Package 'mbest'

April 24, 2025

Version 0.6.1

Title Moment-Based Estimation for Hierarchical Models

Description Fast moment-based hierarchical model fitting. Implements methods from the papers
``Fast Moment-Based Estimation for Hierarchical Models,'' by Perry (2017) and
``Fitting a Deeply Nested Hierarchical Model to a Large Book Review Dataset Using a Moment-Based Estimator,'' by Zhang, Schmaus, and Perry (2018).

Depends nlme (>= 3.1-124), R (>= 3.3.0)

Imports abind, bigmemory, foreach, reformulas, logging

Suggests testthat, lme4

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URL https://github.com/patperry/r-mbest

Encoding UTF-8

NeedsCompilation yes

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Repository CRAN

Date/Publication 2025-04-24 11:50:09 UTC

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`mbest-package` *Moment-Based Estimation for Hierarchical Models*

Description

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Details

Package:	mbest
Version:	0.6.1
Title:	Moment-Based Estimation for Hierarchical Models
Authors@R:	c(person(c("Patrick", "O."), "Perry", role = "aut"), person("Timothy", "Sweetser", role = "ctb"), person("Kyle"
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LazyData:	Yes
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URL:	https://github.com/patperry/r-mbest
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Basic usage is to call mhglm.

References

P. O. Perry (2017) "Fast moment-based estimation for hierarchical models."

N. Zhang, K. Schmaus, and P. O. Perry (2018) "Fitting deeply-nested hierarchical models to a large book review dataset using moment-based estimators."

See Also

mhglm, fixef.mhglm, ranef.mhglm, VarCorr.mhglm, predict.mhglm.

---

effects *Mixed Effects*

Description

Get the fixed effects, random effect variances, and empirical Bayes random effect estimates.

Usage

```
## S3 method for class 'mhglm'
fixef(object, ...)

## S3 method for class 'mhglm'
ranef(object, condVar = FALSE, ...)

## S3 method for class 'mhglm'
vcov(object, ...)

## S3 method for class 'mhglm'
VarCorr(x, sigma = 1, ...)

## S3 method for class 'mhglm_ml'
fixef(object, ...)

## S3 method for class 'mhglm_ml'
ranef(object, condVar = FALSE, ...)

## S3 method for class 'mhglm_ml'
vcov(object, ...)

## S3 method for class 'mhglm_ml'
VarCorr(x, sigma = 1, ...)
```

Arguments

object, x	an mhglm object.
sigma	a factor by which to scale the random effect variance-covariance matrix.
condVar	a logical indicating whether conditional covariance matrices for the random effects should be returned.
...	further arguments passed to or from other methods.

Details

fixef returnes the fixed effects, while vcov returns the variance-covariance matrix of the fixed effect estimates.

VarCorr returns the random effect covariance matrix. ranef returns the empirical Bayes random effect estimates.

These functions behave like their counterparts in the **nlme** package.

See Also

fixef, ranef, VarCorr, from package **nlme**.

---

firthglm.fit *Fitting Generalized Linear Models with Firth's Bias Reduction*

Description

A drop-in replacement for glm.fit which uses Firth's bias-reduced estimates instead of maximum likelihood.

Usage

```
firthglm.fit(x, y, weights = rep(1, nobs),
             start = NULL, etastart = NULL, mustart = NULL,
             offset = rep(0, nobs), family = gaussian(),
             control = list(...), intercept = TRUE, singular.ok = TRUE, ...)

firthglm.control(epsilon = 1e-8, maxit = 25, qr.tol = 1e-7,
                 improve.tol = 1e-4, curvature.tol = 0.9,
                 linesearch.method = "linesearch",
                 linesearch.maxit = 20, trace = FALSE)
```

Arguments

x, y, weights, start, etastart, mustart, offset, family, control, intercept, singular.ok, ...	arguments that have the same functions as for glm.fit.
qr.tol	tolerance parameter for determining the rank of x.

`epsilon`, `maxit` convergence parameters for the quasi-Newton method.

`linesearch.method`
line search methods (one of "linesearch", "backtrack", or "blindsearch")

`improve.tol`, `curvature.tol`, `linesearch.maxit`
tolerance parameters for the linesearch procedure.

`trace` logical indicating if output should be produced for each iteration.

Details

Firth's modified score function gives rise to estimates with smaller biases than their maximum likelihood counterparts. Unlike the maximum likelihood estimates, if the design matrix is of full rank, then the Firth bias-reduced estimate is finite.

By default, the fitting procedure uses a quasi-Newton optimization method, with a More-Thuente linesearch.

Value

`firthglm.fit` returns an object having the same components that a call to `glm.fit` would produce.

Note

Currently, only families with canonical link functions are supported.

Author(s)

Patrick O. Perry

References

Firth, D. (1993) Bias reduction of maximum likelihood estimates. *Biometrika* **80**, 27-–38.

More, J. J. and Thuente, D. J. (1994) Line search algorithms with guaranteed sufficient decrease. *ACM Transactions on Mathematical Software* **20** 286–307.

See Also

`logistf` (package **logistf**) and `brglm` (package **brglm**) for alternative implementations of Firth's bias-reduced estimators.

Examples

```
## Dobson (1990) Page 93: Randomized Controlled Trial :
counts <- c(18,17,15,20,10,20,25,13,12)
outcome <- gl(3,1,9)
treatment <- gl(3,3)

## Use firthglm.fit instead of glm.fit:
glm.D93 <- glm(counts ~ outcome + treatment, family = poisson(),
               method="firthglm.fit")
summary(glm.D93)
```

mhglm *Fitting Moment Hierarchical Generalized Linear Models*

Description

mhglm is used to fit a moment hierarchical generalized linear model of one level. mhglm_ml is used to fit a moment hierarchical generalized linear model of arbitrary number of levels (including one level).

Usage

```
mhglm(formula, family = gaussian, data, weights, subset,
      na.action, start = NULL, etastart, mustart, offset,
      control = list(), model = TRUE, method = "mhglm.fit",
      x = FALSE, z = FALSE, y = TRUE, group = TRUE,
      contrasts = NULL)

mhglm.fit(x, z, y, group, weights = rep(1, nobs),
          start = NULL, etastart = NULL, mustart = NULL,
          offset = rep(0, nobs), family = gaussian(),
          control = list(), intercept = TRUE, dispersion = NULL)

mhglm_ml(formula, family = gaussian, data, weights, subset,
         na.action, start = NULL, etastart, mustart, offset,
         control = list(), model = TRUE, method = "mhglm_ml.fit",
         x = FALSE, z = FALSE, y = TRUE, group = TRUE,
         contrasts = NULL)

mhglm_ml.fit(x, z, y, group, weights = rep(1, nobs),
             start = NULL, etastart = NULL, mustart = NULL,
             offset = rep(0, nobs), family = gaussian(),
             control = list(), intercept = TRUE)
```

Arguments

formula, family, data, weights, subset, na.action, start, etastart, mustart, offset, model, contrasts, intercept
: These arguments are analogous to the similarly-named arguments for the glm and glm.fit functions.

control
: a list of parameters for controlling the fitting process. For mhglm.fit this is passed to mhglm.control.

method
: the method to be used in fitting the model. The default method "mhglm.fit" uses moment-based estimates; the alternative "model.frame" returns the model frame and does no fitting.

`x`, `z`, `y`, `group` — For `mhglm` and `mhglm_ml`: logical values indicating whether the response vector, model matrices, and grouping factor used in the fitting process should be returned as components of the returned value.

For `mhglm.fit`: `x` is a fixed effect design matrix of dimension `n * p`, `z` is a random effect design matrix of dimension `n * q`, `y` is a vector of observations of length `n`, and `group` is a grouping factor vector of length `n`.

For `mhglm_ml.fit`: `x` is a fixed effect design matrix of dimension `n * p`, `z` is a list of L elements, with L the depth of nested hierarchies, each element of `z` is a random effect design matrix of dimension `n * q_i`, with `q_i` the feature dimension on tree depth `i`, `y` is a vector of observations of length `n`, and `group` is a list of L elements (same L as `z`), each element of `group` is a grouping factor vector of length `n`.

`dispersion` — If `NULL`, will estimate from data; otherwise use this argument as dispersion parameter.

Details

These functions are analogues of `glm` and `glm.fit`, meant to be used for fitting hierarchical generalized linear models. A typical predictor has the form `response ~ terms + (reterms | group)` where `response` is the (numeric) response vector, `terms` is a series of terms which specifies a linear predictor for `response`, `reterms` is a series of terms with random coefficients (effects), and `group` is a grouping factor; observations with the same grouping factor share the same random effects.

`mhglm` and `mhglm.fit` only allow one random effect term, along with a single level of hierarchy. `mghlm_ml` and `mhglm_ml.fit` allow multiple random effect terms so long as levels of random effects are hierarchically nested. If the random effect design matrices are the same for each level, a predictor has the form `response ~ terms + (reterms | g1/.../gQ)`. If the random effects design matrices differ from level to level, colons are used to delineate the nesting structure; for example, `response ~ fe + (re1 | g1) + (re2 | g2:g1) + (re3 | g3:g2:g1)`.

`mhglm` allows `||` in the formula `response ~ terms + (reterms || group)` to indicate that random effects are independent, that is the random effects covariance matrix has non-zero value only on its diagonal. `mhglm_ml` currently does not support `||`, to indicate indpenedent random effects, set `control=list(diagcov = TRUE)`.

Value

`mhglm` returns an object of class inheriting from `"mhglm"`. `mhglm_ml` returns an object of class inheriting from `"mhglm_ml"`.

The function `summary` can be used to obtain or print a summary of the results.

The generic accessor functions `fixef`, `ranef`, `VarCorr`, `sigma`, `fitted.values` and `residuals` can be used to extract various useful features of the value returned by `mhglm` or `mhglm_ml`.

Note

If the moment-based random effect covariance is not positive-semidefinite, then a warning will be issued, and a projection of the estimate to the positive-semidefinite cone will be used instead.

References

P. O. Perry (2017) "Fast moment-based estimation for hierarchical models."

N. Zhang, K. Schmaus, and P. O. Perry (2018) "Fitting deeply-nested hierarchical models to a large book review dataset using moment-based estimators."

See Also

`terms.mhglm`, `model.matrix.mhglm`, and `predict.mhglm` for `mhglm` methods, and the generic functions `fitted.values`, `residuals`, `summary`, `vcov`, and `weights`.

Generic functions `fixef`, `ranef`, `VarCorr`, and `sigma` for features related to mixed effect models.

`glmer` (package **lme4**) for fitting generalized linear mixed models with likelihood-based estimates.

Examples

```
if (require("lme4")) {
## The following examples are adapted from lme4:
(fm1 <- mhglm(Reaction ~ Days + (Days | Subject), gaussian, sleepstudy))

(fm2 <- mhglm(Reaction ~ Days + (Days || Subject), gaussian, sleepstudy))

(gm <- mhglm(cbind(incidence, size - incidence) ~ period + (1 | herd),
             data = cbpp, family = binomial))

## The following examples are for multilevel models
g_data <- mhglm_sim(n = 30, m_per_level = c(10, 5, 2), sd_intercept = c(1, 1, 1),
                    sd_slope = c(1, 1, 1), family = "gaussian", seed = 12345)

(m1 <- mhglm_ml(y ~ 1 + x + (1 + x | g1/g2/g3), gaussian, g_data))
# or equivalent
(m2 <- mhglm_ml(y ~ 1 + x + (1 + x | g1) + (1 + x | g2:g1) + (1 + x | g3:g2:g1),
                gaussian, g_data))
}
```

`mhglm.control` *Auxiliary for Controlling Moment Heirarchical GLM Fitting*

Description

Auxiliary function for `mhglm` fitting. Typically only used internally by `mhglm.fit`, but may be used to construct a control argument to either function.

Usage

```
mhglm.control(standardize = TRUE, steps = 1, parallel = FALSE, diagcov = FALSE,
      fit.method = "firthglm.fit", fixef.rank.warn = FALSE, cov.rank.warn = FALSE,
        cov.psd.warn = TRUE, fit.control = list(...), ...)
```

```
mhglm_ml.control(standardize = FALSE, steps = 1, parallel = FALSE, diagcov = FALSE,
          fit.method = "firthglm.fit", fixef.rank.warn = FALSE, cov.rank.warn = FALSE,
               cov.psd.warn = FALSE, fit.control = list(...), ...)
```

Arguments

`standardize`	logitcal indicating if predictors should be standardized before moment-based fitted
`steps`	number of refinement steps
`parallel`	fit the group-specific estimates in parallel rather than sequentially
`diagcov`	estimate random effect covairance matrix with diagonal approximation
`fit.method`	method for obtaining group-specific effect estimates
`fixef.rank.warn`	if TRUE, print warnings when fixef is unidentifiable
`cov.rank.warn`	if TRUE, print warnings when covariance matrix is unidentifiable
`cov.psd.warn`	if TRUE, print warnings when moment based estimates of covariance matrix is not positive semi-definite
`fit.control`	control parameters for `fit.method`
`...`	arguments to be used to form the `fit.control` argument if it is not supplied directly.

Details

Setting `standardize = TRUE` ensures that the procedure is equivariant, and generally leads to better estimation performance. Right now `standardize = TRUE` is not allowed for `mhglm_ml`.

The `steps` argument gives the number of refinement steps for the moment based parameters. In each step, the previous fixed effect and random effect covariance matrix estimates are used to weight the subpopulation-specific effect estimates. In principle, higher values of `steps` could lead to more accurate estimates, but in simulations, the differences are negligible.

Value

A list with components named as the arguments.

See Also

`mhglm.fit`, the fitting procedure used by `mhglm`.

`firthglm.fit`, the default subpopulation-specific fitting method.

Examples

```
if(requireNamespace("lme4")) { # for cbpp data

 data("cbpp", package = "lme4")
# The default fitting method uses Firth's bias-corrected estimates
(gm.firth <- mhglm(cbind(incidence, size - incidence) ~ period + (1 | herd),
```

```
                  data = cbpp, family = binomial,
                  control=mhglm.control(fit.method="firthglm.fit")))

# Using maximum likelihood estimates is less reliable
(gm.ml <- mhglm(cbind(incidence, size - incidence) ~ period + (1 | herd),
                data = cbpp, family = binomial,
                control=mhglm.control(fit.method="glm.fit")))
}
```

---

mhglm_sim *Simulate response patterns*

Description

Simulate response patterns for generalized linear models of `gaussian` or `binomial` families, with both an intercept and slope covariate. Used primarily for testing purposes.

Usage

```
mhglm_sim(n, m_per_level, sd_intercept, sd_slope,
          family = c("gaussian", "binomial"), seed)
```

Arguments

`n`	an integer scalar, the number of observations at the lowest grouping level.
`m_per_level`	an integer vector, the number of grouping levels nested under the level above.
`sd_intercept`	a numeric vector, the standard deviations of the intercept random effects.
`sd_slope`	a numeric vector, the standard deviations of the slope random effects.
`family`	a character scalar, either `"gaussian"` or `"binomial"`.
`seed`	a single value, interpreted as an integer, or `NULL` as in `set.seed`.

Details

returns a data.frame with design matrix, response, and group levels.

Examples

```
mhglm_sim(n = 2, m_per_level = c(3, 3), sd_intercept = c(1, 2),
        sd_slope = c(2, 1), family = "gaussian", seed = 123)

mhglm_sim(n = 2, m_per_level = c(3, 3), sd_intercept = c(1, 2),
        sd_slope = c(2, 1), family = "binomial", seed = 123)
```

model.matrix.mhglm *Terms and Model Matrix*

Description

Get the terms or model matrix from an mhglm object.

Usage

```
  ## S3 method for class 'mhglm'
model.matrix(object, type = c("fixed", "random"), ...)
  ## S3 method for class 'mhglm_ml'
model.matrix(object, type = c("fixed", "random"), ...)

  ## S3 method for class 'mhglm'
terms(x, type = c("fixed", "random"), ...)
  ## S3 method for class 'mhglm_ml'
terms(x, type = c("fixed", "random"), ...)
```

Arguments

object, x	an mhglm object.
type	which terms to get (for the fixed or for the random effects).
...	further arguments passed to or from other methods.

See Also

model.matrix, terms

predict *Prediction*

Description

predict gives empirical Bayes predictions of the response, while sigma gives the dispersion parameter.

Usage

```
## S3 method for class 'mhglm'
predict(object, newdata = NULL, type = c("link", "response"),
      se.fit = FALSE, na.action = na.pass, ...)

## S3 method for class 'mhglm'
sigma(object, ...)
```

Arguments

`object`	an `mhglm` object
`newdata, type, se.fit, na.action`	these arguments behave as in `predict.glm`. See Details, below.
`...`	further arguments passed to or from other methods.

Details

The `predict` function gives empirical Bayes posterior mean estimates of response values. If `se.fit = TRUE`, then the conditional variances of the random effects are used along with the fixed effect variance-covariance matrix to estimate the standard errors.

The `sigma` function gives the square root of the dispersion parameter or the model; for linear models, this is the error standard deviation.

See Also

`predict`, `sigma`

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