Package 'h.likelihood'

September 20, 2010	
Type Package	
Title Statistical Modeling and Inference via Hierarchical Likelihood	
Version 2010.9.20	
Date 2010-09-20	
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Description The package provides a top interface of hierarchical likelihood (h-likelihood) based models. It currently covers the estimation of hierarchical generalized linear models (HGLMs) and fraility models.	
License GPL	
LazyLoad yes	
Depends hglm, HGLMMM, lattice, Matrix, numDeriv	
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h.likelihood-package Statistical Modeling and Inference via Hierarchical Likelihood

Description

The package provides a top interface of hierarchical likelihood (h-likelihood) based models. It currently covers the estimation of hierarchical generalized linear models (HGLMs) and frailty models.

Details

h.likelihood-package

Package: h.likelihood Type: Package Version: 2010.9.20 Date: 2010-00-20 License: GPL LazyLoad: ves

Depends: hglm, HGLMMM

Author(s)

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References

Ha, I.D. and Lee, Y. (2003). Estimating frailty models via Poisson Hierarchical generalized linear models. Journal of Computational and Graphical Statistics, 12, 663-681.

Ha, I.D. and Lee, Y. (2005). Comparison of hierarchical likelihood versus orthodox best linear unbiased predictor approaches for frailty models. *Biometrika*, 92, 717-723.

Ha, I.D., Lee, Y. and Song, J.-K. (2001). Hierarchical likelihood approach for frailty models. Biometrika, 88, 233-243.

Lee, Y. and Nelder, J.A. (1996). Hierarchical generalized linear models (with discussion). Journal of the Royal Statistical Society. Series B (Methological) 58, 619-678.

Lee, Y. and Nelder, J.A. (2001). Hierarchical generalised linear models: A synthesis of generalised linear models, random-effect models and structured dispersions. Biometrika 88, 987-1006.

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Molas, M. and Lesaffre, E. (2010). Hierarchical Generalized Linear Models: the R Package HGLMMM . Submitted.

Noh, M. and Lee, Y. (2007). REML estimation for binary data in GLMMs. *Journal of Multivariate Analysis* 98, 896-915.

Ronnegard, L., Shen, X. and Alam, M. (2010). hglm: A Package for Fitting Hierarchical Generalized Linear Models. *The R Journal.* (to appear)

See Also

hglm-package, HGLMMM-package

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cqd

Chronic Granulomatous Disease (CGD) Infection Data

Description

The CGD data set in Fleming and Harrington (1991) consists of a placebo-controlled randomized trial of gamma interferon in chronic granulomatous disease.

Usage

```
data (cgd)
```

Format

A data frame with 203 observations on the following 16 variables.

OBS There were 203 observations.

HOSPITAL Hospitial ID: There were 13 hospotials where trials were conducted.

PATIENT Patient ID: There were 128 patients.

TIME The recurrent infection times of each pateient from the different hospitals.

DEL Censoring indicator(1 = uncensored, 0 = censored).

TRT Treatment Code(1 = gamma interferon, 0 = placebo).

INHERIT Pattern of inheritance (0 = autosomal recessive, 1 = X-linked).

AGE Age of each patient, years.

HEIGHT Height of each patient, in cm.

WEIGHT Weight of each patient, in kg.

STEROIDS Using corticosteroids at times of study centry(1 = Yes, 0 = No).

PROPYLAC Using prophylactic antibiotics at time of study entry(1 = Yes, 0 = No).

SEX Sex of each patient(0 = male, 1 = female).

H Hospital region(0 = U.S.A., 1 = Europe).

LONGI A longitudinal variable representing the accumulated time from the firts infection(in years).

ENUM Sequence number. For each patient, the infection records are in sequence number.

References

Fleming and Harrington (1991). Counting processes and survival analysis. Wiley: New York.

Examples

```
data (cqd)
```

Frailty.h

Frailty.h	Fitting Frailty Models using H-likelihood Approach

Description

Frailty, is used to fit frailty models using h-likelihood approach. The package fits Cox's proportional hazards models with random effects (or frailties), Here, for the frailty distribution lognomator or gamma is allowed. The h-likelihood obviates the need for marginalization over the frailty distribution, leading to a statistically efficient procedure for various random-effect models including frailty models.

Usage

```
Frailty.h(formulaMain, censor, DataMain, RandDist = "Normal", mord = 0, dord = 1, Maxiter = 200, convergence = 1e-7, contrasts = NULL)
```

Arguments

formulaMain	Specify the formula for the mean structure of the model. e.g. y $\sim x + (1 id)$, y: response, x: fixed covariate, id: random effect.
censor	One vector for censoring indicator ($1 = uncensored$, $0 = censored$).
DataMain	Dataframe for formulaMain.
RandDist	Distribution for random effect ("Normal" or "Gamma").
mord	Order for the mean model (0 or 1); $default = 0$.
dord	Order for the dispersion components (1 or 2); default = 1.
Maxiter	Maximum number of iterations; default = 1.
convergence	Specify the convergence criterion, the default is 1e-7.
contrasts	Caution as it is currently not fully developed.

Details

Frailty.h produces estimates of fixed effects and frailty parameters as well as their standard errors. Also, Frailty.h makes it possible to fit models where the frailty distribution is not necessarily normal and estimate variance components when frailty structure is shared or nested.

Author(s)

Il Do Ha

References

Ha, I.D. and Lee, Y. (2003). Estimating frailty models via Poisson Hierarchical generalized linear models. *Journal of Computational and Graphical Statistics*, 12, 663-681.

Ha, I.D. and Lee, Y. (2005). Comparison of hierarchical likelihood versus orthodox best linear unbiased predictor approaches for frailty models. *Biometrika*, 92, 717-723.

Ha, I.D., Lee, Y. and Song, J.-K. (2001). Hierarchical likelihood approach for frailty models. Biometrika. 88, 233-243.

Lee, Y., Nelder, J.A., and Pawitan, Y. (2006). Generalized Linear Models with Random Effects. Boca Raton: Chapman & Hall/CRC. HGLM 5

Examples

```
## Not run:
data (cod)
data(kidnev)
#### Frailty model
#### Analysis of Kidney data
kln11 <- Frailty.h(time ~ sex + age + (1|patient)-1, kidney$censor, kidney,
                   RandDist = "Normal", mord = 1, dord = 1)
kln12 <- Frailty.h(time ~ sex + age + (1|patient) - 1, kidney$censor, kidney,
                   RandDist = "Normal", mord = 1, dord = 2)
kq11 <- Frailty.h(time ~ sex + age + (1|patient) - 1, kidney$censor, kidney,
                  RandDist = "Gamma", mord = 1, dord = 1)
kg12 <- Frailty.h(time ~ sex + age + (1|patient) - 1, kidneyScensor, kidney,
                  RandDist = "Gamma", mord = 1, dord = 2)
#### Analysis of CGD data
cgd11 <- Frailty.h(TIME ~ TRT + (1|HOSPITAL) + (1|PATIENT) - 1, cgd$DEL, cgd,
                   RandDist = "Normal", mord = 1, dord = 1)
cgd12 <- Frailty.h(TIME ~ TRT + (1|HOSPITAL) + (1|PATIENT) - 1, cgd$DEL, cgd,
                   RandDist = "Normal", mord = 1, dord = 2)
## End(Not run)
```

HGLM

Fitting Hierarchical Generalized Linear Models

Description

This function fits hierarchical generalized linear models (HGLMs) using various approximation methods.

Usage

```
HGLM(y = NULL, X = NULL, Z = NULL, X.disp = NULL,
    family = gaussian(link = identity),
    random.family = gaussian(link = identity), method = "EQL",
    conv = le-04, maxit = 20, fixed = NULL, random = NULL,
    disp = NULL, link.disp = "log", disp.random = NULL,
    data = NULL, data.random = NULL, fix.disp = NULL,
    Offset = NULL, Weights = NULL, disp.start = 0, binomial.N = NULL,
    start.fixed = NULL, start.random = NULL, start.disp = NULL,
    start.disp.random = NULL, info = TRUE, debug = FALSE,
    contrasts = NULL
```

Arguments

```
y the dependent variable, only available when method = 'FQL'.

X a design matrix for the fixed effects, only available when method = 'EQL'.

Z and design matrix for the random effects, only available when method = 'EQL'.
```

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X.disp	a design matrix for the fixed effects in the dispersion part of the model, only available when method = 'EQL'.
family	a description of the error distribution and link function to be used in the mean part of the model. (See family for details of family functions.)
random.famil	Y
	a description of the error distribution and link function to be used in the variance part of the model.
method	estimation method, which can be 'EQL', 'HL01', or 'HL11', where 'EQL' can ONLY be used when ONLY ONE random effect term is specified. 'EQL' is the method of interconnected GLMs presented in Lee et al. (2006), and for 'HL01' and 'HL11', see Lee and Nelder (2001).
conv	convergence criterion, the default is 1e-4, for models with many random effects could be set less strict.
maxit	maximum number of iterations in the IWLS algorithm, only available when method = 'EQL'.
fixed	a formula specifying the fixed effects part of the model, and the format is Response ~ Fixed.Effect.1 + + Fixed.Effect.p.
random	a one-sided formula specifying the random effects part of the model, and the format is - (Random.Effect.1 Subject.1) + + (Random.Effect.q Subject.q).
disp	a one-sided formula specifying the fixed effects in the dispersion part of the model, and the format is ~ Effect.1 + + Effect.N.
link.disp	the link function for the dispersion part of the model, only available when method = 'EQL'.
disp.random	a list of one-sided formulae for the dispersion structure of each random effects, which has the format of list (one = ~ Effect.1.1 +, two = ~ Effect.2.1 +, three =,), only available when method = 'HLO1' or 'HLI1'.
data	the data frame to be used together with fixed and random.
data.random	a list of data.frames for disp.random, which has the format of list (one = data.Random.1, two = data.Random.2,), only available when method = 'H101' or '
Weights	prior weights to be specified in weighted regression, only available when $method = ' EQL'$.
fix.disp	a numeric value if the dispersion parameter of the mean model is known for example 1 for binomial and Poisson models.
Offset	an offset for the linear predictor of the mean model.
disp.start	(starting) values for the overdispersion structure - vector of length equal to the number of parameters in the overdispersion structure, only available when fix.disp = NULL and method = 'HLO1' or 'HL11'.
binomial.N	the number of trials for each observation for binomial models.
start.fixed	optional starting values for fixed effects in the mean structure (one vector of numeric values).
start.random	optional starting values for random effects in the mean structure (one vector of numeric values).
start.disp	optional starting values for parameters of dispersion components of the residuals (one vector of numeric values).

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start.disp.random

optional starting values for parameters of dispersion components of random ef-

fects (one vector of numeric values).

info a request to display of iteration information if TRUE, only available when method

= 'HL01' or 'HL11'.

debug a request to display of iteration mechanism progress in detail if TRUE, only

available when method = 'HL01' or 'HL11'.

contrasts see lm, caution as it is currently not fully developed, only available when method

= 'HL01' or 'HL11'.

Details

When method = 'EQL', all the model checking functions in the hglm-package are available on the object returned; Otherwise, all the model checking functions in the HGLMMM-package are available on the object returned.

Value

When method = 'EQL', an object of class hglm is returned, see hglm; Otherwise, an object of class HGLM is returned, see HGLMfit.

Note

The function provides a unified interface to the hglm-package developed by Moudud Alam, Lars Ronnegard and Xia Shen, and the HGLMMM-package developed by Marek Molas.

Author(s)

Xia Shen and Marek Molas

References

Lee, Y. and Nelder, J.A. (1996). Hierarchical generalized linear models (with discussion). Journal of the Royal Statistical Society. Series B (Methological) 58, 619-678.

Lee, Y. and Nelder, J.A. (2001). Hierarchical generalised linear models: A synthesis of generalised linear models, random-effect models and structured dispersions. *Biometrika* 88, 987-1006.

Lee, Y., Nelder, J.A., and Pawitan, Y. (2006). Generalized Linear Models with Random Effects. Boca Raton: Chapman & Hall/CRC.

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Ronnegard, L., Shen, X. and Alam, M. (2010). hglm: A Package for Fitting Hierarchical Generalized Linear Models. *The R Journal.* (to appear)

Molas, M. and Lesaffre, E. (2010). Hierarchical Generalized Linear Models: the R Package HGLMMM. Submitted.

See Also

hglm-package, HGLMMM-package, hglm, HGLMfit.

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Examples

```
data(semiconductor)
# ---- use 'EOL'
h.gamma.normal <- HGLM(fixed = y \sim x1 + x3 + x5 + x6,
                       random = ~ 1|Device,
                       family = Gamma(link = log),
                       disp = ~ x2 + x3, data = semiconductor)
summary (h.gamma.normal)
plot(h.gamma.normal, cex = .6, pch = 1,
     cex.axis = 1/.6, cex.lab = 1/.6,
     cex.main = 1/.6, mar = c(3, 4.5, 0, 1.5))
# ---- use 'HL(0,1)'
RSC <- data.frame(int = rep(1, 16))
h.gamma.normal <- HGLM(fixed = y \sim x1 + x3 + x5 + x6,
                       random = ~ 1|Device,
                       family = Gamma(link = log),
                       disp = \sim x2 + x3, data = semiconductor,
                       method = 'HL01', disp.start = c(0, 0, 0),
                       disp.random = list(one = ~ 1), data.random = list(RSC))
# ---- use 'HL(1,1)'
RSC <- data.frame(int = rep(1, 16))
h.gamma.normal <- HGLM(fixed = y \sim x1 + x3 + x5 + x6,
                       random = ~ 1|Device,
                       family = Gamma(link = log),
                       disp = ~ x2 + x3, data = semiconductor,
                       method = 'HL11', disp.start = c(0, 0, 0),
                       disp.random = list(one = ~ 1), data.random = list(RSC))
```

kidney

Kidney Infection Data

Description

The data presented by McGilchrist and Aisbett (1991) consist of times to the first and second recurrences of infection in 38 kidney patients using a portable dialysis machine.

Usage

```
data(kidney)
```

Format

A data frame with 76 observations on the following 9 variables.

time Survival time: Time to infection since insertion of the catheter

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```
censor Censoring indicator(1 = uncensored, 0 = censored). obs There were 76 observations. patient Patient ID: There were 38 patients with two recurrences. age Age of each patient. sex Sex of each patient (1 = male, 2 = female). \\ dy0 GN type of disease(1 = Yes, 0 = No). \\ dy1 AN type of disease(1 = Yes, 0 = No). \\ dy2 PKD type of disease(1 = Yes, 0 = No).
```

References

McGilchrist and Aisbett(1991). Regression with frailty in survival analysis. *Biometrics*, 47, 461-466.

Examples

data(kidney)

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