# Package 'speedglm'

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Type Package

Title Fitting Linear and Generalized Linear Models to Large Data Sets

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Depends Matrix, MASS

Imports methods, stats

Description

Fitting linear models and generalized linear models to large data sets by updating algorithms.

License GPL

LazyLoad yes

NeedsCompilation no

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### Description

Fits Linear and Generalized Linear Models to large data sets. For data loaded in R memory the fitting is usually fast, especially if R is linked against an optimized BLAS. For data sets of size greater of R memory, the fitting is made by an updating algorithm.

### Details

Package:	speedglm
Type:	Package
Version:	0.3-1
Date:	2015-09-09
Depends:	Matrix, stats, MASS
License:	GPL
LazyLoad:	yes

### Author(s)

Marco Enea <emarco76@libero.it>, with contributions from Ronen Meiri and Tomer Kalimi (on behalf of DMWay Analytics LTD).

Maintainer: Marco Enea <emarco76@libero.it>

add1.speedlm	Wrappers to the add1 and drop1 methods for speedlm and speedglm
	objects

# Description

These are adviced to be used for speedlm and speedglm models fitted on moderately large data sets. It is also possible to use stepAIC function from package MASS.

# Usage

```
drop1(object, scope, scale = 0, all.cols = TRUE,
                        test = c("none","Chisq", "F"), k = 2, data, ...)
## S3 method for class 'speedlm'
extractAIC(fit, scale = 0, k=2,...)
## S3 method for class 'speedlm'
nobs(object, use.fallback = FALSE, ...)
## S3 method for class 'speedglm'
## S3 method for class 'speedglm'
add1(object, scope, scale = 0, test = c("none", "Rao", "LRT",
                    "Chisq", "F"), x = NULL, k = 2, weights=rep(1,object$n), ...)
## S3 method for class 'speedglm'
drop1(object, scope, scale = 0, test = c("none", "Rao", "LRT",
                         "Chisq", "F"), k = 2, weights=rep(1,object$n), ...)
## S3 method for class 'speedglm'
extractAIC(fit, scale = 0, k=2,...)
## S3 method for class 'speedglm'
nobs(object, use.fallback = FALSE, ...)
```

### Arguments

object	a speedlm or speedglm object for which model=TRUE was previously set to.
fit	a speedlm or speedglm object
scope	see add1 from package stats.
scale	see add1 from package stats.
all.cols	see drop1 from package stats.
test	see add1 from package stats.
x	see add1 from package stats.
k	see add1 from package stats.
data	the data that the model was previously fitted to. If not provided, these will be searched in the parent environment.
weights	the model weights, if provided in the speedglm object
use.fallback	logical. Should fallback methods be used to try to guess the value?
	further optional arguments.

# Details

It is possible to use functions step() and stepAIC() for both speedlm and speedglm objects but objects fitted using updateWithMoreData()

### Warnings

Note that these functions have been poorly tested and need to be checked out more carefully.

control

### Author(s)

Ronen Meiri and Marco Enea

### Examples

```
## Not run:
set.seed(10)
n <- 1000
k <- 3
x <- round(matrix(rnorm(n * k), n, k), digits = 3)</pre>
beta <- c(0.05,0.5,0.8,1.3,1.8)
y \leq drop(tcrossprod(cbind(1,x,x[,2]*x[,3]),t(beta))) + rnorm(n,,0.2)
colnames(x) <- c("s1", "s2", "s3")
da <- data.frame(y, x)</pre>
m0 <- speedlm(y ~ 1, data = da,model=TRUE, y=TRUE)</pre>
m0.1 <- add1(m0,scope=~(s1+s2+s3)^2, data = da)</pre>
m1 <- step(m0,scope=~(s1+s2+s3)^3)</pre>
m1
m2 <- speedglm(y ~ 1, data = da, model=TRUE, y=TRUE)</pre>
m2.1 <- add1(m2,scope=~(s1+s2+s3)^2, data = da)</pre>
m3 <- step(m2,scope=~(s1+s2+s3)^3)</pre>
m3
## End(Not run)
```

control

#### Miscellanea of functions

### Description

Utility functions for least squares estimation in large data sets.

### Usage

# Arguments

В	a squared matrix.
symmetric	logical, is B symmetric?
tol.values	tolerance to be consider eigenvalues equals to zero.
tol.vectors	tolerance to be consider eigenvectors equals to zero.

### control

out.B	Have the matrix B to be returned?
method	the method to check for singularity. By default is "eigen", and an eigendecom- position of X'X is made. The "Cholesky" method is faster than "eigen" and does not use tolerance, but the former seems to be more stable for opportune tolerance values.
Х	the model matrix.
w	a weights vector.
sparse	logical, is X sparse?
sparselim	a real in the interval [0; 1]. It indicates the minimal proportion of zeroes in the data matrix X in order to consider X as sparse eigendec Logical. Do you want to investigate on rank of X? You may set to
row.chunk	an integer which indicates the total rows number compounding each of the first g-1 blocks. If row.chunk is not a divisor of $nrow(X)$ , the g-th block will be formed by the remaining data.
camp	the sample proportion of elements of X on which the survey will be based.

### Details

Function control makes an eigendecomposition of B according established values of tolerance. Function cp makes the cross-product X'X by partitioning X in row-blocks. When an optimized BLAS, such as ATLAS, is not installed, the function represents an attempt to speed up the calculation and avoid overflows with medium-large data sets loaded in R memory. The results depending on processor type. Good results are obtained, for example, with an AMD Athlon dual core 1.5 Gb RAM by setting row.chunk to some value less than 1000. Try the example below by changing the matrix size and the value of row.chunk. If the matrix X is sparse, it will have class "dgCMatrix" (the package Matrix is required) and the cross-product will be made without partitioning. However, good performances are usually obtained with a very high zeroes proportion. Function is.sparse makes a quick sample survey on sample proportion of zeroes in X.

### Value

for the function control, a list with the following elements:

ХТХ	the matrix product B without singularities (if there are).
rank	the rank of B
pivot	an ordered set of column indeces of B with, if the case, the last $rank + 1,, p$ columns which indicate possible linear combinations.
for the function cp:	

given).

new.B	the matrix product X'X	(weighted, if w is

for the function is.sparse:

sparse a logical value which indicates if the sample proportion of zeroes is greater than sparselim, with the sample proportion as attribute.

control

### Author(s)

Marco ENEA

# See Also

eigen, chol, qr, crossprod

# Examples

```
#### example 1.
n <- 100000
k <- 100
x <- round(matrix(rnorm(n*k),n,k),digits=4)</pre>
y <- rnorm(n)
# if an optimized BLAS is not installed, depending on processor type, cp() may be
# faster than crossprod() for large matrices.
system.time(a1 <- crossprod(x))</pre>
system.time(a2 <- cp(x, row.chunk = 500))
all.equal(a1, a2)
#### example 2.1.
n <- 100000
k <- 10
x <- matrix(rnorm(n*k),n,k)</pre>
x[,2] <- x[,1] + 2*x[,3] # x has rank 9
y <- rnorm(n)
# estimation by least squares
A <- function(){
  A1 <- control(crossprod(x))
  ok <- A1$pivot[1:A1$rank]</pre>
  as.vector(solve(A1$XTX,crossprod(x[,ok],y)))
}
# estimation by QR decomposition
B <- function(){</pre>
 B1 <- qr(x)
  qr.solve(x[,B1$pivot[1:B1$rank]],y)
}
system.time(a <- A())</pre>
system.time(b <- B())</pre>
all.equal(a,b)
### example 2.2
x <- matrix(c(1:5, (1:5)<sup>2</sup>), 5, 2)
x <- cbind(x, x[, 1] + 3*x[, 2])</pre>
m <- crossprod(x)</pre>
qr(m)$rank # is 2, as it should be
control(m,method="eigen")$rank # is 2, as it should be
```

# data1

```
control(m,method="Cholesky")$rank # is wrong
#### example 3.
n <- 10000
fat1 <- gl(20,500)
y <- rnorm(n)
da <- data.frame(y,fat1)
m <- model.matrix(y ~ factor(fat1),data = da)
is.sparse(m)</pre>
```

data1

A toy dataset

# Description

The data1 dataset has 100 rows and 4 columns.

# Usage

data(data1)

## Format

A data frame with 100 observations on the following 4 variables.

y a gamma-distributed response variable

fat1 a four-level factor

- x1 a numeric covariate
- x2 a numeric covariate

# Details

This is a toy dataset used to show how function shglm works.

# Examples

data(data1)

predict.speedglm Predict method for a speedglm object

# Description

summary The method is currently under construction but some functionalities are available.

# Usage

### Arguments

object	an object of class 'speedlgm'.
newdata	An optional data frame with new data or the original data.
type	Type of prediction.
na.action	function determining what should be done with missing values in newdata.
	further optional arguments

# Details

If newdata is omitted prediction are based on the data used for the fit only if argument fitted was previously set to TRUE in the speedglm object. Currently the method does not work for function shglm.

### Value

pred a vector of predictions.

# Author(s)

Tomer Kalimi and Marco Enea

### See Also

speedglm

# predict.speedlm

### Examples

```
set.seed(10)
y <- rgamma(20,1.5,1)
x <-round( matrix(rnorm(20*2),20,2),digits=3)
colnames(x) <-c("s1","s2")
da <- data.frame(y,x)
mod <- speedglm(y~s1+s2, data=da, family=Gamma(log), fitted=TRUE)
predict(mod)</pre>
```

predict.speedlm Predict method for a speedlm object

### Description

summary The method is currently under construction but some functionalities are available.

### Usage

```
## S3 method for class 'speedlm'
predict(object, newdata, na.action = na.pass, ...)
```

### Arguments

object	an object of class 'speedlm'.
newdata	An optional data frame with new data or the original data.
na.action	function determining what should be done with missing values in newdata.
	further optional arguments

### Details

If newdata is omitted prediction are based on the data used for the fit only if argument fitted was previously set to TRUE in the speedlm object.

### Value

predictor a vector of predictions.

### Author(s)

Tomer Kalimi and Marco Enea

### See Also

speedlm

### Examples

```
set.seed(10)
x <- round( matrix(rnorm(20*3),20,3),digits=3)
colnames(x) <-c("y","s1","s2")
da <- as.data.frame(x)
mod <- speedlm(y~s1+s2, data=da, fitted=TRUE)
predict(mod)</pre>
```

speedglm

Fitting Generalized Linear Models for Large Data Sets

### Description

speedglm and speedglm.wfit fit GLMs to medium-large data sets, that is those storable into the R memory. The highest performances, in terms of computation time, are obtained when R is linked against an optimized BLAS, such as ATLAS. The function shglm is for a data set stored into a file of size greater than the available memory, and takes as argument a function to manipulate connections.

### Usage

```
## S3 method for class 'data.frame':
speedglm(formula,data,family=gaussian(),weights=NULL,start=NULL,
         etastart=NULL,mustart=NULL,offset=NULL,maxit=25, k=2,
         sparse=NULL,set.default=list(), trace=FALSE,
         method=c('eigen','Cholesky','qr'), model=FALSE, y=FALSE,
         fitted=FALSE,...)
## S3 method for class 'matrix':
speedglm.wfit(y, X, intercept=TRUE, weights=NULL,row.chunk=NULL,
              family=gaussian(), start=NULL, etastart=NULL,
              mustart=NULL, offset=NULL, acc=1e-08, maxit=25, k=2,
              sparselim=.9,camp=.01, eigendec=TRUE, tol.values=1e-7,
              tol.vectors=1e-7, tol.solve=.Machine$double.eps,
              sparse=NULL,method = c('eigen','Cholesky','qr'),
              trace=FALSE,...)
## S3 method for class 'function':
shglm(formula, datafun, family = gaussian(), weights.fo = NULL, start = NULL,
     etastart = NULL, mustart = NULL, offset = NULL, maxit = 25, k = 2,
     chunksize = 5000, sparse = NULL, trace = FALSE, all.levels = FALSE,
      set.default = list(),...)
```

# speedglm

# Arguments

-	Most of arguments are the same of glm or bigglm but with some difference.
	the same of glm.
<b>đeta</b> ula	a data frame.
datafun	a function which uses connections. See the example below.
family	the same of glm, but it must be specified with brackets.
start	the same of glm.
weights	the same of glm, but it must be specified as data\$weights.
weights.fo	weights for the response. It must be specified as a formula (see the example below).
etastart	the same of glm.
mustart	the same of glm.
offset	the same of glm.
intercept	the same of glm.
Х	the same of x in glm.fit.
У	the same of glm and glm.fit.
maxit	the same of glm.
k	numeric, the penalty per parameter to be used; the default $k = 2$ is the classical AIC.
trace	logical. Do you want to be informed about the model estimation progress?
sparse	logical. Is the model matrix sparse? By default is NULL, so a quickly sample survey will be made.
chunksize	an integer indicates the number of rows of the data file to read at time.
all.levels	logical, are all factor's levels present in each data chunk?
set.default	a list in which to specify the below parameters.
sparselim	a real in the interval [0, 1]. It indicates the minimal proportion of zeroes in the data matrix X in order to consider X as sparse.
camp	see the function is.sparse.
eigendec	logical. Do you want to check the rank of X? You may set it to false if you are sure that X is full rank.
row.chunk	an integer, see the function cp for details.
асс	tolerance to be used for the estimation.
tol.solve	see the function solve.
tol.values	see the function control.
tol.vectors	see the function control.
method	the chosen method to detect for singulatity.
model	logical. If TRUE the model frame will be returned.
fitted	logical. If TRUE the fitted values will be returned.
	further optional arguments.

### Details

The function shglm works like biglm, but it checks for singularity and does not impose restrictions on factors. Since during the IWLS estimation shglm uses repeated accesses to data file stored, for example, into the hard disk, the estimation time could be very long. Unlike from glm or biglm, the functions of class 'speedglm' do not use the QR decomposition, but directly solve the equations in the form of Iterative(-ly) (Re-)Weighted Least Squares (IWLS). The memory size of an object of class 'speedglm' is  $O(p^2)$ , where p is the number of covariates, unless one or more of argument model, y and fitted are set to TRUE. If an optimized BLAS is not installed, an attempt to speed up calculations might be done by setting row.chunk to some value, usually less than 1000, in set.default. See the function cp for details.

If the model matrix is (very) sparse, the package Matrix could be used. Note that if method 'qr' is chosen, then the qr decomposition will not be applied on matrix X, as in lm, but on X'WX.

### Value

coefficients	the estimated coefficients.
logLik	the log likelihood of the fitted model.
iter	the number of iterations of IWLS used.
tol	the maximal value of tolerance reached.
convergence	a logical value which indicates if convergence was reached.
family	the family object used.
link	the link function used.
df	the degrees of freedom of the model.
XTX	the product X'X (weighted, if the case).
dispersion	the estimated dispersion parameter of the model.
ok	the set of column indeces of the model matrix where the model has been fitted.
rank	the rank of the model matrix.
RSS	the estimated residual sum of squares of the fitted model.
aic	the estimated Akaike Information Criterion.
sparse	a logical value which indicates if the model matrix is sparse.
deviance	the estimated deviance of the fitted model.
nulldf	the degrees of freedom of the null model.
nulldev	the estimated deviance of the null model.
ngoodobs	the number of non-zero weighted observations.
n	the number of observations.
intercept	a logical value which indicates if an intercept has been used.
terms	the terms object used.
call	the matched call.
model	Either NULL or, if model was previously set to TRUE, the model frame.
У	Either NULL or, if y was previously set to TRUE, the response variable.
linear.predict	
	Fither NILL or if fitted was previously set to TPLIE the fitted values

Either NULL or, if fitted was previously set to TRUE, the fitted values.

### speedglm

### Note

All the above functions make an object of class 'speedglm'. In the current package version, arguments start, mustart and etastart of function shglm have been disabled. These will be restored in future.

### Author(s)

Marco Enea. Ronen Meiri contributed with method 'qr'

### References

Enea, M. (2009) Fitting Linear Models and Generalized Linear Models with large data sets in R. In book of short papers, conference on "Statistical Methods for the analysis of large data-sets", Italian Statistical Society, Chieti-Pescara, 23-25 September 2009, 411-414.

Bates, D. (2009) Comparing Least Square Calculations. Technical report. Available at http: //cran.rakanu.com/web/packages/Matrix/vignettes/Comparisons.pdf

Lumley, T. (2009) biglm: bounded memory linear and generalized linear models. *R package version* 0.7. https://CRAN.R-project.org/package=biglm.

#### See Also

speedlm, bigglm, glm

### Examples

```
## Not run:
# The following comparison among glm(), bigglm() and speedglm() cannot be considered rigorous
# and exhaustive, but it is only to give an idea of the computation time.
# It may take a long time.
require(biglm)
n<-50000
k<-80
y <- rgamma(n, 1.5, 1)
x <-round( matrix(rnorm(n*k),n,k),digits=3)</pre>
colnames(x) <-paste("s",1:k,sep = "")</pre>
da<- data.frame(y,x)</pre>
fo <- as.formula(paste("y~",paste(paste("s",1:k,sep=""),collapse="+")))</pre>
system.time(m1 <- glm(fo,data=da,family=Gamma(log)))</pre>
system.time(m2 <- bigglm(fo,data=da,family=Gamma(log)))</pre>
system.time(m3 <- speedglm(fo,data=da,family=Gamma(log)))</pre>
# You may also try speedglm when R is linked against an optimized BLAS,
# otherwise try to run the following function. In some computers, it is
# faster for large data sets.
system.time(m4 <- speedglm(fo,data=da,family=Gamma(log),set.default=list(row.chunk=1000)))</pre>
```

```
## End(Not run)
#####################
## Not run:
## An example of function using a connection to an out-memory file
## This is a slightly modified version of the function from the bigglm's help page
make.data<-function(filename, chunksize,...){</pre>
     conn<-NULL
     function(reset=FALSE){
     if(reset){
       if(!is.null(conn)) close(conn)
       conn<<-file(filename,open="r")</pre>
     } else{
       rval<-read.table(conn, nrows=chunksize,...)</pre>
       if ((nrow(rval)==0)) {
            close(conn)
            conn<<-NULL
            rval<-NULL
       }
       return(rval)
     }
 }
}
# data1 is a small toy dataset
data(data1)
write.table(data1,"data1.txt",row.names=FALSE,col.names=FALSE)
rm(data1)
da<-make.data("data1.txt",chunksize=50,col.names=c("y","fat1","x1","x2"))</pre>
# Caution! make sure to close the connection once you have run command #1
da(reset=T) #1: opens the connection to "data1.txt"
da(reset=F) #2: reads the first 50 rows (out of 100) of the dataset
da(reset=F) #3: reads the second 50 rows (out of 100) of the dataset
da(reset=F) #4: is NULL: this latter command closes the connection
require(biglm)
# fat1 is a factor with four levels
b1<-shglm(y~factor(fat1)+x1,weights=~I(x2^2),datafun=da,family=Gamma(log))</pre>
b2<-bigglm(y~factor(fat1)+x1,weights=~I(x2^2),data=da,family=Gamma(log))</pre>
summary(b1)
summary(b2)
file.remove("data1.txt")
## End(Not run)
```

speedlm

Fitting Linear Models to Large Data Sets

### speedlm

### Description

The functions of class 'speedlm' may speed up the fitting of LMs to large data sets. High performances can be obtained especially if R is linked against an optimized BLAS, such as ATLAS.

### Usage

```
# S3 method of class 'data.frame'
speedlm(formula, data, weights = NULL, offset = NULL, sparse = NULL,
        set.default = list(), method=c('eigen','Cholesky','qr'),
       model = FALSE, y = FALSE, fitted = FALSE, subset=NULL, ...)
# S3 method of class 'matrix'
speedlm.fit(y, X, intercept = FALSE, offset = NULL, row.chunk = NULL,
            sparselim = 0.9, camp = 0.01, eigendec = TRUE,
            tol.solve = .Machine$double.eps, sparse = NULL, tol.values = 1e-07,
            tol.vectors = 1e-07, method=c('eigen','Cholesky','qr'), ...)
speedlm.wfit(y, X, w, intercept = FALSE, offset = NULL, row.chunk = NULL,
             sparselim = 0.9, camp = 0.01, eigendec = TRUE,
            tol.solve = .Machine$double.eps, sparse = NULL, tol.values = 1e-07,
             tol.vectors = 1e-07, method=c('eigen', 'Cholesky', 'qr'), ...)
# S3 method of class 'speedlm' (object) and 'data.frame' (data)
## S3 method for class 'speedlm'
update(object, formula, data, add=TRUE, evaluate=TRUE,
                         subset=NULL,...)
# S3 method of class 'speedlm' (object) and 'data.frame' (data)
updateWithMoreData(object, data, weights = NULL, offset = NULL, sparse = NULL,
```

```
all.levels = FALSE, set.default = list(), subset=NULL, ...)
```

### Arguments

	Most of arguments are the same of functions lm but with some difference.
	the same of function 1m.
<b>đetm</b> ula	the same of function 1m, but it must always specified.
weights	the same of function lm, but it must be specified as data\$weights.
w	the same of weights.
intercept	a logical value which indicates if an intercept is used.
offset	the same of function 1m.
Х	the same of x in function lm.
У	the same of lm,lm.wfit and lm.fit.
sparse	logical. Is the model matrix sparse? By default is NULL, so a quickly sample survey will be made.
set.default	a list in which to specify the parameters to pass to the functions cp, control and is.sparse.

sparselim	a value in the interval [0, 1]. It indicates the minimal proportion of zeroes, in the model matrix X, in order to consider X as sparse.
camp	see function is.sparse.
eigendec	logical. Do you want to investigate on rank of X? You may set it to false if you are sure that X is full rank.
row.chunk	an integer, see the function cp for details.
tol.solve	see function solve.
tol.values	see function control.
tol.vectors	see function control.
method	see function control.
object	an object of class 'speedlm'.
all.levels	are all levels of eventual factors present in each data chunk? If so, set all.levels to true to speed up the fitting.
model	logical. Should the model frame be returned?
fitted	logical. Should the fitted values be returned?
subset	the same of function lm
add	logical. Are additional data coming from a new chunk provided?
evaluate	logical. If true evaluate the new call else return the call.
	further optional arguments.

### Details

Unlikely from lm or biglm, the functions of class 'speedlm' do not use the QR decomposition but directly solve the normal equations. Further, the most recent version of the package include method 'qr'. However such qr decomposition is not applied directly on matrix X, but on X'WX. In some extreme case, this might have some problem of numerical stability but may take advantage from the use of an optimized BLAS. The memory size of an object of class 'speedlm' is  $O(p^2)$ , where p is the number of covariates. If an optimized BLAS library is not installed, an attempt to speed up calculations may be done by setting row.chunk to some value, usually less than 1000, in set.default. See the function cp for details. Factors are permitted without limitations.

In the most recent versions, function update.speedlm is now a wrapper to call either updateWithMoreData (the new name of the old update.speedlm, for additional data chunks), or update from package stats.

### Value

coefficients	the estimated coefficients.
df.residual	the residual degrees of freedom.
ХТХ	the product X'X (weighted, if the case).
А	the product X'X (weighted, if the case) not checked for singularity.
Ху	the product X'y (weighted, if the case).
ok	the set of column indeces of the model matrix where the model has been fitted.

# speedlm

rank	the numeric rank of the fitted linear model.
pivot	see the function control.
RSS	the estimated residual sums of squares of the fitted model.
sparse	a logical value indicating if the model matrix is sparse.
deviance	the estimated deviance of the fitted model.
weigths	the weights used in the last updating.
zero.w	the number of non-zero weighted observations.
nobs	the number of observations.
nvar	the number of independent variables.
terms	the terms object used.
intercept	a logical value which indicates if an intercept has been used.
call	the matched call.
model	Either NULL or the model frame, if model was previously set to TRUE.
У	Either NULL or the response variable, if y was previously set to TRUE.
fitted.values	Either NULL or the fitted values, if fitted was previously set to TRUE.
	others values necessary to update the estimation.

## Note

All the above functions make an object of class 'speedlm'.

### Author(s)

Marco Enea, with contribution from Ronen Meiri.

### References

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# See Also

summary.speedlm,speedglm, lm, and biglm

speedlm

### Examples

```
## Not run:
n <- 1000
k <- 3
y <- rnorm(n)
x <- round(matrix(rnorm(n * k), n, k), digits = 3)</pre>
colnames(x) <- c("s1", "s2", "s3")
da <- data.frame(y, x)</pre>
do1 <- da[1:300,]
do2 <- da[301:700,]
do3 <- da[701:1000,]
m1 <- speedlm(y \sim s1 + s2 + s3, data = do1)
m1 <- update(m1, data = do2)
m1 <- update(m1, data = do3)</pre>
m2 <- lm(y \sim s1 + s2 + s3, data = da)
summary(m1)
summary(m2)
## End(Not run)
## Not run:
  # as before but recursively
  make.data <- function(filename, chunksize,...){</pre>
    conn <- NULL
    function(reset=FALSE, header=TRUE){
      if(reset){
        if(!is.null(conn)) close(conn)
         conn<<-file(filename,open="r")</pre>
      } else{
        rval <- read.table(conn, nrows=chunksize,header=header,...)</pre>
        if (nrow(rval)==0) {
          close(conn)
          conn<<-NULL
          rval<-NULL
        }
        return(rval)
      }
    }
  }
  write.table(da,"da.txt",col.names=TRUE,row.names=FALSE,quote=FALSE)
  x.names <- c("s1", "s2", "s3")
  dat <- make.data("da.txt",chunksize=300,col.names=c("y",x.names))</pre>
  dat(reset=TRUE)
  da2 <- dat(reset=FALSE)</pre>
  # the first model runs on the first 300 rows.
  m3 <- speedlm(y ~ s1 + s2 + s3, data=da2)
  # the last three models run on the subsequent 300, 300 and 100 rows, respectively
```

# summary.speedglm

```
for (i in 1:3){
    da2 <- dat(reset=FALSE, header=FALSE)
    m3 <- update(m3, data=da2, add=TRUE)
    }
    all.equal(coef(m1),coef(m3))
    file.remove("da.txt")
## End(Not run)</pre>
```

summary.speedglm Methods to summarize Generalized Linear Models fits

# Description

summary method for the class 'speedglm'.

# Usage

```
## S3 method for class 'speedglm'
summary(object,correlation=FALSE,...)
## S3 method for class 'speedglm'
coef(object,...)
## S3 method for class 'speedglm'
vcov(object,...)
## S3 method for class 'speedglm'
logLik(object,...)
## S3 method for class 'speedglm'
AIC(object,...)
```

# Arguments

object	an object of class 'speedglm'.
correlation	logical. Do you want to print the correlation matrix? By default it is false.
	further optional arguments

### Value

coefficients	the matrix of coefficients, standard errors, z-statistics and two-side p-values.
df.residual	the component from object.
df.null	the component from object.
null.deviance	the component from object.
deviance	the component from object.
family	the component from object.
call	the component from object.

AIC	the Akaike Information Criterion.
RSS	Residuals sums of squares.
correlation	(only if correlation is true.) The correlations of the estimated coefficients.
logLik	the log-likelihood value.
rank	the component from object.
dispersion	the estimated dispersion parameter of the fitted model.
convergence	the component from object.
iter	the component from object.
tol	the component from object.

### Author(s)

Marco ENEA

# See Also

speedglm

### Examples

```
n<-1000
k<-5
y <- rgamma(n,1.5,1)
x <-round( matrix(rnorm(n*k),n,k),digits=3)
colnames(x) <-paste("s",1:k,sep = "")
da<- data.frame(y,x)
fo <- as.formula(paste("y~",paste(paste("s",1:k,sep=""),collapse="+")))
m4 <- speedglm(fo,data=da,family=Gamma(log))
summary(m4)
```

summary.speedlm Methods to summarize Linear Models fits

# Description

summary method for class 'speedlm'.

# Usage

```
## S3 method for class 'speedlm'
summary(object, correlation = FALSE,...)
## S3 method for class 'speedlm'
coef(object,...)
## S3 method for class 'speedlm'
vcov(object,...)
```

# summary.speedlm

```
## S3 method for class 'speedlm'
logLik(object,...)
## S3 method for class 'speedlm'
AIC(object,...,k = 2)
```

# Arguments

object	an object of class 'speedlm'.
correlation	logical. Do you want to print the correlation matrix? By default it is false.
k	numeric, the penalty per parameter to be used; the default $k = 2$ is the classical AIC.
	further optional arguments

# Value

coefficients	the matrix of coefficients, standard errors, t-statistics and two-side p-values.
rdf	degrees of freedom of the fitted model. It is a component from object.
call	the component from object.
r.squared	R^2, the fraction of variance explained by the model.
adj.r.squared	the "adjusted" R^2 statistic, penalizing for higher p.
fstatistic	(for models including non-intercept terms) a 3-vector with the value of the F-statistic with its numerator and denominator degrees of freedom.
f.pvalue	p-value of the F-statistic.
RSS	Residual sum of squares.
var.res	estimated variance of residuals.
rank	the component from object.
correlation	(only if correlation is true) the correlations of the estimated parameters.
	the results from the functions logLik, AIC and vcov.

# Author(s)

Marco ENEA

### See Also

speedlm

# Examples

```
y <- rnorm(100,1.5,1)
x <- round(matrix(rnorm(200), 100, 2), digits = 3)
colnames(x) <- c("s1","s2")
da <- data.frame(y, x)
m <- speedlm(y ~ s1 + s2,da)
summary(m)</pre>
```

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