# Package 'mclogit'

July 1, 2020

Type Package

**Title** Multinomial Logit Models, with or without Random Effects or Overdispersion

**Version** 0.8.5.1 **Date** 2020-06-27

**Author** Martin Elff

Maintainer Martin Elff <mclogit@elff.eu>

**Description** Provides estimators for multinomial logit models in their

conditional logit and baseline logit variants, with or without random effects, with or without overdispersion.

Random effects models are estimated using the PQL technique (based on a Laplace approximation)

or the MQL technique (based on a Solomon-Cox approximation). Estimates should be treated with caution if the group sizes are small.

License GPL-2

**Depends** stats, Matrix

Imports memisc, methods

Suggests MASS, nnet

LazyLoad Yes

URL http://www.elff.eu/software/mclogit/,http://github.com/melff/mclogit/

BugReports http://github.com/melff/mclogit/issues

RoxygenNote 7.1.0

NeedsCompilation no

Repository CRAN

**Date/Publication** 2020-07-01 16:40:07 UTC

2 dispersion

## **R** topics documented:

Index		17
	Transport	15
	simulate.mclogit	
	mclogit.fit	13
	mclogit.control	12
	mclogit	9
	mblogit	6
	getSummary-methods	4
	electors	3
	dispersion	2

## Description

dispersion

The function dispersion() extracts the dispersion parameter from a multinomial logit model or computes a dispersion parameter estimate based on a given method. This dispersion parameter can be attached to a model using update(). It can also given as an argument to summary().

Overdispersion in Multinomial Logit Models

### Usage

```
dispersion(object,method, ...)
## S3 method for class 'mclogit'
dispersion(object,method=NULL, ...)
```

### Arguments

object	an object that inherits class "mclogit". When passed to dispersion(), it should be the result of a call of mclogit() of mblogit(), without random effects.
method	a character string, either "Afroz", "Fletcher", "Pearson", or "Deviance", that specifies the estimator of the dispersion; or NULL, in which case the default estimator, "Afroz" is used. The estimators are discussed in Afroz et al. (2019).
	other arguments, ignored or passed to other methods.

### References

Afroz, Farzana, Matt Parry, and David Fletcher. (2019). "Estimating Overdispersion in Sparse Multinomial Data." *Biometrics*, early view. https://doi.org/10.1111/biom.13194.

electors 3

### **Examples**

```
library(MASS) # For 'housing' data
# Note that with a factor response and frequency weighted data,
# Overdispersion will be overestimated:
house.mblogit <- mblogit(Sat ~ Infl + Type + Cont, weights = Freq,</pre>
                         data = housing)
dispersion(house.mblogit,method="Afroz")
dispersion(house.mblogit,method="Deviance")
summary(house.mblogit)
phi.Afroz <- dispersion(house.mblogit,method="Afroz")</pre>
summary(house.mblogit, dispersion=phi.Afroz)
summary(update(house.mblogit, dispersion="Afroz"))
# In order to be able to estimate overdispersion accurately,
# data like the above (which usually comes from applying
# 'as.data.frame' to a contingency table) the model has to be
# fitted with the optional argument 'from.table=TRUE':
house.mblogit.corrected <- mblogit(Sat ~ Infl + Type + Cont, weights = Freq,</pre>
                                    data = housing, from.table=TRUE,
                                    dispersion="Afroz")
# Now the estimated dispersion parameter is no longer larger than 20,
# but just bit over 1.0.
summary(house.mblogit.corrected)
```

electors

Class, Party Position, and Electoral Choice

### **Description**

This is an artificial data set on electoral choice as influenced by class and party positions.

### Usage

```
data(electors)
```

### Format

A data frame containing the following variables:

```
class class position of voters
```

party party that runs for election

Freq frequency by which each party list is chosen by members of each class

time time variable, runs from zero to one

```
econ.left economic-policy "leftness" of each partywelfare emphasis of welfare expansion of each partyauth position on authoritarian issues
```

### **Examples**

```
data(electors)
summary(mclogit(
  cbind(Freq,interaction(time,class))~econ.left+welfare+auth,
  data=electors))
summary(mclogit(
  \verb|cbind| (\texttt{Freq}, \texttt{interaction}(\texttt{time}, \texttt{class})) \\ ^- econ. \\ \texttt{left/class+welfare/class+auth/class}, \\
  data=electors))
summary(mclogit(
  cbind(Freq,interaction(time,class))~econ.left/class+welfare/class+auth/class,
  random=~1|party.time,
  data=within(electors,party.time<-interaction(party,time))))</pre>
summary(mclogit(
  cbind(Freq,interaction(time,class))~econ.left/(class*time)+welfare/class+auth/class,
  random=~1|party.time,
  data=within(electors,{
         party.time <-interaction(party,time)</pre>
         econ.left.sq <- (econ.left-mean(econ.left))^2</pre>
         })))
```

getSummary-methods

'getSummary' Methods

### **Description**

getSummary methods for use by mtable

### Usage

getSummary-methods 5

### **Arguments**

obj an object returned by mblogit or mclogit

alpha level of the confidence intervals; their coverage should be 1-alpha/2

rearrange an optional named list of character vectors. Each element of the list designates

a column in the table of estimates, and each element of a character vector refers to a coefficient. Names of list elements become column heads and names of the

character vector elements become coefficient labels.

... further arguments; ignored.

### **Examples**

```
## Not run:
summary(classd.model <- mclogit(cbind(Freq,choice.set)~</pre>
                   (econdim1.sq+nonmatdim1.sq+nonmatdim2.sq)+
                   (econdim1+nonmatdim1+nonmatdim2)+
                   (econdim1+nonmatdim1+nonmatdim2):classd,
                  data=mvoteint.classd,random=~1|mvoteint/eb,
                  subset=classd!="Farmers"))
myGetSummary.classd <- function(x)getSummary.mclogit(x,rearrange=list(</pre>
        "Econ. Left/Right"=c(
                    "Squared effect"="econdim1.sq",
                    "Linear effect"="econdim1",
                    " x Intermediate/Manual worker"="econdim1:classdIntermediate",
                    " x Service class/Manual worker"="econdim1:classdService class",
                    " x Self-employed/Manual worker"="econdim1:classdSelf-employed"
                    ),
        "Lib./Auth."=c(
                    "Squared effect"="nonmatdim1.sq",
                    "Linear effect"="nonmatdim1",
                    " x Intermediate/Manual worker"="nonmatdim1:classdIntermediate",
                    " x Service class/Manual worker"="nonmatdim1:classdService class",
                    " x Self-employed/Manual worker"="nonmatdim1:classdSelf-employed"
        "Mod./Trad."=c(
                    "Squared effect"="nonmatdim2.sq",
                    "Linear effect"="nonmatdim2",
                    " x Intermediate/Manual worker"="nonmatdim2:classdIntermediate",
                    " x Service class/Manual worker"="nonmatdim2:classdService class",
```

6 mblogit

```
" x Self-employed/Manual worker"="nonmatdim2:classdSelf-employed"
      ))
library(memisc)
mtable(classd.model,getSummary=myGetSummary.classd)
# Output would look like so:
# -----
                            Econ. Left/Right Lib./Auth. Mod./Trad.
                                              0.008
# Squared effect
                                 0.030
                                                             -0.129**
                                (0.081)
                                               (0.041)
                                                              (0.047)
# Linear effect
                                -0.583***
                                              -0.038
                                                              0.137**
                                              (0.041)
                                (0.063)
                                                              (0.045)
  x Intermediate/Manual worker
                                0.632***
                                              -0.029
                                                             -0.015
                                              (0.020)
                                (0.026)
                                                             (0.019)
#
  x Service class/Manual worker
                                1.158***
                                              0.084**
                                                              0.000
                                (0.040)
                                               (0.032)
                                                             (0.030)
  x Self-employed/Manual worker
                                1.140***
                                               0.363***
                                                              0.112***
                                (0.035)
                                               (0.027)
                                                              (0.026)
# Var(mvoteint)
                                1.080***
                                (0.000)
                                 0.118***
# Var(mvoteint x eb)
                                (0.000)
# Dispersion
                                    1.561
# Deviance
                                 15007.0
                                173445
## End(Not run)
```

mblogit

Baseline-Category Logit Models for Categorical and Multinomial Responses

### **Description**

The function mblogit fits baseline-category logit models for categorical and multinomial count responses with fixed alternatives.

### Usage

```
mblogit(
  formula,
  data = parent.frame(),
  random = NULL,
  subset,
  weights = NULL,
```

mblogit 7

```
na.action = getOption("na.action"),
model = TRUE,
x = FALSE,
y = TRUE,
contrasts = NULL,
method = NULL,
estimator = c("ML", "REML"),
dispersion = FALSE,
from.table = FALSE,
groups = NULL,
control = if (length(random)) mmclogit.control(...) else mclogit.control(...),
...
)
```

### Arguments

formula the model formula. The response must be a factor or a matrix of counts.

data an optional data frame, list or environment (or object coercible by as.data.frame

to a data frame) containing the variables in the model. If not found in data, the variables are taken from environment(formula), typically the environment

from which glm is called.

random an optional formula that specifies the random-effects structure or NULL.

subset an optional vector specifying a subset of observations to be used in the fitting

process.

weights an optional vector of weights to be used in the fitting process. Should be NULL

or a numeric vector.

na.action a function which indicates what should happen when the data contain NAs. The

default is set by the na.action setting of options, and is na.fail if that is unset. The 'factory-fresh' default is na.omit. Another possible value is NULL,

no action. Value na. exclude can be useful.

model a logical value indicating whether *model frame* should be included as a compo-

nent of the returned value.

x, y logical values indicating whether the response vector and model matrix used in

the fitting process should be returned as components of the returned value.

contrasts an optional list. See the contrasts.arg of model.matrix.default.

method NULL or a character string, either "PQL" or "MQL", specifies the type of the

quasilikelihood approximation to be used if a random-effects model is to be

estimated.

estimator a character string; either "ML" or "REML", specifies which estimator is to be

used/approximated.

dispersion a logical value or a character string; whether and how a dispersion parameter

should be estimated. For details see dispersion.

from. table a logical value; do the data represent a contingency table, e.g. were created by

applying as.data.frame() a the result of table() or xtabs(). This relevant only if the response is a factor. This argument should be set to TRUE if the data do

8 mblogit

	come from a contingency table. Correctly setting from. table=TRUE in this case, will lead to efficiency gains in computing, but more importantly overdispersion will correctly be computed if present.
groups	an optional formula that specifies groups of observations relevant for the specification of overdispersed response counts.
control	a list of parameters for the fitting process. See mclogit.control
	arguments to be passed to mclogit.control or mmclogit.control

#### **Details**

The function mblogit internally rearranges the data into a 'long' format and uses mclogit. fit to compute estimates. Nevertheless, the 'user data' is unaffected.

#### Value

mblogit returns an object of class "mblogit", which has almost the same structure as an object of class "glm". The difference are the components coefficients, residuals, fitted.values, linear.predictors, and y, which are matrices with number of columns equal to the number of response categories minus one.

### References

```
Agresti, Alan (2002). Categorical Data Analysis. 2nd ed, Hoboken, NJ: Wiley. https://doi.org/10.1002/0471249688
```

Breslow, N.E. and D.G. Clayton (1993). "Approximate Inference in Generalized Linear Mixed Models". *Journal of the American Statistical Association* 88 (421): 9-25. https://doi.org/10.1080/01621459.1993.10594284

#### See Also

The function multinom in package **nnet** also fits multinomial baseline-category logit models, but has a slightly less convenient output and does not support overdispersion or random effects. However, it provides some other options. Baseline-category logit models are also supported by the package **VGAM**, as well as some reduced-rank and (semi-parametric) additive generalisations. The package **mnlogit** estimates logit models in a way optimized for large numbers of alternatives.

### **Examples**

mclogit 9

```
summary(house.mult)
summary(house.mblogit)
mtable(house.mblogit)
```

mclogit

Conditional Logit Models and Mixed Conditional Logit Models

### Description

mclogit fits conditional logit models and mixed conditional logit models to count data and individual choice data, where the choice set may vary across choice occasions.

Conditional logit models without random effects are fitted by Fisher-scoring/IWLS. Models with random effects (mixed conditional logit models) are estimated via maximum likelihood with a simple Laplace approximation (aka PQL).

### Usage

### **Arguments**

formula

a model formula: a symbolic description of the model to be fitted. The left-hand side contains is expected to be a two-column matrix. The first column contains the choice counts or choice indicators (alternative is chosen=1, is not chosen=0). The second column contains unique numbers for each choice set.

If individual-level data is used, choice sets correspond to the individuals, if aggregated data with choice counts are used, choice sets may e.g. correspond to covariate classes within clusters.

10 mclogit

The right-hand of the formula contains choice predictors. It should be noted that constants are deleted from the formula as are predictors that do not vary within choice sets. data an optional data frame, list or environment (or object coercible by as.data.frame to a data frame) containing the variables in the model. If not found in data, the variables are taken from environment (formula), typically the environment from which glm is called. random an optional formula that specifies the random-effects structure or NULL. weights an optional vector of weights to be used in the fitting process. Should be NULL or a numeric vector. offset an optional model offset. Currently only supported for models without random subset an optional vector specifying a subset of observations to be used in the fitting process. na.action a function which indicates what should happen when the data contain NAs. The default is set by the na.action setting of options, and is na.fail if that is unset. The 'factory-fresh' default is na. omit. Another possible value is NULL, no action. Value na. exclude can be useful. start an optional numerical vector of starting values for the conditional logit parameters. mode1 a logical value indicating whether model frame should be included as a component of the returned value. logical values indicating whether the response vector and model matrix used in х, у the fitting process should be returned as components of the returned value. contrasts an optional list. See the contrasts.arg of model.matrix.default. method NULL or a character string, either "PQL" or "MQL", specifies the type of the quasilikelihood approximation to be used if a random-effects model is to be estimated. a character string; either "ML" or "REML", specifies which estimator is to be estimator used/approximated. dispersion a real number used as dispersion parameter; a character vector that specifies the method to compute the dispersion; a logical value – if TRUE the default method ("Afroz") is used, if FALSE, the dispersion parameter is set to 1, that is, no dispersion. For details see dispersion. a list of parameters for the fitting process. See mclogit.control control arguments to be passed to mclogit.control or mmclogit.control object an object that inherits class "mclogit". When passed to dispersion(), it should be the result of a call of mclogit() of mblogit(), without random efa changes to the model formula, see update.default and update.formula. formula. correlation logical; see summary.lm.

symbolic.cor

logical; see summary.lm.

mclogit 11

#### Value

mclogit returns an object of class "mclogit", which has almost the same structure as an object of class "glm".

#### Note

Covariates that are constant within choice sets are automatically dropped from the model formula specified by the formula argument of mclogit.

If the model contains random effects, these should

- either vary within choice sets (e.g. the levels of a factor that defines the choice sets should not be nested within the levels of factor)
- or be random coefficients of covariates that vary within choice sets.

In earlier versions of the package (prior to 0.6) it will lead to a failure of the model fitting algorithm if these conditions are not satisfied. Since version 0.6 of the package, the function mclogit will complain about such model a misspecification explicitely.

#### References

Agresti, Alan (2002). Categorical Data Analysis. 2nd ed, Hoboken, NJ: Wiley. https://doi.org/10.1002/0471249688

Breslow, N.E. and D.G. Clayton (1993). "Approximate Inference in Generalized Linear Mixed Models". *Journal of the American Statistical Association* 88 (421): 9-25. https://doi.org/10.1080/01621459.1993.10594284

Elff, Martin (2009). "Social Divisions, Party Positions, and Electoral Behaviour". *Electoral Studies* 28(2): 297-308. https://doi.org/10.1016/j.electstud.2009.02.002

McFadden, D. (1973). "Conditionial Logit Analysis of Qualitative Choice Behavior". Pp. 105-135 in P. Zarembka (ed.). *Frontiers in Econometrics*. New York: Wiley. https://eml.berkeley.edu/reprints/mcfadden/zarembka.pdf

#### See Also

Conditional logit models are also supported by **gmnl**, **mlogit**, and **survival**. **survival** supports conditional logit models for binary panel data and case-control studies. **mlogit** and **gmnl** treat conditional logit models from an econometric perspective. Unlike the present package, they focus on the random utility interpretation of discrete choice models and support generalisations of conditional logit models, such as nested logit models, that are intended to overcome the IIA (indipendence from irrelevant alterantives) assumption. Mixed multinomial models are also supported and estimated using simulation-based techniques. Unlike the present package, mixed or random-effects extensions are mainly intended to fit repeated choices of the same individuals and not aggregated choices of many individuals facing identical alternatives.

### **Examples**

```
data(Transport)
summary(mclogit(
```

12 mclogit.control

```
cbind(resp,suburb)~distance+cost,
  data=Transport
  ))

data(electors)

summary(mclogit(
  cbind(Freq,interaction(time,class))~econ.left/class+welfare/class+auth/class,
  random=~1|party.time,
  data=within(electors,party.time<-interaction(party,time))))</pre>
```

mclogit.control

Control Parameters for the Fitting Process

### **Description**

mclogit.control returns a list of default parameters that control the fitting process of mclogit.

### Usage

### **Arguments**

break.on.negative

positive convergence tolerance  $\epsilon$ ; the iterations converge when  $|dev-dev_{old}|/(|dev|+$ epsilon 0.1) <  $\epsilon$ . maxit integer giving the maximal number of IWLS or PQL iterations. trace logical indicating if output should be produced for each iteration. logical; indicating if output should be produced for each inner iteration of the trace.inner PQL method. avoid.increase logical; should an increase of the deviance be avoided by step truncation? break.on.increase logical; should an increase of the deviance be avoided by stopping the algorithm? break.on.infinite logical; should an infinite deviance stop the algorithm instead of leading to step truncation?

logical; should a negative deviance stop the algorithm?

mclogit.fit 13

### Value

A list.

mclogit.fit Internal functions used for model fit.

### **Description**

These functions are exported and documented for use by other packages. They are not intended for end users.

### Usage

### Arguments

У	a response vector. Should be binary.
S	a vector identifying individuals or covariate strata
W	a vector with observation weights.
Χ	a model matrix; required.
dispersion	a logical value or a character string; whether and how a dispersion parameter should be estimated. For details see dispersion.
Z	the random effects design matrix.
groups	a list of grouping factors.
start	an optional numerical vector of starting values for the coefficients.
offset	an optional model offset. Currently only supported for models without random effects.
method	a character string, either "PQL" or "MQL", specifies the type of the quasilikelihood approximation.
estimator	a character string; either "ML" or "REML", specifies which estimator is to be used/approximated.

a list of parameters for the fitting process. See mclogit.control

### Value

control

A list with components describing the fitted model.

14 simulate.mclogit

simulate.mclogit	Simulating responses from baseline-category and conditional logit
	models.

#### Description

The simulate() methods allow to simulate responses from models fitted with mclogit() and mblogit(). Currently only models *without* random effects are supported for this.

#### Usage

```
## S3 method for class 'mblogit'
simulate(object, nsim = 1, seed = NULL, ...)
## S3 method for class 'mclogit'
simulate(object, nsim = 1, seed = NULL, ...)

# These methods are currently just 'stubs', causing an error
# message stating that simulation from models with random
# effects are not supported yet
## S3 method for class 'mmblogit'
simulate(object, nsim = 1, seed = NULL, ...)
## S3 method for class 'mmclogit'
simulate(object, nsim = 1, seed = NULL, ...)
```

### Arguments

object	an object from the relevant class
nsim	a number, specifying the number of simulated responses for each observation.
seed	an object specifying if and how the random number generator should be initialized ('seeded'). The interpetation of this argument follows the default method, see link[stats]{simulate}
	other arguments, ignored.

#### Value

The result of the simulate method for objects created by mclogit is a data frame with one variable for each requested simulation run (their number is given by the nsim= argument). The contents of the columns are counts (or zero-one values), with group-wise multinomial distribution (within choice sets) just like it is assumed for the original response.

The shape of the result of the simulate method for objects created by mblogit is also a data frame. The variables within the data frame have a mode or shape that corresponds to the response to which the model was fitted. If the response is a matrix of counts, then the variables in the data frame are also matrices of counts. If the response is a factor and mblogit was called with an argument from.table=FALSE, the variables in the data frame are factors with the same factor levels as the response to which the model was fitted. If instead the function was called with from.table=TRUE, the variables in the data frame are counts, which represent frequency weights that would result from applying as.data.frame to a contingency table of simulated frequency counts.

Transport 15

### **Examples**

```
library(MASS)
(house.mblogit <- mblogit(Sat ~ Infl + Type + Cont,</pre>
                             data = housing,
                             weights=Freq,
                             from.table=TRUE))
sm <- simulate(house.mblogit,nsim=7)</pre>
housing.long <- housing[rep(seq.int(nrow(housing)),housing$Freq),]</pre>
(housel.mblogit <- mblogit(Sat ~ Infl + Type + Cont,</pre>
                             data=housing.long))
sml <- simulate(housel.mblogit,nsim=7)</pre>
housing.table <- xtabs(Freq~.,data=housing)</pre>
housing.mat <- memisc::to.data.frame(housing.table)</pre>
head(housing.mat)
(housem.mblogit <- mblogit(cbind(Low, Medium, High) ~</pre>
                                  Infl + Type + Cont,
                              data=housing.mat))
smm <- simulate(housem.mblogit,nsim=7)</pre>
str(sm)
str(sml)
str(smm)
head(smm[[1]])
```

Transport

Choice of Means of Transport

### **Description**

This is an artificial data set on choice of means of transport based on cost and walking distance.

### Usage

```
data(Transport)
```

### **Format**

A data frame containing the following variables:

```
transport means of transportation that can be chosen.
suburb identifying number for each suburb
distance walking distance to bus or train station
cost cost of each means of transportation
working size of working population of each suburb
```

Transport Transport

**prop.true** true choice probabilities**resp** choice frequencies of means of transportation

# **Index**

```
* datasets
                                                mmclogit.fitPQLMQL (mclogit.fit), 13
    electors, 3
                                                mtable, 4
    Transport, 15
                                                multinom, 8
* models
                                                na.exclude, 7, 10
    mclogit, 9
                                                na.fail, 7, 10
* regression
                                                na.omit, 7, 10
    mclogit, 9
                                                options, 7, 10
AIC.mclogit (mclogit), 9
anova.mclogit(mclogit), 9
                                                predict.mblogit(mblogit), 6
as.data.frame, 7, 10, 14
                                                predict.mblogit, (mblogit), 6
                                                predict.mclogit(mclogit), 9
BIC.mclogit (mclogit), 9
                                                print.mblogit(mblogit), 6
deviance.mclogit (mclogit), 9
                                                print.mclogit (mclogit), 9
dispersion, 2, 7, 10, 13
                                                print.mmblogit(mblogit), 6
                                                print.summary.mblogit(mblogit), 6
electors, 3
                                                print.summary.mclogit(mclogit), 9
                                                print.summary.mmblogit(mblogit), 6
fitted.mblogit(mblogit), 6
                                                print.summary.mmclogit(mclogit), 9
fitted.mclogit (mclogit), 9
                                                 residuals.mclogit (mclogit), 9
getSummary, 4
getSummary-methods, 4
                                                 simulate, 14
getSummary.mblogit
                                                 simulate.mblogit(simulate.mclogit), 14
        (getSummary-methods), 4
                                                 simulate.mclogit, 14
getSummary.mclogit
                                                 simulate.mmblogit(simulate.mclogit), 14
        (getSummary-methods), 4
                                                 simulate.mmclogit(simulate.mclogit), 14
getSummary.mmblogit
                                                 summary.lm, 10
        (getSummary-methods), 4
                                                 summary.mblogit(mblogit), 6
getSummary.mmclogit
                                                summary.mclogit(mclogit), 9
        (getSummary-methods), 4
                                                 summary.mmblogit(mblogit), 6
glm, 8, 11
                                                 summary.mmclogit (mclogit), 9
logLik.mclogit (mclogit), 9
                                                Transport, 15
mblogit, 5, 6, 14
                                                update.default, 10
mclogit, 5, 9, 14
                                                update.formula, 10
mclogit.control, 8, 10, 12, 13
                                                update.mclogit(mclogit), 9
mclogit.fit, 8, 13
mmclogit.control (mclogit.control), 12
                                                vcov.mclogit (mclogit), 9
```

18 INDEX

```
\begin{tabular}{ll} weights.mblogit (mblogit), 6 \\ weights.mclogit (mclogit), 9 \\ \end{tabular}
```