Package ‘glmmTMB’

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Title Generalized Linear Mixed Models using Template Model Builder

Version 1.0.1

Description Fit linear and generalized linear mixed models with various extensions, including zero-inflation. The models are fitted using maximum likelihood estimation via ‘TMB’ (Template Model Builder). Random effects are assumed to be Gaussian on the scale of the linear predictor and are integrated out using the Laplace approximation. Gradients are calculated using automatic differentiation.

License AGPL-3

Depends R (>= 3.2.0)

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LinkingTo TMB, RcppEigen

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Methods have been written that allow glmmTMB objects to be used with several downstream packages that enable different forms of inference. In particular,

- `car::Anova` constructs type-II and type-III Anova tables for the fixed effect parameters of the conditional model (this might work with the fixed effects of the zero-inflation or dispersion models, but has not been tested)
- the `effects` package computes graphical tabular effect displays (again, for the fixed effects of the conditional component)
- the `emmeans` package computes estimated marginal means (aka least-squares means) for the fixed effects of the conditional component

**Usage**

```r
Anova.glmmTMB(
  mod,
  type = c("II", "III", 2, 3),
  test.statistic = c("Chisq", "F"),
  component = "cond",
  vcov. = vcov(mod)[[component]],
  singular.ok,
  ...
)
```

```r
Effect.glmmTMB(focal.predictors, mod, ...)
```

```r
recover_data.glmmTMB(object, ...)
```

```r
emm_basis.glmmTMB(object, trms, xlev, grid, component = "cond", ...)
```

**Arguments**

- `mod` a glmmTMB model
- `type` type of test, "II", "III", 2, or 3. Roman numerals are equivalent to the corresponding Arabic numerals. See `Anova` for details.
- `test.statistic` unused: only valid choice is "Chisq" (i.e., Wald chi-squared test)
- `component` which component of the model to compute emmeans for (conditional ("cond"), zero-inflation ("zi"), or dispersion ("disp"))
- `vcov.` variance-covariance matrix (usually extracted automatically)
- `singular.ok` OK to do ANOVA with singular models (unused)?
- `...` Additional parameters that may be supported by the method.
confint.glmmTMB

Calculate confidence intervals

Description
Calculate confidence intervals

Usage

## S3 method for class 'glmmTMB'
confint(
  object, 
  parm = NULL, 
  level = 0.95, 
  method = c("wald", "Wald", "profile", "uniroot"), 
  component = c("all", "cond", "zi", "other"), 
  estimate = TRUE, 
  parallel = c("no", "multicore", "snow"), 
  ncpus = getOption("profile.ncpus", 1L), 
  cl = NULL, 
  full = FALSE, 
  ... 
)

Arguments

object glmmTMB fitted object.

parm which parameters to profile, specified

• by index (position) [after component selection for confint, if any]
• by name (matching the row/column names of vcov(object, full=TRUE))
Parameter indexing by number may give unusual results when some parameters have been fixed using the map argument: please report surprises to the package maintainers.

level  
Confidence level.

method  
'wald', 'profile', or 'uniroot': see Details function)

component  
Which of the three components 'cond', 'zi' or 'other' to select. Default is to select 'all'.

estimate  
(logical) add a third column with estimate ?

parallel  
method (if any) for parallel computation

ncpus  
number of CPUs/cores to use for parallel computation

c1  
cluster to use for parallel computation

full  
CIs for all parameters (including dispersion) ?

...  
arguments may be passed to profile.merMod or tmbroot

Details

Available methods are

"wald" These intervals are based on the standard errors calculated for parameters on the scale of their internal parameterization depending on the family. Derived quantities such as standard deviation parameters and dispersion parameters are back-transformed. It follows that confidence intervals for these derived quantities are typically asymmetric.

"profile" This method computes a likelihood profile for the specified parameter(s) using profile.glmmTMB; fits a spline function to each half of the profile; and inverts the function to find the specified confidence interval.

"uniroot" This method uses the uniroot function to find critical values of one-dimensional profile functions for each specified parameter.

At present, "wald" returns confidence intervals for variance parameters on the standard deviation/correlation scale, while "profile" and "uniroot" report them on the underlying ("theta") scale: for each random effect, the first set of parameter values are standard deviations on the log scale, while remaining parameters represent correlations on the scaled Cholesky scale (see the

Examples

data(sleepstudy, package="lme4")
model <- glmmTMB(Reaction ~ Days + (1|Subject), sleepstudy)
model2 <- glmmTMB(Reaction ~ Days + (1|Subject), sleepstudy, 
dispformula= ~I(Days>8))
confint(model)  ## Wald/delta-method CIs
confint(model.parm="theta_.")  ## Wald/delta-method CIs
confint(model.parm=1,method="profile")
Description

Extended version of the epil dataset of the MASS package. The three transformed variables Visit, Base, and Age used by Booth et al. (2003) have been added to epil.

Usage

epil2

Format

A data frame with 236 observations on the following 12 variables:

- `y` an integer vector.
- `trt` a factor with levels "placebo" and "progabide".
- `base` an integer vector.
- `age` an integer vector.
- `V4` an integer vector.
- `subject` an integer vector.
- `period` an integer vector.
- `lbase` a numeric vector.
- `lage` a numeric vector.
- `Visit` \((\text{rep}(1:4,59) - 2.5) / 5\).
- `Base` \(\log(\text{base}/4)\).
- `Age` \(\log(\text{age})\).

References


Examples

```r
epil2$subject <- factor(epil2$subject)
op <- options(digits=3)
(fm <- glmmTMB(y ~ Base*trt + Age + Visit + (Visit|subject),
data=epil2, family=nbinom2))
meths <- methods(class = class(fm))
if((Rv <- getRversion()) > "3.1.3") {
  (funs <- attr(meths, "info")[, "generic"])
  for(F in funs[is.na(match(funs, "getME"))]) {
```
findReTrmClasses

```r
findReTrmClasses

```
Value

an object of class `fixef.glmmTMB` comprising a list of components (cond, zi, disp), each containing a (possibly zero-length) numeric vector of coefficients

Examples

data(sleepstudy, package = "lme4")
fm1 <- glmmTMB(Reaction ~ Days, sleepstudy)
(f1 <- fixef(fm1))
f1$cond
## show full coefficients, including dispersion parameter
unlist(f1)
print.default(f1)

---

formatVC

Format the 'VarCorr' Matrix of Random Effects

Description

"format()" the 'VarCorr' matrix of the random effects – for print(ing) and show(ing)

Usage

```r
formatVC(
  varcor,
  digits = max(3, getOption("digits") - 2),
  comp = "Std.Dev.",
  formatter = format,
  useScale = attr(varcor, "useSc"),
  ...
)
```

Arguments

- `varcor` a `VarCorr` (-like) matrix with attributes.
- `digits` the number of significant digits.
- `comp` character vector of length one or two indicating which columns out of "Variance" and "Std.Dev." should be shown in the formatted output.
- `formatter` the function to be used for formatting the standard deviations and or variances (but not the correlations which (currently) are always formatted as "0.000"
- `useScale` whether to report a scale parameter (e.g. residual standard deviation)
- `...` optional arguments for `formatter(*)` in addition to the first (numeric vector) and digits.

Value

a character matrix of formatted VarCorr entries from `varc`. 

**formula.glmmTMB**

Extract the formula of a glmmTMB object

**Description**

Extract the formula of a glmmTMB object

**Usage**

```r
## S3 method for class 'glmmTMB'
formula(x, fixed.only = FALSE, component = c("cond", "zi", "disp"), ...)
```

**Arguments**

- `x`: a glmmTMB object
- `fixed.only`: (logical) drop random effects, returning only the fixed-effect component of the formula?
- `component`: formula for which component of the model to return (conditional, zero-inflation, or dispersion)
- `...`: unused, for generic consistency

**getCapabilities**

List model options that glmmTMB knows about

**Description**

List model options that glmmTMB knows about

**Usage**

```r
getCapabilities(what = "all", check = FALSE)
```

**Arguments**

- `what`: (character) which type of model structure to report on ("all","family","link","covstruct")
- `check`: (logical) do brute-force checking to test whether families are really implemented (only available for what="family")

**Value**

if check==FALSE, returns a vector of the names (or a list of name vectors) of allowable entries; if check==TRUE, returns a logical vector of working families

**Note**

these are all the options that are defined internally; they have not necessarily all been implemented (FIXME!)
**getME.glmmTMB**  
*Extract or Get Generalize Components from a Fitted Mixed Effects Model*

### Description

Extract or Get Generalize Components from a Fitted Mixed Effects Model

### Usage

```r
## S3 method for class 'glmmTMB'
getME(object, name = c("X", "Xzi", "Z", "Zzi", "Xd", "theta", "beta"), ...)
```

### Arguments

- **object**: a fitted glmmTMB object
- **name**: of the component to be retrieved
- **...**: ignored, for method compatibility

### See Also

- `getME` Get generic and re-export:

**getReStruc**  
*Calculate random effect structure Calculates number of random effects, number of parameters, block size and number of blocks. Mostly for internal use.*

### Description

Calculate random effect structure Calculates number of random effects, number of parameters, block size and number of blocks. Mostly for internal use.

### Usage

```r
getReStruc(reTrms, ss = NULL)
```

### Arguments

- **reTrms**: random-effects terms list
- **ss**: a character string indicating a valid covariance structure. Must be one of `names(glmmTMB::.valid_covstruct)`; default is to use an unstructured variance-covariance matrix ("us") for all blocks).
Value

- a list
  - blockNumTheta: number of variance covariance parameters per term
  - blockSize: size (dimension) of one block
  - blockReps: number of times the blocks are repeated (levels)
  - covCode: structure code

Examples

data(sleepstudy, package="lme4")
rt <- lme4::lFormula(Reaction~Days+(1|Subject)+(0+Days|Subject), sleepstudy)$reTrms
rt2 <- lme4::lFormula(Reaction~Days+(Days|Subject), sleepstudy)$reTrms
getReStruc(rt)

Description

Create X and random effect terms from formula

Usage

g getXReTrms(formula, mf, fr, ranOK = TRUE, type = "", contrasts)

Arguments

- formula: current formula, containing both fixed & random effects
- mf: matched call
- fr: full model frame
- ranOK: random effects allowed here?
- type: label for model type
- contrasts: a list of contrasts (see ?glmmTMB)

Value

- a list composed of
  - X: design matrix for fixed effects
  - Z: design matrix for random effects
  - reTrms: output from mkReTrms from lme4
  - offset: offset vector, or vector of zeros if offset not specified
get_cor translate vector of correlation parameters to correlation values

Description
translate vector of correlation parameters to correlation values

Usage
get_cor(theta)

Arguments
theta vector of internal correlation parameters

Details
This function follows the definition at http://kaskr.github.io/adcomp/classUNSTRUCTURED_CORR_t.html: if \( L \) is the lower-triangular matrix with 1 on the diagonal and the correlation parameters in the lower triangle, then the correlation matrix is defined as \( \Sigma = D^{-1/2} L L^\top D^{-1/2} \), where \( D = \text{diag}(L L^\top) \). For a single correlation parameter \( \theta_0 \), this works out to \( \rho = \theta_0/\sqrt{1+\theta_0^2} \).

The function returns the elements of the lower triangle of the correlation matrix, in column-major order.

Value
a vector of correlation values

Examples
```r
th0 <- 0.5
stopifnot(all.equal(get_cor(th0), th0/sqrt(1+th0^2)))
get_cor(c(0.5,0.2,0.5))
```

---

Fit Models with TMB

Description
Fit a generalized linear mixed model (GLMM) using Template Model Builder (TMB).
Usage

```r
glmmTMB(
  formula,
  data = NULL,
  family = gaussian(),
  ziformula = ~0,
  dispformula = ~1,
  weights = NULL,
  offset = NULL,
  contrasts = NULL,
  na.action = na.fail,
  se = TRUE,
  verbose = FALSE,
  doFit = TRUE,
  control = glmmTMBControl(),
  REML = FALSE,
  start = NULL,
  map = NULL
)
```

Arguments

- **formula**: combined fixed and random effects formula, following lme4 syntax.
- **data**: optional data frame containing model variables.
- **family**: a family function, a character string naming a family function, or the result of a call to a family function (variance/link function) information. See `family` for a generic discussion of families or `family_glmmTMB` for details of `glmmTMB`-specific families.
- **ziformula**: a one-sided (i.e., no response variable) formula for zero-inflation combining fixed and random effects: the default ~0 specifies no zero-inflation. Specifying ~. sets the zero-inflation formula identical to the right-hand side of formula (i.e., the conditional effects formula); terms can also be added or subtracted. **When using ~. as the zero-inflation formula in models where the conditional effects formula contains an offset term, the offset term will automatically be dropped.** The zero-inflation model uses a logit link.
- **dispformula**: a one-sided formula for dispersion containing only fixed effects: the default ~1 specifies the standard dispersion given any family. The argument is ignored for families that do not have a dispersion parameter. For an explanation of the dispersion parameter for each family, see `sigma`. The dispersion model uses a log link. In Gaussian mixed models, dispformula=~0 fixes the residual variance to be 0 (actually a small non-zero value: at present it is set to sqrt(.Machine$double.eps)), forcing variance into the random effects.
- **weights**: weights, as in `glm`. Not automatically scaled to have sum 1.
- **offset**: offset for conditional model (only).
- **contrasts**: an optional list, e.g., list(fac1="contr.sum"). See the contrasts.arg of `model.matrix.default`.
na.action how to handle missing values, see \texttt{na.action} and \texttt{model.frame}. From \texttt{lm}:
“The default is set by the \texttt{na.action} setting of \texttt{options}, and is \texttt{na.fail} if that is unset. The ‘factory-fresh’ default is \texttt{na.omit}.”

\texttt{se} whether to return standard errors.

\texttt{verbose} whether progress indication should be printed to the console.

\texttt{doFit} whether to fit the full model, or (if FALSE) return the preprocessed data and parameter objects, without fitting the model.

\texttt{control} control parameters, see \texttt{glmmTMBControl}.

\texttt{REML} whether to use REML estimation rather than maximum likelihood.

\texttt{start} starting values, expressed as a list with possible components \texttt{beta}, \texttt{betazi}, \texttt{betad} (fixed-effect parameters for conditional, zero-inflation, dispersion models); \texttt{b}, \texttt{bzi} (conditional modes for conditional and zero-inflation models); \texttt{theta}, \texttt{thetazi} (random-effect parameters, on the standard deviation/Cholesky scale, for conditional and z-i models); \texttt{thetaf} (extra family parameters, e.g., shape for Tweedie models).

\texttt{map} a list specifying which parameter values should be fixed to a constant value rather than estimated. \texttt{map} should be a named list containing factors corresponding to a subset of the internal parameter names (see \texttt{start} parameter). Distinct factor values are fitted as separate parameter values, NA values are held fixed: e.g., \texttt{map=list(beta=factor(c(1,2,3,NA))}) would fit the first three fixed-effect parameters of the conditional model and fix the fourth parameter to its starting value. In general, users will probably want to use \texttt{start} to specify non-default starting values for fixed parameters. See \texttt{MakeADFun} for more details.

Details

Binomial models with more than one trial (i.e., not binary/Bernoulli) can either be specified in the form \texttt{prob \sim ...}, \texttt{weights = N}, or in the more typical two-column matrix \texttt{cbind(successes,failures) \sim ...} form.

Behavior of \texttt{REML=TRUE} for Gaussian responses matches \texttt{lme4::lmer}. It may also be useful in some cases with non-Gaussian responses (Millar 2011). Simulations should be done first to verify.

Because the \texttt{df.residual} method for \texttt{glmmTMB} currently counts the dispersion parameter, one would need to multiply by \texttt{sqrt(nobs(fit) / (1+df.residual(fit)))} when comparing with \texttt{lm}.

By default, vector-valued random effects are fitted with unstructured (general positive definite) variance-covariance matrices. Structured variance-covariance matrices can be specified in the form \texttt{struc(terms\mid group)}, where \texttt{struc} is one of

- \texttt{diag} (diagonal, heterogeneous variance)
- \texttt{ar1} (autoregressive order-1, homogeneous variance)
- \texttt{cs} (compound symmetric, heterogeneous variance)
- \texttt{ou} (* Ornstein-Uhlenbeck, homogeneous variance)
- \texttt{exp} (* exponential autocorrelation)
- \texttt{gau} (* Gaussian autocorrelation)
- \texttt{mat} (* Matérn process correlation)
• toep (* Toeplitz)

Structures marked with * are experimental/untested.

For backward compatibility, the family argument can also be specified as a list comprising the name of the distribution and the link function (e.g. list(family="binomial",link="logit")). However, this alternative is now deprecated; it produces a warning and will be removed at some point in the future. Furthermore, certain capabilities such as Pearson residuals or predictions on the data scale will only be possible if components such as variance and linkfun are present, see family.

Note

For more information about the glmmTMB package, see Brooks et al. (2017) and the vignette(package="glmmTMB") collection. For the underlying TMB package that performs the model estimation, see Kristensen et al. (2016).

References


Examples

(m1 <- glmmTMB(count ~ mined + (1|site),
    zi=~mined,
    family=poisson, data=Salamanders))
summary(m1)

## Zero-inflated negative binomial model
(m2 <- glmmTMB(count ~ spp + mined + (1|site),
    zi=~spp + mined,
    family=nbinom2, data=Salamanders))

## Hurdle Poisson model
(m3 <- glmmTMB(count ~ spp + mined + (1|site),
    zi=~spp + mined,
    family=truncated_poisson, data=Salamanders))

## Binomial model
data(cbpp, package="lme4")
(bovine <- glmmTMB(cbind(incidence, size-incidence) ~ period + (1|herd),
    family=binomial, data=cbpp))

## Dispersion model
sim1 <- function(nfac=40, nt=100, facsd=0.1, tsd=0.15, mu=0, residsd=1)
```r
{  
  dat <- expand.grid(fac=factor(letters[1:nfac]), t=1:nt)  
  n <- nrow(dat)  
  dat$REfac <- rnorm(nfac, sd=facsd)[dat$fac]  
  dat$REt <- rnorm(nt, sd=tsd)[dat$t]  
  dat$x <- rnorm(n, mean=mu, sd=resisd) + dat$REfac + dat$REt  
  dat
}
set.seed(101)
  d1 <- sim1(mu=100, resisd=10)
  d2 <- sim1(mu=200, resisd=5)
  d1$sd <- "ten"
  d2$sd <- "five"
  dat <- rbind(d1, d2)
  m0 <- glmmTMB(x ~ sd + (1|t), dispformula=~sd, data=dat)
  fixef(m0)$disp
  c(log(5^2), log(10^2)-log(5^2)) # expected dispersion model coefficients
}
```
### glmmTMBControl

- **optimizer**
  Function to use in model fitting. See Details for required properties of this function.

- **profile**
  Logical; Experimental option to improve speed and robustness when a model has many fixed effects.

- **collect**
  Logical; Experimental option to improve speed by recognizing duplicated observations.

- **parallel**
  Numeric; Setting number of OpenMP threads to evaluate the negative log-likelihood in parallel.

### Details

The general non-linear optimizer `nlminb` is used by `glmmTMB` for parameter estimation. It may sometimes be necessary to tweak some tolerances in order to make a model converge. For instance, the warning 'iteration limit reached without convergence' may be fixed by increasing the number of iterations using something like

```r
glmmTMBControl(optCtrl=list(iter.max=1e3,eval.max=1e3)).
```

The argument `profile` allows `glmmTMB` to use some special properties of the optimization problem in order to speed up estimation in cases with many fixed effects. Enable this option using

```r
glmmTMBControl(profile=TRUE).
```

Control parameters may depend on the model specification, because each control component is evaluated inside `TMBStruc`, the output of `mkTMBStruc`. To specify that `profile` should be enabled for more than 5 fixed effects one can use

```r
glmmTMBControl(profile=quote(length(parameters$beta)>=5)).
```

The optimizer argument can be any optimization (minimizing) function, provided that:

- the first three arguments, in order, are the starting values, objective function, and gradient function;
- it also takes a control argument;
- it returns a list with elements (at least) convergence (0 if convergence is successful) and message.

### Examples

```r
## fit with default (nlminb) and alternative (optim/BFGS) optimizer
m1 <- glmmTMB(count~ mined, family=poisson, data=Salamanders)
m1B <- update(m1, control=glmmTMBControl(optimizer=optim,
                        optArgs=list(method="BFGS")))
## estimates are *nearly* identical:
all.equal(fixef(m1), fixef(m1B))
```
**isLMM.glmmTMB**

**support methods for parametric bootstrapping**

**Description**

see `refit` and `isLMM` for details

**Usage**

```r
## S3 method for class 'glmmTMB'
isLMM(object)

## S3 method for class 'glmmTMB'
refit(object, newresp, ...)
```

**Arguments**

- `object` a fitted `glmmTMB` object
- `newresp` a new response vector
- `...` additional arguments (for generic consistency; ignored)

**Details**

These methods are still somewhat experimental (check your results carefully!), but they should allow parametric bootstrapping. They work by copying and replacing the original response column in the data frame passed to `glmmTMB`, so they will only work properly if (1) the data frame is still available in the environment and (2) the response variable is specified as a single symbol (e.g. proportion or a two-column matrix constructed on the fly with `cbind()`). Untested with binomial models where the response is specified as a factor.

**Examples**

```r
if (requireNamespace("lme4")) {
  ## Not run:
  fm1 <- glmmTMB(count~mined+(1|spp),
                 ziformula=~mined,
                 data=Salamanders,
                 family=nbinom1)
  b1 <- lme4::bootMer(fm1, FUN=function(x) fixef(x)$zi, nsim=20, .progress="txt")
  if (requireNamespace("boot")) {
    boot.ci(b1, type="perc")
  }

  ## End(Not run)
}
```
Description

Family functions for glmmTMB

Usage

nbinom2(link = "log")
nbinom1(link = "log")
compois(link = "log")
truncated_compois(link = "log")
genpois(link = "log")
truncated_genpois(link = "log")
truncated_poisson(link = "log")
truncated_nbinom2(link = "log")
truncated_nbinom1(link = "log")
beta_family(link = "logit")
betabinomial(link = "logit")
tweedie(link = "log")
ziGamma(link = "inverse")

Arguments

link (character) link function for the conditional mean ("log", "logit", "probit", "inverse", "cloglog", "identity", or "sqrt")

Details

If specified, the dispersion model uses a log link. Denoting the variance as $V$, the dispersion parameter as $\phi = \exp(\eta)$ (where $\eta$ is the linear predictor from the dispersion model), and the predicted mean as $\mu$:

**gaussian** (from base R): constant $V = \phi$
\textbf{Gamma} (from base R) phi is the shape parameter. \( V = \mu \phi \)

\textbf{ziGamma} a modified version of Gamma that skips checks for zero values, allowing it to be used to fit hurdle-Gamma models

\textbf{nbinom2} Negative binomial distribution: quadratic parameterization (Hardin & Hilbe 2007). \( V = \mu(1 + \mu/\phi) = \mu + \mu^2/\phi \).

\textbf{nbinom1} Negative binomial distribution: linear parameterization (Hardin & Hilbe 2007). \( V = \mu(1 + \phi) \)

\textbf{compois} Conway-Maxwell Poisson distribution: parameterized with the exact mean (Huang 2017), which differs from the parameterization used in the \texttt{COMPoissonReg} package (Sellers & Shmueli 2010, Sellers & Lotze 2015). \( V = \mu \phi \).

\textbf{genpois} Generalized Poisson distribution (Consul & Famoye 1992). \( V = \mu \exp(\eta) \). (Note that Consul & Famoye (1992) define \( \phi \) differently.)

\textbf{beta} Beta distribution: parameterization of Ferrari and Cribari-Neto (2004) and the \texttt{betareg} package (Cribari-Neto and Zeileis 2010); \( V = \mu(1 - \mu)\phi \)

\textbf{betabinomial} Beta-binomial distribution: parameterized according to Morris (1997). \( V = \mu(1 - \mu)(n(\phi + n)/(\phi + 1)) \)

\textbf{tweedie} Tweedie distribution: \( V = \phi \mu^p \). The power parameter is restricted to the interval \( 1 < p < 2 \)

\textbf{Value}

returns a list with (at least) components

\textbf{family} length-1 character vector giving the family name
\textbf{link} length-1 character vector specifying the link function
\textbf{variance} a function of either 1 (mean) or 2 (mean and dispersion parameter) arguments giving a value proportional to the predicted variance (scaled by \texttt{sigma(.)})

\textbf{References}

numFactor

Factor with numeric interpretable levels.

Description
Create a factor with numeric interpretable factor levels.

Usage
numFactor(x, ...)

parseNumLevels(levels)

Arguments
x Vector, matrix or data.frame that constitute the coordinates.
...
Additional vectors, matrices or data.frames that constitute the coordinates.
levels Character vector to parse into numeric values.

Details
Some glmmTMB covariance structures require extra information, such as temporal or spatial coordinates. numFactor allows to associate such extra information as part of a factor via the factor levels. The original numeric coordinates are recoverable without loss of precision using the function parseNumLevels. Factor levels are sorted coordinate wise from left to right: first coordinate is fastest running.

Value
Factor with specialized coding of levels.

Examples
## 1D example
numFactor(sample(1:5,20,TRUE))
## 2D example
coords <- cbind( sample(1:5,20,TRUE), sample(1:5,20,TRUE) )
(f <- numFactor(coords))
parseNumLevels(levels(f)) ## Sorted
## Used as part of a model.matrix
model.matrix(~ f )
## parseNumLevels( colnames(model.matrix(~ f )) )
## Error: 'Failed to parse numeric levels: (Intercept)'
parseNumLevels( colnames(model.matrix(~ f-1 )) )
Description

Begging by owl nestlings

Usage

data(Owls)

Format

The Owls data set is a data frame with 599 observations on the following variables:

- **Nest** a factor describing individual nest locations
- **FoodTreatment** (factor) food treatment: Deprived or Satiated
- **SexParent** (factor) sex of provisioning parent: Female or Male
- **ArrivalTime** a numeric vector
- **SiblingNegotiation** a numeric vector
- **BroodSize** brood size
- **NegPerChick** number of negotiations per chick

Note

Access to data kindly provided by Alain Zuur

Source


References


Examples

data(Owls, package = "glmmTMB")
require("lattice")
bwplot(reorder(Nest,NegPerChick) ~ NegPerChick | FoodTreatment:SexParent, data=Owls)
dotplot(reorder(Nest,NegPerChick) ~ NegPerChick| FoodTreatment:SexParent, data=Owls)
## Not run:
## Fit negative binomial model with "constant" Zero Inflation:
owls_nb1 <- glmmTMB(SiblingNegotiation ~ FoodTreatment*SexParent +
                      (1|Nest)+offset(log(BroodSize)),
                      family = nbinom1(), zi = ~1, data=Owls)
owls_nb1 bs <- update(owls_nb1,
                      . ~ . - offset(log(BroodSize)) + log(BroodSize))
fixef(owls_nb1 bs)
## End(Not run)

---

**predict.glmmTMB**

**prediction**

### Description
prediction

### Usage
## S3 method for class 'glmmTMB'
predict(
  object,
  newdata = NULL,
  newparams = NULL,
  se.fit = FALSE,
  re.form = NULL,
  allow.new.levels = FALSE,
  type = c("link", "response", "conditional", "zprob", "zlink", "disp"),
  zitype = NULL,
  na.action = na.pass,
  debug = FALSE,
  ...
)

### Arguments
- **object**: a glmmTMB object
- **newdata**: new data for prediction
- **newparams**: new parameters for prediction
- **se.fit**: return the standard errors of the predicted values?
- **re.form**: NULL to specify individual-level predictions; ~0 or NA to specify population-level predictions (i.e., setting all random effects to zero)
- **allow.new.levels**: allow previously unobserved levels in random-effects variables? see details.
- **type**: Denoting \( \mu \) as the mean of the conditional distribution and \( p \) as the zero-inflation probability, the possible choices are:
"link" conditional mean on the scale of the link function, or equivalently the linear predictor of the conditional model

"response" expected value; this is \( \mu \ast (1 - p) \) for zero-inflated models and \( \mu \) otherwise

"conditional" mean of the conditional response; \( \mu \) for all models (i.e., synonymous with "response" in the absence of zero-inflation

"zprob" the probability of a structural zero (gives an error for non-zero-inflated models)

"zlink" predicted zero-inflation probability on the scale of the logit link function

"disp" dispersion parameter however it is defined for that particular family as described in \( \sigma_{\text{glmmTMB}} \)

zitype deprecated: formerly used to specify type of zero-inflation probability. Now synonymous with type

na.action how to handle missing values in newdata (see na.action); the default (na.pass) is to predict NA

deprecated: formerly used to specify type of zero-inflation probability. Now synonymous with type

... unused - for method compatibility

**Details**

- To compute population-level predictions for a given grouping variable (i.e., setting all random effects for that grouping variable to zero), set the grouping variable values to NA. Finer-scale control of conditioning (e.g. allowing variation among groups in intercepts but not slopes when predicting from a random-slopes model) is not currently possible.

- Prediction of new random effect levels is possible as long as the model specification (fixed effects and parameters) is kept constant. However, to ensure intentional usage, a warning is triggered if allow.new.levels=FALSE (the default).

- Prediction using "data-dependent bases" (variables whose scaling or transformation depends on the original data, e.g. poly, ns, or poly) should work properly; however, users are advised to check results extra-carefully when using such variables. Models with different versions of the same data-dependent basis type in different components (e.g. formula= y ~ poly(x,3), dispformula= ~poly(x,2)) will probably not produce correct predictions.

**Examples**

data(sleepstudy, package="lme4")
g0 <- glmmTMB(Reaction~Days+(Days|Subject), sleepstudy)
predict(g0, sleepstudy)
## Predict new Subject
nd <- sleepstudy[1,]
nd$Subject <- "new"
predict(g0, newdata=nd, allow.new.levels=TRUE)
## population-level prediction
nd_pop <- data.frame(Days=unique(sleepstudy$Days), Subject=NA)
predict(g0, newdata=nd_pop)
`print.VarCorr.glmmTMB`  
Printing The Variance and Correlation Parameters of a `glmmTMB`

Description

Printing The Variance and Correlation Parameters of a `glmmTMB`

Usage

```r
## S3 method for class 'VarCorr.glmmTMB'
print(
  x,
  digits = max(3, getOption("digits") - 2),
  comp = "Std.Dev.",
  formatter = format,
  ...
)
```

Arguments

- `x`: a result of `VarCorr(<glmmTMB>).`
- `digits`: number of significant digits to use.
- `comp`: a string specifying the component to format and print.
- `formatter`: a function.
- `...`: optional further arguments, passed the next `print` method.

`profile.glmmTMB`  
Compute likelihood profiles for a fitted model

Description

Compute likelihood profiles for a fitted model

Usage

```r
## S3 method for class 'glmmTMB'
profile(
  fitted,
  parm = NULL,
  level_max = 0.99,
  npts = 8,
  stepfac = 1/4,
  stderr = NULL,
  trace = FALSE,
```
parallel = c("no", "multicore", "snow"),
ncpus = getOption("profile.ncpus", 1L),
cl = NULL,
...)

## S3 method for class 'profile.glmmTMB'
confint(object, parm = NULL, level = 0.95, ...)

Arguments

fitted
  a fitted glmmTMB object

parm
  which parameters to profile, specified
  • by index (position)
  • by name (matching the row/column names of vcov(object, full=TRUE))
  • as "theta_" (random-effects variance-covariance parameters) or "beta_"
    (conditional and zero-inflation parameters)

level_max
  maximum confidence interval target for profile

npts
  target number of points in (each half of) the profile (approximate)

stepfac
  initial step factor (fraction of estimated standard deviation)

stderr
  standard errors to use as a scaling factor when picking step sizes to compute the
  profile; by default (if stderr is NULL, or NA for a particular element), uses the
  estimated (Wald) standard errors of the parameters

trace
  print tracing information? If trace=FALSE or 0, no tracing; if trace=1, print
  names of parameters currently being profiled; if trace>1, turn on tracing for the
  underlying tmbprofile function

parallel
  method (if any) for parallel computation

ncpus
  number of CPUs/cores to use for parallel computation

cl
  cluster to use for parallel computation

object
  a fitted profile (profile.glmmTMB) object

level
  confidence level

Details

Fits natural splines separately to the points from each half of the profile for each specified parameter
(i.e., values above and below the MLE), then finds the inverse functions to estimate the endpoints
of the confidence interval

Value

An object of class profile.glmmTMB, which is also a data frame, with columns .par (parameter
being profiled), .focal (value of focal parameter), value (negative log-likelihood).
Examples

```r
## Not run:
m1 <- glmmTMB(count~ mined + (1|site),
    zi=mined, family=poisson, data=Salamanders)
salamander_prof1 <- profile(m1, parallel="multicore",
    ncpus=2, trace=1)
## testing
salamander_prof1 <- profile(m1, trace=1, parm=1)
salamander_prof1M <- profile(m1, trace=1, parm=1, npts = 4)
salamander_prof2 <- profile(m1, parm="theta_")
```

## End(Not run)
salamander_prof1 <- readRDS(system.file("example_files","salamander_prof1.rds",package="glmmTMB"))
if (require("ggplot2")) {
  ggplot(salamander_prof1,aes(.focal,sqrt(value))) +
  geom_point() + geom_line() +
  facet_wrap(~.par,scale="free_x") +
  geom_hline(yintercept=1.96,linetype=2)
}
salamander_prof1 <- readRDS(system.file("example_files","salamander_prof1.rds",package="glmmTMB"))
confint(salamander_prof1)
confint(salamander_prof1, level=0.99)
```

---

### ranef.glmmTMB

**Extract Random Effects**

**Description**

Extract random effects from a fitted `glmmTMB` model, both for the conditional model and zero inflation.

**Usage**

```r
## S3 method for class 'glmmTMB'
ranef(object, condVar = TRUE, ...)

## S3 method for class 'ranef.glmmTMB'
as.data.frame(x, ..., stringsAsFactors = default.stringsAsFactors())

## S3 method for class 'glmmTMB'
coef(object, condVar = FALSE, ...)
```

**Arguments**

- `object` a `glmmTMB` model.
- `condVar` whether to include conditional variances in result.
- `...` some methods for this generic function require additional arguments.
x a ranef.glmmTMB object (i.e., the result of running ranef on a fitted glmmTMB model)
stringsAsFactors see data.frame

Value

• For ranef, an object of class ranef.glmmTMB with two components:
  cond a list of data frames, containing random effects for the conditional model.
  zi a list of data frames, containing random effects for the zero inflation.
If condVar=TRUE, the individual list elements within the cond and zi components (corresponding to individual random effects terms) will have associated condVar attributes giving the conditional variances of the random effects values. These are in the form of three-dimensional arrays: see ranef.merMod for details. The only difference between the packages is that the attributes are called ‘postVar’ in lme4, vs. ‘condVar’ in glmmTMB.

• For coef.glmmTMB: a similar list, but containing the overall coefficient value for each level, i.e., the sum of the fixed effect estimate and the random effect value for that level. Conditional variances are not yet available as an option for coef.glmmTMB.

• For as.data.frame: a data frame with components
  component part of the model to which the random effects apply (conditional or zero-inflation)
  grpvar grouping variable
  term random-effects term (e.g., intercept or slope)
  grp group, or level of the grouping variable
  condval value of the conditional mode
  condsd conditional standard deviation

Note

When a model has no zero inflation, the ranef and coef print methods simplify the structure shown, by default. To show the full list structure, use print(ranef(model), simplify=FALSE) or the analogous code for coef. In all cases, the full list structure is used to access the data frames, see example.

See Also

fixef.glmmTMB.

Examples

if (requireNamespace("lme4")) {
data(sleepstudy, package="lme4")
model <- glmmTMB(Reaction ~ Days + (1|Subject), sleepstudy)
rr <- ranef(model)
print(rr, simplify=FALSE)
## extract Subject conditional modes for conditional model
rr$cond$Subject
as.data.frame(rr)
}
residuals.glmmTMB

Compute residuals for a glmmTMB object

Description

Compute residuals for a glmmTMB object

Usage

```r
## S3 method for class 'glmmTMB'
residuals(object, type = c("response", "pearson"), ...)
```

Arguments

- **object**: a “glmmTMB” object
- **type**: (character) residual type
- **...**: ignored, for method compatibility

Salamanders

Repeated counts of salamanders in streams

Description

A data set containing counts of salamanders with site covariates and sampling covariates. Each of 23 sites was sampled 4 times. When using this data set, please cite Price et al. (2016) as well as the Dryad data package (Price et al. 2015).

Usage

```r
data(Salamanders)
```

Format

A data frame with 644 observations on the following 10 variables:

- **site**: name of a location where repeated samples were taken
- **mined**: factor indicating whether the site was affected by mountain top removal coal mining
- **cover**: amount of cover objects in the stream (scaled)
- **sample**: repeated sample
- **DOP**: Days since precipitation (scaled)
- **Wtemp**: water temperature (scaled)
- **DOY**: day of year (scaled)
- **spp**: abbreviated species name, possibly also life stage
- **count**: number of salamanders observed
References

Examples
```r
require("glmmTMB")
data(Salamanders)

zmpm3 = glmmTMB(count~spp * mined + (1|site), zi=~spp * mined, Salamanders, family="poisson")
```

sigma.glmmTMB

*Extract residual standard deviation or dispersion parameter*

Description
For Gaussian models, `sigma()` returns the value of the residual standard deviation; for other families, it returns the dispersion parameter, *however it is defined for that particular family*. See details for each family below.

Usage
```r
## S3 method for class 'glmmTMB'
sigma(object, ...)
```

Arguments
- `object` a “glmmTMB” fitted object
- `...` (ignored; for method compatibility)

Details
The value returned varies by family:

- **gaussian** returns the *maximum likelihood* estimate of the standard deviation (i.e., smaller than the results of `sigma(lm(...))` by a factor of (n-1)/n)
- **nbinom1** returns an overdispersion parameter (usually denoted $\alpha$ as in Hardin and Hilbe (2007)): such that the variance equals $\mu(1 + \alpha)$.
- **nbinom2** returns an overdispersion parameter (usually denoted $\theta$ or $k$); in contrast to most other families, larger $\theta$ corresponds to a *lower* variance which is $\mu(1 + \mu/\theta)$. 
**Gamma** Internally, `glmmTMB` fits Gamma responses by fitting a mean and a shape parameter; `sigma` is estimated as \((1/\sqrt{\text{shape}})\), which will typically be close (but not identical to) that estimated by `stats::sigma.default`, which uses \(\sqrt{\text{deviance}/\text{df.residual}}\).

**beta** returns the value of \(\phi\), where the conditional variance is \(\mu(1 - \mu)/(1 + \phi)\) (i.e., increasing \(\phi\) decreases the variance.) This parameterization follows Ferrari and Cribari-Neto (2004) (and the `betareg` package):

**betabinomial** This family uses the same parameterization (governing the Beta distribution that underlies the binomial probabilities) as `beta`.

**genpois** returns the index of dispersion \(\phi^2\), where the variance is \(\mu\phi^2\) (Consul & Famoye 1992).

**compois** returns the value of \(1/\nu\), When \(\nu = 1\), compois is equivalent to the Poisson distribution. There is no closed form equation for the variance, but it is approximately undersidpersed when \(1/\nu < 1\) and approximately oversidpersed when \(1/\nu > 1\). In this implementation, \(\mu\) is exactly the mean (Huang 2017), which differs from the `COMPoissonReg` package (Sellers & Lotze 2015).

**tweedie** returns the value of \(\phi\), where the variance is \(\phi\mu^p\). The value of \(p\) can be extracted using the internal function `glmmTMB::.tweedie_power`.

The most commonly used GLM families (binomial, poisson) have fixed dispersion parameters which are internally ignored.

**References**


---

### simulate.glmmTMB

**Simulate from a glmmTMB fitted model**

#### Description

Simulate from a glmmTMB fitted model

#### Usage

```r
## S3 method for class 'glmmTMB'
simulate(object, nsim = 1, seed = NULL, ...)
```
Arguments

- **object**: glmmTMB fitted model
- **nsim**: number of response lists to simulate. Defaults to 1.
- **seed**: random number seed
- **...**: extra arguments

Details

Random effects are also simulated from their estimated distribution. Currently, it is not possible to condition on estimated random effects.

Value

returns a list of vectors. The list has length `nsim`. Each simulated vector of observations is the same size as the vector of response variables in the original data set. In the binomial family case each simulation is a two-column matrix with success/failure.

terms.glmmTMB

Methods for extracting developer-level information from glmmTMB models

Description

Methods for extracting developer-level information from glmmTMB models

Usage

```r
## S3 method for class 'glmmTMB'
terms(x, component = "cond", part = "fixed", ...)
```  

```r
## S3 method for class 'glmmTMB'
model.matrix(object, component = "cond", part = "fixed", ...)
```  

Arguments

- **x**: a fitted glmmTMB object
- **component**: model component ("cond", "zi", or "disp"; not all models contain all components)
- **part**: whether to return results for the fixed or random effect part of the model (at present only part="fixed" is implemented for most methods)
- **...**: additional arguments (ignored or passed to `model.frame`)
- **object**: a fitted glmmTMB object
vcov.glmmTMB

`vcov.glmmTMB` is an S3 method for class `glmmTMB` that calculates the variance-covariance matrix for a fitted glmmTMB model.

### Description

Calculate Variance-Covariance Matrix for a Fitted glmmTMB model.

### Usage

```
## S3 method for class 'glmmTMB'
vcov(object, full = FALSE, ...)
```

### Arguments

- **object**: a `glmmTMB` fit
- **full**: return a full variance-covariance matrix?
- **...**: ignored, for method compatibility

### Value

By default (`full==FALSE`), a list of separate variance-covariance matrices for each model component (conditional, zero-inflation, dispersion). If `full==TRUE`, a single square variance-covariance matrix for all top-level model parameters (conditional, dispersion, and variance-covariance parameters).

---

weights.glmmTMB

`weights.glmmTMB` extracts weights from a glmmTMB object.

### Description

Extract weights from a glmmTMB object.

### Usage

```
## S3 method for class 'glmmTMB'
weights(object, type = "prior", ...)
```

### Arguments

- **object**: a fitted glmmTMB object
- **type**: weights type
- **...**: additional arguments (not used; for methods compatibility)
Details

At present only explicitly specified prior weights (i.e., weights specified in the weights argument) can be extracted from a fitted model.

- Unlike other GLM-type models such as glm or glmer, weights() does not currently return the total number of trials when binomial responses are specified as a two-column matrix.
- Since glmmTMB does not fit models via iteratively weighted least squares, working weights (see weights.glm) are unavailable.
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