

Package ‘bootStepAIC’

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Title Bootstrap stepAIC

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Description Model selection by bootstrapping the stepAIC() procedure.

Depends R(>= 4.0.0), MASS

LazyLoad yes

License GPL (>= 2)

NeedsCompilation no

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boot.stepAIC	<i>Bootstraps the Stepwise Algorithm of stepAIC() for Choosing a Model by AIC</i>
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Description

Implements a Bootstrap procedure to investigate the variability of model selection under the stepAIC() stepwise algorithm of package MASS.

Usage

```
boot.stepAIC(object, data, B = 100, alpha = 0.05, direction = "backward",  
             k = 2, verbose = FALSE, seed = 1L, ...)
```

Arguments

object	an object representing a model of an appropriate class; currently, "lm", "aov", "glm", "negbin", "polr", "survreg", and "coxph" objects are supported.
data	a data.frame or a matrix that contains the response variable and covariates.
B	the number of Bootstrap samples.
alpha	the significance level.
direction	the direction argument of stepAIC().
k	the k argument of stepAIC().
verbose	logical; if TRUE information about the evolution of the procedure is printed in the screen.
seed	numeric scalar denoting the seed used to create the Bootstrap samples.
...	extra arguments to stepAIC(), e.g., scope.

Details

The following procedure is replicated B times:

Step 1: Simulate a new data-set taking a sample with replacement from the rows of data.

Step 2: Refit the model using the data-set from Step 1.

Step 3: For the refitted model of Step 2 run the stepAIC() algorithm.

Summarize the results by counting how many times (out of the B data-sets) each variable was selected, how many times the estimate of the regression coefficient of each variable (out of the times it was selected) it was statistically significant in significance level alpha, and how many times the estimate of the regression coefficient of each variable (out of the times it was selected) changed signs (see also Austin and Tu, 2004).

Value

An object of class BootStep with components

Covariates	a numeric matrix containing the percentage of times each variable was selected.
Sign	a numeric matrix containing the percentage of times the regression coefficient of each variable had sign + and -.
Significance	a numeric matrix containing the percentage of times the regression coefficient of each variable was significant under the alpha significance level.
OrigModel	a copy of object.
OrigStepAIC	the result of applying stepAIC() in object.
direction	a copy of the direction argument.
k	a copy of the k argument.
BootStepAIC	a list of length B containing the results of stepAIC() for each Bootstrap data-set.

Author(s)

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References

Austin, P. and Tu, J. (2004). Bootstrap methods for developing predictive models, *The American Statistician*, **58**, 131–137.

Venables, W. N. and Ripley, B. D. (2002). *Modern Applied Statistics with S*, 4th ed. Springer, New York.

See Also

stepAIC in package MASS

Examples

```
## lm() Example ##
n <- 350
x1 <- runif(n, -4, 4)
x2 <- runif(n, -4, 4)
x3 <- runif(n, -4, 4)
x4 <- runif(n, -4, 4)
x5 <- runif(n, -4, 4)
x6 <- runif(n, -4, 4)
x7 <- factor(sample(letters[1:3], n, rep = TRUE))
y <- 5 + 3 * x1 + 2 * x2 - 1.5 * x3 - 0.8 * x4 + rnorm(n, sd = 2.5)
data <- data.frame(y, x1, x2, x3, x4, x5, x6, x7)
rm(n, x1, x2, x3, x4, x5, x6, x7, y)

lmFit <- lm(y ~ (. - x7) * x7, data = data)
boot.stepAIC(lmFit, data)

#####

## glm() Example ##
n <- 200
x1 <- runif(n, -3, 3)
x2 <- runif(n, -3, 3)
x3 <- runif(n, -3, 3)
x4 <- runif(n, -3, 3)
x5 <- factor(sample(letters[1:2], n, rep = TRUE))
eta <- 0.1 + 1.6 * x1 - 2.5 * as.numeric(as.character(x5) == levels(x5)[1])
y1 <- rbinom(n, 1, plogis(eta))
y2 <- rbinom(n, 1, 0.6)
data <- data.frame(y1, y2, x1, x2, x3, x4, x5)
rm(n, x1, x2, x3, x4, x5, eta, y1, y2)

glmFit1 <- glm(y1 ~ x1 + x2 + x3 + x4 + x5, family = binomial(), data = data)
glmFit2 <- glm(y2 ~ x1 + x2 + x3 + x4 + x5, family = binomial(), data = data)

boot.stepAIC(glmFit1, data, B = 50)
boot.stepAIC(glmFit2, data, B = 50)
```

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* **regression**

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