Package 'MBRM'

September 8, 2025

Type Package			
Fitle Mixed Regression Models with Generalized Log-Gamma Random Effects			
Version 0.1.0			
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Description ``This package fits binary mixed regression models with random effects based on the generalized log-gamma distribution. Estimation is performed by maximizing the log-likelihood using numerical optimization."			
License GPL-3			
Encoding UTF-8			
LazyData true			
RoxygenNote 7.3.2			
LinkingTo Rcpp			
mports Rcpp, stats, Formula, tibble, dplyr, ggplot2			
Depends R (>= 3.5)			
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Arthritis1

Arthritis1 Dataset

Description

This dataset contains binary response data from a longitudinal study on rheumatoid arthritis. The data include repeated measurements on 38 individuals, each with 5 time points.

Usage

```
Arthritis1
```

Format

A data frame with 190 rows and 6 variables:

```
Ind Subject identifier (integer from 1 to 38)
y Binary response variable (0 or 1)
Sex Sex indicator (0 = Female, 1 = Male)
Age Age indicator (0 = <=55 years 1 = > 55 uears)
Treatment Treatment group (0 = Placebo, 1 = Auranofin)
```

Time Time indicator (0 = baseline, 1 = follow-up)

Details

The original data was collected to evaluate the effect of a treatment over time on arthritis severity. Covariates include demographic and clinical variables, and the response is binary (presence/absence of symptoms).

Source

Derived from the original dataset included in:

Arthritis.txt an internal clinical dataset used in Bernoulli-GLG modeling study.

References

Fitzmaurice, G. M. and Lipsitz, S. (1995). A model for binary time series data with serial odds ratio patterns. *Journal of the Royal Statistical Society: Series B*, 44, 51–61.

Examples

```
## Not run:
data(Arthritis1)
head(Arthritis1)

# Fit the Bernoulli-LGG model
fit <- MRMfit(y ~ Sex + Age + Treatment + Time, data = Arthritis1)
summary(fit)

## End(Not run)</pre>
```

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envelope.MRM

Compute simulation envelopes for MRM model

Description

This function computes the envelopes simulation of the randomized quantile residuals for objects of class MRM.

Usage

```
envelope.MRM(object, R = 100, ...)
```

Arguments

object An object of class MRM, typically returned from MRMfit.

R Integer. Number of replications to simulate the envelopes (default is 100).

Additional arguments passed to MRMfit when envelope = TRUE.

Value

A QQ-plot with envelope.

Examples

```
data(Arthritis1)
fit <- MRMfit(y ~ Sex + Age + Treatment + Time, data = Arthritis1)
envelope.MRM(fit)</pre>
```

MRMfit

Fit Mixed Regression Model with Log-Gamma Random Effects

Description

This function fits a mixed regression model for binary outcomes with random effects following a generalized log-gamma distribution. The estimation is performed by maximizing a custom log-likelihood using numerical optimization via optim.

Usage

```
MRMfit(formula, data, hessian = TRUE, method = "BFGS", ...)
```

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Arguments

formula A symbolic description of the model to be fitted, e.g., $y \sim x1 + x2$. The response variable must be binary (0/1). data A data frame containing the variables in the model. The data must include an Ind column indicating cluster or subject IDs for the random effects. hessian Logical. Should a numerically differentiated Hessian matrix be returned? method Optimization method to be used in optim. One of "Nelder-Mead", "BFGS", "CG", "L-BFGS-B", "SANN", or "Brent". Default is "BFGS".

Additional arguments passed to optim, such as control, lower, or upper (when . . .

supported by the chosen method).

Value

An object of class "MRM" containing:

The matched function call. call

formula The model formula.

coefficients Estimated fixed effects coefficients.

scale Estimated scale parameter for the random effects distribution.

loglik Maximized log-likelihood value. Number of clusters or subjects. n

Vector with the number of observations per cluster. m

Estimated standard errors of the parameters. ер iter Number of iterations used by the optimizer.

method Optimization method used. The original data frame used. data

Examples

```
## Not run:
# Simulated data
data1 <- rMRM(n = 50, m = rep(3, 50),
              theta = c(0.8, 1, -1),
              X = data.frame(x1 = rnorm(150), x2 = rnorm(150)))
# Fit using BFGS (default)
fit1 <- MRMfit(y \sim x1 + x2, data = data1)
summary(fit1)
# Fit using L-BFGS-B with bounds
fit2 <- MRMfit(y \sim x1 + x2, data = data1,
               method = "L-BFGS-B",
               lower = c(1e-5, rep(-Inf, 3)),
               upper = rep(Inf, 4),
               control = list(factr = 1e7))
summary(fit2)
## End(Not run)
```

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residuals.MRM	Compute the randomized quantile residuals for MRM model	

Description

This function computes the randomized quantile residuals for objects of class MRM.

Usage

```
## S3 method for class 'MRM'
residuals(object, ...)
```

Arguments

object An object of class MRM, typically returned from MRMfit.
... Additional arguments passed to MRMfit.

Value

A numeric vector of randomized quantile residuals.

Examples

```
data(Arthritis1)
fit <- MRMfit(y ~ Sex + Age + Treatment + Time, data = Arthritis1)
summary(residuals(fit))</pre>
```

rMRM

Simulate Data from a Mixed Regression Model with GLG Random Effects

Description

This function simulates clustered binary response data from a mixed regression model with random effects. The model allows for different link functions and random effect distributions.

Usage

```
rMRM(n, m, theta, X)
```

Arguments

Χ

n	Integer. Number of clusters or subjects.
m	Integer vector of length n. Each element indicates the number of observations per cluster.
theta	Numeric vector. The first element is the scale or dispersion parameter for the random effects, and the remaining values are the fixed effects coefficients, including the intercept.

A data frame or matrix of covariates with n rows (one per cluster). Should not

include the intercept.

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Value

A tibble containing the simulated dataset with the following columns:

```
Ind Cluster or subject ID (integer from 1 to n).
```

y Binary response variable (0 or 1).

x1, x2, ... Covariates as defined in X, repeated according to cluster size.

The output also has an attribute "proportions" indicating the proportions of 0's and 1's in y.

Examples

```
## Not run:
set.seed(123)
n <- 500
m <- rep(3,n)
theta <- c(0.5,1,-2,1)
set.seed(123)
X <- cbind(runif(n),rnorm(n))
set.seed(456)
data1 <- rMRM(n,m,theta,X)
head(data1)
## End(Not run)</pre>
```

toenail

toenail Dataset

Description

A data frame with 1908 observations on the following 4 variables.

Usage

toenail

Format

A data frame with 1908 rows and 4 variables:

```
Ind Subject identifier (integer from 1 to 38)
```

y Binary response is the severity of infection, 0 (not severe) and 1(severe).

```
treatment Treatment group (0 = Treatment A, 1 = Treatment B)
```

month a numeric vector giving the time of the visit (not exactly monthly intervals hence not round numbers)

visit a numeric vector giving the number of the visit

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Details

The Toenail data discussed in Molenberghs and Verbeke (2010) were obtained from a randomized, double-blind, parallel-group, multicenter study for the comparison of two oral treatments (coded as A and B) for Toenail Dermatophyte Onychomycosis (TDO). The present study aimed to compare the efficacy of 12 weeks of continuous therapy with Treatment A or B. In total, 294 patients, distributed over 36 centers, were randomized. Subjects were followed during 12 weeks (3 months) of treatment and followed further up to 48 weeks (12 months). Measurements were taken at baseline every month during treatment and every 3 months afterward, resulting in a maximum of 7 measurements per subject.

References

Molenberghs G, Verbeke G (2010). Models for Discrete Longitudinal data. Springer, New York.

Examples

```
## Not run:
data(toenail)
head(toenail)

# Fit the Bernoulli-LGG model
fit <- MRMfit(y ~ treatment + month + treatment:month, data = toenail)
summary(fit)

## End(Not run)</pre>
```

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