side of the formula, though. Make sure there is no a na.action set to na.omit in global.model. This can result with models fitted to different data sets, if there are NA's present. dredge cannot handle nested model designs (formulas such as  $y \sim a/b$ ).

### Author(s)

Kamil Bartoń

#### References

Burnham, K. P. and Anderson, D. R (2002) *Model selection and multimodel inference: a practical information-theoretic approach*. 2nd ed.

#### See Also

get.models, model.avg. QAIC has examples of using custom rank function.

#### **Examples**

```
# Example from Burnham and Anderson (2002), page 100:
data(Cement)
lm1 <- lm(y ~ ., data = Cement)
dd <- dredge(lm1)
dd

#models with delta.aicc < 4
model.avg(get.models(dd, subset = delta < 4)) # get averaged coefficients

#or as a 95% confidence set:
top.models <- get.models(dd, cumsum(weight) <= .95)

model.avg(top.models) # get averaged coefficients

#topmost model:
top.models[[1]]</pre>
```

dRedging.functions Helper functions for package dRedging

### Description

beta.weights-computes standardized coefficients (beta weights) for the model; coeffs-gets model coefficients; getAllTerms-extracts independent variables' names from a fitted model; tTable-returns summary table of a fitted model with coefficients, standard errors, and p-values.

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### Usage

```
beta.weights (model)
coeffs (model)
getAllTerms (x, ...)
tTable (model, ...)
```

### **Arguments**

model a fitted model object

x a fitted model object or a formula

... other arguments, not used

### Author(s)

Kamil Bartoń

# See Also

dredge

get.models

Get models

## **Description**

Gets list of models from a model.selection object

# Usage

```
get.models(dd, subset = delta <= 4, ...)</pre>
```

### **Arguments**

dd object returned by dredge

subset subset of models

... additional parameters passed to update, for example, in lme/lmer one may

want to use method = "REML" while using "ML" for model selection

### Value

list of models.

## Author(s)

Kamil Bartoń

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#### See Also

```
dredge, model.avg
```

### **Examples**

```
# Mixed models:
require(nlme)
fm2 <- lme(distance ~ age + Sex, data = Orthodont, random = ~ 1 | Subject, method="ML")
dd2 <- dredge(fm2)
# Get top-most models, but fitted by REML:
(top.models.2 <- get.models(dd2, subset = delta < 4, method = "REML"))</pre>
```

model.avg

Model averaging

### **Description**

averages models according to an information criterion

### Usage

```
model.avg(ml, ..., beta = FALSE, method = c("0", "NA"), rank = NULL, rank.args = N
```

### **Arguments**

m1	a fitted model object or a list of such objects
	more fitted model objects
beta	logical, should standardized coefficients be returned rather than normal ones?
method	if set to "0" (default), terms missing in one model are assumed to be 0's, otherwise they are omitted from the weighted average $\frac{1}{2}$
rank	custom rank function (information criterion) to use instead of AICc, e.g. QAIC or BIC, may be omitted if m1 is a list returned by dredge. See 'Details'
rank.args	optional list of arguments for the rank function
alpha	significance level for calculating confidence intervals

#### **Details**

rank is found by a call to match. fun and typically is specified as a function or a symbol (e.g. a backquoted name) or a character string specifying a function to be searched for from the environment of the call to lapply.

Function rank must be able to accept model as a first argument and must always return a scalar.

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#### Value

model.avg returns a list with elements:

```
coefficients model coefficients

variance variance of coefficients

avg.model averaged model summary (data.frame with columns: coef - averaged coefficients, var - unconditional variance estimator, ase - adjusted standard error estimator, lci, uci - unconditional confidence intervals)

relative.importance relative variable importances

beta (logical) were standardized coefficients used?
```

#### Author(s)

Kamil Bartoń

#### References

Burnham, K. P. and Anderson, D. R (2002) Model selection and multimodel inference: a practical information-theoretic approach. 2nd ed.

#### See Also

dredge, get.models. QAIC has examples of using custom rank function.

### **Examples**

```
# Example from Burnham and Anderson (2002), page 100:
data(Cement)
lm1 <- lm(y \sim ., data = Cement)
dd <- dredge(lm1)
dd
#models with delta.aicc < 4
model.avg(get.models(dd, subset = delta < 4)) # get averaged coefficients</pre>
#or as a 95% confidence set:
top.models <- get.models(dd, cumsum(weight) <= .95)
model.avg(top.models) # get averaged coefficients
#topmost model:
top.models[[1]]
## Not run:
# using BIC (Schwarz's Bayesian criterion) to rank the models
BIC <- function(x) AIC(x, k = log(length(residuals(x))))
model.avg(top.models, rank=BIC)
```

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```
## End(Not run)
```

MuMIn-package

Multi model inference

#### **Description**

The MuMIn contains functions for model selection and model averaging, based on information criteria (AIC alike).

#### **Details**

User level functions include:

```
model.avg - does model averaging.
get.models - evaluates models from the table returned by dredge.
dredge - runs models with combinations of terms of the supplied 'global' model.
AICc - calculates second-order Akaike information criterion for one or several fitted model objects.
```

#### Author(s)

Kamil Bartoń (kbarton@zbs.bialowieza.pl)

#### References

Burnham, K. P. and Anderson, D. R (2002) *Model selection and multimodel inference: a practical information-theoretic approach*. 2nd ed.

#### See Also

```
AIC, step
```

# **Examples**

```
fm1 <- lm(Fertility ~ . , data = swiss)

dd <- dredge(fm1)
dd

#list of models with delta.aicc < 4
top.models.1 <- get.models(dd, subset = delta < 4)
model.avg(top.models.1) # get averaged coefficients

#or as a 95% confidence set:
top.models.2 <- get.models(dd, cumsum(weight) <= .95)

model.avg(top.models.2) # get averaged coefficients</pre>
```

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```
# Mixed models:
data(Orthodont, package="nlme")
require(nlme)
fm2 <- lme(distance ~ age + Sex, data = Orthodont, random = ~ 1 | Subject, method="ML")
dredge(fm2)

require(lme4)
fm3 <- lmer(distance ~ age + Sex + (1 | Subject), data = Orthodont, REML=FALSE)
dd3 <- dredge(fm3)

# Get top-most models, but fitted by REML:
(top.models.3 <- get.models(dd3, subset = delta < 4, REML=TRUE))
# use: method = "REML" for older versions of lme4</pre>
```

par.avg

Parameter averaging

### **Description**

averages single parameter based on given weights

#### Usage

```
par.avg (x, se, npar, weight, alpha = 0.05)
```

#### **Arguments**

x vector of parametersse vector of standard errors

npar vector giving numbers of estimated parameters

weight vector of weights

alpha significance level for calculating confidence intervals

#### Value

par.avg returns a vector with named elements:

Coefficient model coefficients

Variance unconditional variance of coefficients

Unconditional SE

Lower CI, Upper CI
relative variable importances

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#### Author(s)

Kamil Bartoń

#### References

Burnham, K. P. and Anderson, D. R (2002) *Model selection and multimodel inference: a practical information-theoretic approach*. 2nd ed.

#### See Also

model.avg for averaging models.

QAIC

Quasi AIC

### **Description**

Calculates "quasi AIC" for one or several fitted model objects. This function is provided just as an example of custom rank function for use with model.avg and dredge

### Usage

```
QAIC(object, ..., chat)
```

### **Arguments**

object a fitted model object.

... optionally more fitted model objects.

chat c-hat

#### **Details**

rank is specified as a function or a symbol (e.g. a backquoted name) or a character string specifying a function.

Function rank must be able to accept model as a first argument and must always return a scalar.

#### Value

If just one object is provided, returns a numeric value with the corresponding QAIC; if more than one object are provided, returns a data.frame with rows corresponding to the objects.

#### Author(s)

Kamil Bartoń

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### **Examples**

```
budworm <- data.frame(ldose = rep(0:5, 2), numdead = c(1, 4, 9, 13, 18, 20, 0, 2, 6, 10, 12, budworm$SF <- cbind(budworm$numdead, 20 - budworm$numdead)
budworm.qlg <- glm(SF ~ sex*ldose, family = quasibinomial, data = budworm)

dd1 <- dredge(budworm.qlg, rank = "QAIC", chat = summary(budworm.qlg)$dispersion)
gm1 <- get.models(dd1, 1:4)

model.avg(gm1)

model.avg(gm1[[1]], gm1[[2]], rank = "QAIC", rank.args = list(chat = 1))</pre>
```