

# Package ‘nmm’

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**Type** Package

**Title** Nonlinear Multivariate Models

**Version** 0.9

**Description** Estimates a subset of nonlinear multivariate models (NMM):

system of nonlinear regressions (SNR), logit, and a joint model of SNR and logit.

‘nmm’ uniquely accounts for correlations between the error terms from nonlinear regressions and the probabilities from logit models.

It also enables a very flexible design of logit: alternative-specific indirect utilities, individual-specific choice set and number of actual choices.

**Imports** Rdpack, AER, mlogit, Hmisc, stats, gsubfn, abind, tidyverse, dplyr,

**Depends** R (>= 4.0), systemfit, DEoptim, data.table, magrittr, maxLik

**RdMacros** Rdpack

**License** GPL (>= 2)

**Encoding** UTF-8

**LazyData** true

**RoxygenNote** 7.1.1

**Suggests** knitr, rmarkdown, recipes, Stat2Data

**NeedsCompilation** no

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## R topics documented:

addInter . . . . .	2
add_variable . . . . .	3

AICc . . . . .	4
cond_expr . . . . .	5
cont_stats . . . . .	6
convert_attr2exp . . . . .	7
dat4cond_mean_cov_expr . . . . .	7
dataM . . . . .	8
datmaxle . . . . .	9
datmlsem . . . . .	10
diagnostics . . . . .	10
expr_ll_norm . . . . .	11
expr_ll_norm_v2 . . . . .	12
extract_attr_deriv . . . . .	12
formula2string . . . . .	13
f_create . . . . .	14
get_npar . . . . .	15
get_par . . . . .	16
get_start . . . . .	16
grad_hess_eval . . . . .	19
in2nmm . . . . .	20
logLik.nmm . . . . .	21
MAEDtimeExpenditure . . . . .	22
MAEDtravel . . . . .	24
maxle . . . . .	26
maxle_p . . . . .	27
mlsem . . . . .	28
MNlogitf . . . . .	28
nmm . . . . .	30
prepare_data . . . . .	33
pseudoR . . . . .	35
replace_par . . . . .	36
replace_par_wrap . . . . .	37
stats_function . . . . .	38
string2formula . . . . .	39
wxMaxima . . . . .	39

**Index****41**

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addInter*Add interactions*

---

**Description**

`addInter` add interactions into continuous equations.

**Usage**

```
addInter(eqcont, par_c, intv, inter_pars)
```

**Arguments**

eqcont	Vector of strings containing equations.
parl_c	Names of coefficients.
intv	Vector of integers corresponding to coefficients to which interactions should be added.
inter_parl	Names of new coefficients (interactions).

**Value**

list: 1 - expressions of errors, equations, parameters to estimate

**Examples**

```
eq_c <- c("Tw ~ tw*w + ph1*Tc", "Tf1 ~ (1+w)^tw + ph1^3*Tc")
parl <- c("tw", "ph1")
intv <- c(1,0)
inter_parl <- c('yytw','yyph1')
res <- addInter(eq_c, parl, intv, inter_parl)
```

add\_variable

add\_variable *adds columns to the data matrix*

**Description**

add\_variable adds columns to the data matrix

**Usage**

```
add_variable(data = data, dname = "chc", weights = NULL)
```

**Arguments**

data	data.frame, if dname=="chc" columns "chc_i" has to be in the data.
dname	if dname=="chc" (dummy for chosen alternative) dummy for the choice alternative added, if "weights" weights added
weights	Matrix with weights to be added to the data

**Value**

data.frame

**Examples**

```
chc <- c(1,2,1,4,3,1,4)
data <- data.frame(choice=chc, x=rnorm(length(chc)), y=rnorm(length(chc)))
add_variable(data, dname="chc")
ww <- c(1,1,1,2,2,2,3)
add_variable(data, dname="weights", weights=ww)
```

AICc                    *Adjusted Akaike's Information Criterion.*

## Description

Calculates adjusted and Bayesian Information Criterion for nmm object

## Usage

```
AICc(object, ..., k = 2)

## S3 method for class 'nmm'
AICc(object, ..., k = 2)

## Default S3 method:
AICc(object, ..., k = 2)

## S3 method for class 'nmm'
BIC(object, ..., k = 2)
```

## Arguments

object	Fitted nmm model.
...	Not used.
k	Multiplication factor.

## Value

a numeric value with the corresponding AIC, AICc, BIC.

## Examples

```
library(systemfit)
data( ppine , package="systemfit")
hg.formula <- hg ~ exp( h0 + h1*log(tht) + h2*tht^2 + h3*elev)
dg.formula <- dg ~ exp( d0 + d1*log(dbh) + d2*hg + d3*cr)
labels <- list( "height.growth", "diameter.growth" )
model <- list( hg.formula, dg.formula )
start.values <- c(h0=-0.5, h1=0.5, h2=-0.001, h3=0.0001,
                  d0=-0.5, d1=0.009, d2=0.25, d3=0.005)
model.sur <- nlsystemfit( "SUR", model, start.values, data=ppine, eqnlables=labels )
eq_c <- as.character(c(hg.formula, dg.formula))
parl <- c(paste0("h", 0:3), paste0("d", 0:3))
res <- nmm(ppine, eq_c=eq_c, start_v=start.values, par_c=parl,
           eq_type = "cont", best_method = FALSE)
aa <- in2nmm(res, model.sur$b)
AICc(res)
AICc(aa)
```

```
AIC(res)
AIC(aa)
BIC(res)
BIC(aa)
```

**cond\_expr**

*cond\_expr* returns moments of conditional multivariate normal distribution  $X|Y$  (last variable is dependent). Only expression for  $X|Y$ . Requires installation of Maxima software.

**Description**

`cond_expr` returns moments of conditional multivariate normal distribution  $X|Y$  (last variable is dependent). Only expression for  $X|Y$ . Requires installation of Maxima software.

**Usage**

```
cond_expr(neq, sdv, mv, nconeq = neq - 1, tex = FALSE)
```

**Arguments**

<code>neq</code>	Number of equations/variables.
<code>sdv</code>	Vector of standard deviation of normally distributed variables, e.g. <code>c(NA, NA, NA, 1)</code> NA - unknown, any number - know.
<code>mv</code>	Vector of means of normally distributed variables, e.g. <code>rep(0, 4)</code> .
<code>nconeq</code>	Number of continuous equations.
<code>tex</code>	<code>i</code> if TRUE TeX expressions from wxMaxima are returned.

**Value**

List of strings. First element is an expression of conditional mean and covariance. The second element is a TeX formula.

**Examples**

```
# this means that E[y3|y1,y2] and V[y3|y1,y2] will be returned
# all continuous w/ unknown means
## Not run:
# To run this, one needs to install Maxima software
res <- cond_expr(neq=3)
# 3 continuous w/ unknown means and the last one with mean 0 and sd 1, d1|c1c2c3
res <- cond_expr(neq=4, sdv=c(NA, NA, NA, 1), mv=c(NA, NA, NA, 0))
# 2 continuous w/ unknown means and 2 discrete with mean 0 and sd 1, d1|c1c2c3d2
res <- cond_expr(neq=4, sdv=c(NA, NA, 1, 1), mv=c(NA, NA, 0, 0), nconeq=2)

## End(Not run)
```

---

cont_stats	<i>Goodness of fit measures</i>
------------	---------------------------------

---

## Description

Calculate RMSE, MAPE, R<sup>2</sup> and adjusted R<sup>2</sup>

## Usage

```
cont_stats(
  x,
  which = c("all", "RMSE", "MAPE", "Rx2", "Rx2adj"),
  only_total = FALSE
)
```

## Arguments

- x Fitted nmm model.
- which What to calculate. Options: "all", "RMSE", "MAPE", "Rx2", "Rx2adj".
- only\_total If TRUE, calculate statistics only for totals.

## Value

matrix with Goodness of fit measures

## Examples

```
library(systemfit)
data(ppine , package="systemfit")
hg.formula <- hg ~ exp( h0 + h1*log(tht) + h2*tht^2 + h3*elev)
dg.formula <- dg ~ exp( d0 + d1*log(dbh) + d2*hg + d3*cr)
labels <- list( "height.growth", "diameter.growth" )
model <- list( hg.formula, dg.formula )
start.values <- c(h0=-0.5, h1=0.5, h2=-0.001, h3=0.0001,
                  d0=-0.5, d1=0.009, d2=0.25, d3=0.005)
model.sur <- nlsystemfit( "SUR", model, start.values, data=ppine, eqnlabels=labels )
eq_c <- as.character(c(hg.formula, dg.formula))
parl <- c(paste0("h", 0:3), paste0("d", 0:3))
res <- nmm(ppine, eq_c=eq_c, start_v=start.values, par_c=parl,
           eq_type = "cont", best_method = FALSE)
cont_stats(res, which = "all")
```

convert\_attr2exp

*convert\_attr2exp converts symbolic attribute of derivative into expression object.*

**Description**

convert\_attr2exp converts symbolic attribute of derivative into expression object.

**Usage**

```
convert_attr2exp(obj)
```

**Arguments**

obj	Symbolic expression of gradient or hessian
-----	--------------------------------------------

**Value**

combine expression of derivatives

**Examples**

```
eq1 <- parse(text="2*(log(sin(x)/log(x)))+x^4*log(x)+cos(y+x)")  
tt1 <- deriv(eq1, c("x", "y"), hessian=TRUE)  
r1 <- convert_attr2exp(extract_attr_deriv(tt1, "grad"))  
r2 <- convert_attr2exp(extract_attr_deriv(tt1, "hessian"))
```

dat4cond\_mean\_cov\_expr

*Log-likelihood expressions for cont. equations plus 1 discrete*

**Description**

Log-likelihood expressions for cont. equations plus 1 discrete

**Usage**

```
data(dat4cond_mean_cov_expr)
```

**Format**

An object of class list of length 4.

**Examples**

```
data(dat4cond_mean_cov_expr)
```

---

**dataM***Example dataset*

---

## Description

Data "MathPlacement" taken from Stat2Data package.

## Usage

```
data(dataM)
```

## Format

A data frame containing:

**Student** Identification number for each student

**Gender** 0=Female, 1=Male

**PSATM** PSAT score in Math

**SATM** SAT score in Math

**ACTM** ACT Score in Math

**Rank** Adjusted rank in HS class

**Size** Number of students in HS class

**GPAadj** Adjusted GPA

**PlemtScore** Score on math placement exam

**Recommends** Recommended course: R0 R01 R1 R12 R2 R3 R4 R6 R8

**Course** Actual course taken

**Grade** Course grade

**RecTaken** 1=recommended course, 0=otherwise

**TooHigh** 1=took course above recommended, 0=otherwise

**TooLow** 1=took course below recommended, 0=otherwise

**CourseSuccess** 1=B or better grade, 0=grade below B

**DR\_Course** according to recommendations, which level of course was taken: alow - lower, bnornal - recommended, chigh - higher

## Details

Code for data modifications can be found in the example section.

## Examples

```

data(dataM)
library(magrittr)
library(dplyr)
if (requireNamespace("recipes", quietly = TRUE)&requireNamespace("Stat2Data", quietly = TRUE)) {
  data("MathPlacement", package="Stat2Data")
  head(MathPlacement)
  library(recipes)
  # As some of the data is missing, k-nearest neighbors (knn) imputation is
  # used to fill the gaps. This is done with recipes package and function
  # step_knnimpute.
  dataM <- recipe(~ ., data = MathPlacement) %>%
    step_knnimpute(everything()) %>% prep() %>% juice()
  # Afterwards we create a categorical variable that will show whether a
  # student took a course which was too high, too low, the recommended one or
  # something else happened:
  dataM %>>% mutate(Student = 1:n(), DR_Course = case_when(
    TooHigh == 1 ~ "chigh",
    TooLow == 1 ~ "alow",
    RecTaken == 1 ~ "bnormal",
    TRUE ~"dother"
  ))
  # We remove observations with ambiguous course status:
  dataM %>>% filter(DR_Course!="dother")
  dataM %>% select(DR_Course) %>% table %>% t
}

```

datmaxle

*Log-likelihood expressions for cont. equations*

## Description

Log-likelihood expressions for cont. equations

## Usage

```
data(datmaxle)
```

## Format

An object of class list of length 4.

## Examples

```
data(datmaxle)
```

datmlsem

*Log-likelihood expressions for cont. equations sem***Description**

Log-likelihood expressions for cont. equations sem

**Usage**

```
data(datmlsem)
```

**Format**

An object of class `list` of length 4.

**Examples**

```
data(datmlsem)
```

diagnostics

*Goodness of fit measures for both parts***Description**

Calculation RMSE, misclassification and other goodness of fit measures.

**Usage**

```
diagnostics(
  x,
  xdigit = 4,
  which = "all",
  only_total = FALSE,
  cPseudoR = TRUE,
  cRs = TRUE
)
```

**Arguments**

- |                         |                                                                     |
|-------------------------|---------------------------------------------------------------------|
| <code>x</code>          | Fitted <code>nmm</code> model.                                      |
| <code>xdigit</code>     | rounding number                                                     |
| <code>which</code>      | What to calculate. Options: "all", "RMSE", "MAPE", "Rx2", "Rx2adj". |
| <code>only_total</code> | If TRUE, calculate statistics only for the whole system             |
| <code>cPseudoR</code>   | If TRUE, calculate pseudo R <sup>2</sup> s.                         |
| <code>cRs</code>        | Include "AIC", "AICc", "BIC"                                        |

**Value**

matrix with goodness of fit measures. attribute `corr` holds empirical variance-covariance matrix.

**Examples**

```

library(systemfit)
data( ppine , package="systemfit")
hg.formula <- hg ~ exp( h0 + h1*log(tht) + h2*tht^2 + h3*elev)
dg.formula <- dg ~ exp( d0 + d1*log(dbh) + d2*hg + d3*cr)
labels <- list( "height.growth", "diameter.growth" )
model <- list( hg.formula, dg.formula )
start.values <- c(h0=-0.5, h1=0.5, h2=-0.001, h3=0.0001,
                  d0=-0.5, d1=0.009, d2=0.25, d3=0.005)
model.sur <- nlsystemfit( "SUR", model, start.values, data=ppine, eqnlables=labels )
eq_c <- as.character(c(hg.formula, dg.formula))
parl <- c(paste0("h", 0:3), paste0("d", 0:3))
start.values <- c(h0=-0.5, h1=0.5, h2=-0.001, h3=0.0001,
                  d0=-0.5, d1=0.009, d2=0.25, d3=0.005)
res <- nmm(ppine, eq_c=eq_c, start_v=start.values, par_c=parl, eq_type = "cont",
best_method = FALSE)
ressur <- in2nmm(res, new_coef=model.sur$b)
diagnostics(res)
diagnostics(ressur)

#example discrete
library(mlogit)
data("Fishing", package = "mlogit")
Fish <- mlogit.data(Fishing, varying = c(2:9), shape = "wide", choice = "mode")
## a pure "conditional" model
mres <- summary(mlogit(mode ~ price + catch, data = Fish))
data <- prepare_data(Fish %>% data.frame %>% dplyr::select(-idx),
choice="alt", dummy="mode", PeID="chid", mode_spec_var = c("price", "catch"),
type="long")
eq_d <- c("a1 + p1 * price_1 + p2 * catch_2", "a2 + p1 * price_2 + p2 * catch_2",
          "a3 + p1 * price_3 + p2 * catch_3", "a4 + p1 * price_4 + p2 * catch_4")
par_d <- c(paste0("a", 1:4), paste0("p", 1:2))
res <- nmm(data, eq_d=eq_d, par_d=par_d, eq_type="disc")
ncoef <- mres$coefficients
names(ncoef) <- par_d[-1]
resdisc <- in2nmm(res, new_coef = ncoef)
a <- diagnostics(res, xdigit=2)
a2 <- diagnostics(resdisc)
attributes(a2)$corr

```

**Description**

Log-likelihood expressions for cont. equations

**Usage**

```
data(expr_ll_norm)
```

**Format**

An object of class list of length 8.

**Examples**

```
data(expr_ll_norm)
```

expr\_ll\_norm\_v2

*Another Log-likelihood expressions for cont. equations version 2*

**Description**

Another Log-likelihood expressions for cont. equations version 2

**Usage**

```
data(expr_ll_norm_v2)
```

**Format**

An object of class list of length 8.

**Examples**

```
data(expr_ll_norm_v2)
```

extract\_attr\_deriv

*extract\_attr\_deriv converts attributes(hessian/gradient) of deriv() into a matrix of character strings.*

**Description**

`extract_attr_deriv` converts attributes(hessian/gradient) of `deriv()` into a matrix of character strings.

**Usage**

```
extract_attr_deriv(ex, attribute)
```

**Arguments**

ex	Expression of derivative. Results of <code>deriv()</code> .
attribute	"grad" for gradient or "hessian" for the Hessian matrix.

**Value**

Returns a matrix of character strings.

**Examples**

```
eq <- parse(text="2*(log(sin(x)/log(x)))+x^4*log(x)+cos(y+x)")  
tt <- deriv(eq, c("x", "y"), hessian=TRUE)  
g <- tt%>%extract_attr_deriv(., attribute = "grad")  
h <- tt%>%extract_attr_deriv(., attribute = "hessian")
```

**formula2string**

*formula2string removes square brackets from the supplied expressions. Convert par[2] -> par2, sigma[2] -> sigma\_2\_, sigma[2,3] -> sigma\_2x2*

**Description**

`formula2string` removes square brackets from the supplied expressions. Convert `par[2]` -> `par2`, `sigma[2]` -> `sigma_2_`, `sigma[2,3]` -> `sigma_2x2`

**Usage**

```
formula2string(x)
```

**Arguments**

x	String with square brackets.
---	------------------------------

**Value**

String without square brackets.

**Examples**

```
xx <- "par[1]*3+par[2]*par+rho1[1,2]+sigma1[2]"  
formula2string(xx)
```

---

**f\_create***f\_create creates functions for log-likelihood of different models.*

---

## Description

*f\_create* creates functions for log-likelihood of different models.

## Usage

```
f_create(
  mn,
  data,
  fixed = 0,
  cheqs0 = NULL,
  separatenmm = FALSE,
  probt = NULL,
  tformula = NULL,
  hessian = NULL,
  transform = TRUE,
  sume = NULL
)
```

## Arguments

<code>mn</code>	Expression, can be a list of equations.
<code>data</code>	Name of the data frame with which the function will be evaluated.
<code>fixed</code>	Integer, which parameter is fixed to be 0.
<code>cheqs0</code>	If continuous are supplied, include the expressions of errors.
<code>separatenmm</code>	if TRUE, separate log-likelihood for each equations is produced.
<code>probt</code>	Expressions of un-simplified probabilities with quantile transformation(qnorm(ifelse(P))).
<code>tformula</code>	unimplified log(P)
<code>hessian</code>	Adds lines to check the Hessian, hessian should be the name of hessian function.
<code>transform</code>	if TRUE, adds lines to check conditional means
<code>sume</code>	Expression of summed likelihoods.

## Value

Function.

## Examples

```
eq_d <- c("ASC1 * 1 + B11_dur * dur_1" , "ASC2 * 1 + B12_dur * dur_2",
"ASC3 * 1 + B13_dur * dur_3 + B20_cost * cost_3 + B53_parkman * PbAvl_3",
"ASC4 * 1 + B14_dur * dur_4 + B20_cost * cost_4 + B34_serv * servIdx_4 + B44_stop * stopUs1R1_4")
parl <- c(paste0("ASC", 1:4), paste0("B1", 1:4, "_dur"), "B20_cost", "B53_parkman", "B34_serv",
"B44_stop")
obj <- get_par(parl, eq_d)
ffor <- obj$cheqs0
res <- MNlogitf(ffor, separatenmm=FALSE, transform=FALSE)
ff <- f_create(res$formula, data="data", fixed=1)
```

**get\_npar**

*get\_npar* Get number of parameters or vector of parameters in supplied equations. Extracts the number of parameters used in equations. Parameters are given as par[1], ..., par[n].

## Description

**get\_npar** Get number of parameters or vector of parameters in supplied equations. Extracts the number of parameters used in equations. Parameters are given as par[1], ..., par[n].

## Usage

```
get_npar(x, values = FALSE)
```

## Arguments

- |               |                                                                |
|---------------|----------------------------------------------------------------|
| <b>x</b>      | List of strings.                                               |
| <b>values</b> | if TRUE returns the character vector with parameters (par[i]). |

## Value

Number of parameters or vector with parameters (in form par[i]).

## Examples

```
eq_d <- c("ASC1 * 1 + B11_dur * dur_1" , "ASC2 * 1 + B12_dur * dur_2",
"ASC3 * 1 + B13_dur * dur_3 + B20_cost * cost_3 + B53_parkman * PbAvl_3",
"ASC4 * 1 + B14_dur * dur_4 + B20_cost * cost_4 + B34_serv * servIdx_4 + B44_stop * stopUs1R1_4")
parl <- c("ASC1", "B11_dur", "ASC2", "B12_dur", "ASC3", "B13_dur", "B20_cost", "B53_parkman",
"ASC4", "B14_dur", "B20_cost", "B34_serv", "B44_stop") %>% unique
disc_par <- get_par(parl, eq_d)
get_npar(disc_par$cheqs0)
get_npar(disc_par$cheqs0, values=TRUE)
```

**get\_par***get\_par replaces names of parameters with par[i].***Description**

`get_par` replaces names of parameters with `par[i]`.

**Usage**

```
get_par(par, object)
```

**Arguments**

- `par` Names of parameters, vector of character strings.  
`object` List consisting formulas.

**Value**

A list object consisting of equations for errors with `par[i]` (`cheqs0`), the original list with formulas (`eqlab`), vector of parameters (`par1d`, same as `par`), vector of parameters in form of `par[i]` (`parn`), exogenous variables (`exog`), endogenous variables (`endog`, in case of discrete ""), number of parameters in each equation (`neq_par`).

**Examples**

```
# System of Non-linear Regressions
eq_c <- c("hg ~ exp( h0 + h1*log(tht) + h2*tht^2 + h3*elev)",
         "dg ~ exp( d0 + d1*log(dbh) + d2*hg + d3*cr)")
par_c <- c(paste0("h", 0:3), paste0("d", 0:3))
para_cont <- get_par(par_c, eq_c)
# Indirect utility functions for discrete choice:
eq_d <- c("a1 + p1 * price_1 + p2 * catch_2", "a2 + p1 * price_2 + p2 * catch_2",
          "a3 + p1 * price_3 + p2 * catch_3", "a4 + p1 * price_4 + p2 * catch_4")
par_d <- c(paste0("a", 1:4), paste0("p", 1:2))
disc_par <- get_par(par_d, eq_d)
```

**get\_start***get\_start get starting values for discrete or continuous choice model.***Description**

`get_start` get starting values for discrete or continuous choice model.

**Usage**

```
get_start(
  eq_c = NULL,
  eq_d = NULL,
  data = NULL,
  part = "joint",
  datan = "data",
  fixed_term = FALSE,
  weight_paths = TRUE,
  weight_paths_cont = FALSE,
  data_weight = NULL,
  par_c = NULL,
  par_d = NULL,
  best_method = FALSE,
  startvals = NULL,
  DEoptim_run = FALSE,
  hessian = NULL,
  transform = TRUE,
  MNtypef = "logit",
  pardogit = NULL,
  opt_method = "BFGS",
  numerical_deriv = FALSE
)
```

**Arguments**

eq_c	Continuous equations errors.
eq_d	Discrete equations.
data	data.frame is used in the optimization.
part	Type of estimation: "joint", "cont", "disc".
datan	Name of data.frame used in the optimization.
fixed_term	if TRUE, includes fixed term in log-likelihood.
weight_paths	if TRUE, weights paths of the whole system.
weight_paths_cont	if TRUE, weight paths only in continuous part.
data_weight	data.frame with weights for continuous and discrete equations, same dim as data.
par_c	Names of parameters in continuous equations.
par_d	Names of parameters in discrete equations.
best_method	if TRUE, try all possible optimization methods and choose the one with the smallest likelihood.
startvals	Starting values, can be also NULL.
DEoptim_run	if TRUE, runs DEoptim for the optimization.
hessian	Name of hessian function.

```

transform      if TRUE, quantile transformation is applied.
MNtypef       "dogit" or "logit".
pardogit      "dogit" parameters.
opt_method    optimization method to use.
numerical_deriv
              if TRUE, numerical derivatives are calculated in nmm function

```

### Value

Starting values for discrete or continuous blocks.

### Examples

```

# Example of discrete choice model
data("TravelMode", package = "AER")
eq_d <- c("ASC1 * 1 + B2_t * travel_1 + B3_v * vcost_1" ,
         "ASC2 * 1 + B2_t * travel_2 + B3_v * vcost_2",
         "ASC3 * 1 + B2_t * travel_3 + B3_v * vcost_3",
         "ASC4 * 1 + B2_t * travel_4 + B3_v * vcost_4")
parl <- c(paste0("ASC", 1:4), "B2_t", "B3_v")
obj <- get_par(parl, eq_d)

mode_spec_var <- c("wait", "vcost", "travel", "gcost")
data <- TravelMode
data$wait <- as.numeric(data$wait)
data[data$wait==0,"wait"] <- 0.000001 # add a small number to 0
data$travel <- as.numeric(data$travel)
data[data$travel==0,"travel"] <- 0.000001
data$vcost <- as.numeric(data$vcost)
data[data$vcost==0,"vcost"] <- 0.000001
data <- prepare_data(data, choice="mode", dummy="choice", PeID="individual", WeID="",
type="long", mode_spec_var =mode_spec_var, wc=FALSE)
stv <- get_start(eq_d=eq_d, data=data, datan="data", part="disc", par_d = parl,
transform = FALSE)

#example, system of equations
data("CreditCard", package="AER")
cdat <- CreditCard
cdat$income2 <- cdat$income^2
cdat$d_selfemp <- as.numeric(cdat$selfemp)
eq_c <- c("expenditure ~ b1*age + b2*income + b3*income2",
"income ~ a1*age + a2*d_selfemp + a3*dependents + a4*majordcards")
parl <- c(paste0("b", 1:3), paste0("a", 1:4))
para_cont <- get_par(parl, eq_c)
cheqs0 <- para_cont$cheqs0

stv <- get_start(eq_c = eq_c, data=cdat, datan="cdat", part="cont", par_c=parl)

```

---

grad_hess_eval	<i>grad_hess_eval forms function of gradient and Hessian of log-likelihood produced by f_create.</i>
----------------	------------------------------------------------------------------------------------------------------

---

## Description

grad\_hess\_eval forms function of gradient and Hessian of log-likelihood produced by f\_create.

## Usage

```
grad_hess_eval(mn, parnl, hessian = FALSE, fixed = 0, data = "", cheqs0 = NULL)
```

## Arguments

mn	Expression, can be a list of equations.
parnl	Names of parameters.
hessian	if TRUE, returns hessian function, otherwise gradient.
fixed	Integer, which parameter is fixed to be 0.
data	Name of the data frame with which the function will be evaluated.
cheqs0	If continuous are supplied, include the expressions of errors.

## Value

A function for evaluation of gradient or Hessian.

## Examples

```
eq_d <- c("ASC1 * 1 + B11_dur * dur_1" , "ASC2 * 1 + B12_dur * dur_2",
"ASC3 * 1 + B13_dur * dur_3 + B20_cost * cost_3 + B53_parkman * PbAvl_3",
"ASC4 * 1 + B14_dur * dur_4 + B20_cost * cost_4 + B34_serv * servIdx_4 + B44_stop * stopUs1R1_4")
parl <- c(paste0("ASC", 1:4), paste0("B1", 1:4, "_dur"), "B20_cost", "B53_parkman", "B34_serv",
"B44_stop")
disc_par <- get_par(parl, eq_d)
ffor <- disc_par$cheqs0
parld <- disc_par$parld
res <- MNlogitf(ffor, separtenmm=FALSE, transform=FALSE)
parnl <- paste0("par", 1:length(parld))
gf <- grad_hess_eval (res, parnl, data="data", fixed=1)
hf <- grad_hess_eval (res, parnl, data="data", fixed=1, hessian=TRUE)
```

---

**in2nmm***in2nmm convert some estimation results into nmm object.*

---

## Description

**in2nmm** convert some estimation results into **nmm** object.

## Usage

```
in2nmm(to, new_coef)
```

## Arguments

to	<b>nmm</b> object.
new_coef	New coefficients.

## Value

**nmm** object.

## Examples

```
#example continuous nonlinear
library(systemfit)
data( ppine , package="systemfit")
hg.formula <- hg ~ exp( h0 + h1*log(tht) + h2*tht^2 + h3*elev)
dg.formula <- dg ~ exp( d0 + d1*log(dbh) + d2*hg + d3*cr)
labels <- list( "height.growth", "diameter.growth" )
model <- list( hg.formula, dg.formula )
start.values <- c(h0=-0.5, h1=0.5, h2=-0.001, h3=0.0001,
                  d0=-0.5, d1=0.009, d2=0.25, d3=0.005)
model.sur <- nlsystemfit( "SUR", model, start.values, data=ppine, eqnlables=labels )
eq_c <- as.character(c(hg.formula, dg.formula))
parl <- c(paste0("h", 0:3), paste0("d", 0:3))
res <- nmm(ppine, eq_c=eq_c, start_v=start.values, par_c=parl, eq_type = "cont",
best_method = FALSE)
aa <- in2nmm(res, model.sur$b)
summary(res, new_coef=model.sur$b, type="robust")
summary(aa, type="robust")
summary(res, type="robust")
```

---

<code>logLik.nmm</code>	<i>Log-likelihood(LL) with supplied coefficients.</i>
-------------------------	-------------------------------------------------------

---

## Description

Log-likelihood(LL) with supplied coefficients.

## Usage

```
## S3 method for class 'nmm'
logLik(
  object,
  new_coef = NULL,
  separatenmm = FALSE,
  transform = FALSE,
  methodopt = "NA",
  ...
)
```

## Arguments

<code>object</code>	Object of class <code>nmm</code> .
<code>new_coef</code>	"New" coefficients for which LL should be calculated.
<code>separatenmm</code>	if TRUE, returns separate LL for each equation.
<code>transform</code>	if TRUE, do quantile transformation (normal quantiles).
<code>methodopt</code>	"NA" means that automatic algorithm was used in <code>maxLik</code> , if equal to "BHHH" will return LL for each individual.
...	some methods for this generic function require additional arguments.

## Value

Returns log-likelihood.

## Examples

```
#example continuous nonlinear
library(systemfit)
data( ppine , package="systemfit")
hg.formula <- hg ~ exp( h0 + h1*log(tht) + h2*tht^2 + h3*elev)
dg.formula <- dg ~ exp( d0 + d1*log(dbh) + d2*hg + d3*cr)
labels <- list( "height.growth", "diameter.growth" )
model <- list( hg.formula, dg.formula )
start.values <- c(h0=-0.5, h1=0.5, h2=-0.001, h3=0.0001,
                  d0=-0.5, d1=0.009, d2=0.25, d3=0.005)
model.sur <- nlsystemfit( "SUR", model, start.values, data=ppine, eqnlables=labels )
eq_c <- as.character(c(hg.formula, dg.formula))
parl <- c(paste0("h", 0:3), paste0("d", 0:3))
```

```

res <- nmm(ppine, eq_c=eq_c, start_v=start.values, par_c=par1, eq_type = "cont",
best_method = FALSE)
logLik(res)
logLik(res, new_coef=res$estimate)
logLik(res, new_coef=model.sur$b)
#example discrete
library(mlogit)
data("Fishing", package = "mlogit")
Fish <- mlogit.data(Fishing, varying = c(2:9), shape = "wide", choice = "mode")
## a pure "conditional" model
mres <- summary(mlogit(mode ~ price + catch, data = Fish))
data <- prepare_data(Fish %>% data.frame %>% dplyr::select(-idx),
choice="alt", dummy="mode", PeID="chid", mode_spec_var = c("price", "catch"),
type="long")
eq_d <- c("a1 + p1 * price_1 + p2 * catch_2", "a2 + p1 * price_2 + p2 * catch_2",
          "a3 + p1 * price_3 + p2 * catch_3", "a4 + p1 * price_4 + p2 * catch_4")
par_d <- c(paste0("a", 1:4), paste0("p", 1:2))
res <- nmm(data, eq_d=eq_d, eq_type="disc", fixed_term=FALSE, par_d=par_d,
best_method=FALSE)
logLik(res)
logLik(res, new_coef=res$estimate)
logLik(res, new_coef=mres$coefficients)

```

MAEDtimeExpenditure     *Time-use and expenditure dataset*

## Description

Data gathered in Austria in 2015 according to Mobility-Activity-Expenditure-Dairy (MAED), which reported all trips, activities (time use) and expenditures of 737 persons over a whole week

## Usage

```
data(MAEDtimeExpenditure)
```

## Format

A data frame containing:

**PeID** individual index

**PeGenF** gender of the individual

**PeAge** age in years

**PeEduc** education level

**PeEmploy** employment state

**HhCh** type of household: with children or without children

**w** hourly wage rate, EUR/h

**I** income not realted to work, EUR/week

**T<sub>w</sub>** time spent at work, h/week

**T<sub>f1</sub>** freely chosen activities group 1 (leisure), h/week

**T<sub>f2</sub>** freely chosen activities group 2 (eating, shopping, unspecified), h/week

**T<sub>c</sub>** time spent on committed activities (sleep, domestic work, personal care, travel, education, other), h/week

**E<sub>f1</sub>** freely chosen expenditure group 1 (leisure, accommodation, electronics), EUR/week

**E<sub>f2</sub>** freely chosen expenditure group 2 (clothes), EUR/week

**E<sub>f3</sub>** freely chosen expenditure group 3 (savings), EUR/week

**E<sub>c</sub>** committed expenditures (housing, food, mobility, insurance, other, services, health, furniture, education, financing), EUR/week

**t<sub>a</sub>** total time budget = 168 h/week

**T<sub>d</sub>** time spent on domestic chores, h/week. Td is part of Tc.

## Details

Time and expenditure data correspond to weekly totals. Time in hours and expenditure in EUR.

For more on data collection and description see (Aschauer et al. 2018) and (Aschauer et al. 2019).

A variant of this dataset was used in: (Schmid et al. 2019),(Jokubauskaite et al. 2019) and (Hoessinger et al. 2020).

To get the full dataset please contact r.hoessinger@boku.ac.at.

## References

- Aschauer F, Roesel I, Hoessinger R, Kreis BH, Gerike R (2019). “Time use, mobility, expenditure: An innovative survey design for understanding individual trade-off processes.” *Transportation*, **46**, 307–339. doi: [10.1007/s1111601899619](https://doi.org/10.1007/s1111601899619).
- Aschauer F, Hoessinger R, Axhausen KW, Schmid B, Gerike R (2018). “Implications of survey methods on travel and non-travel activities. A comparison of the Austrian national travel survey and an innovative mobility-activity-expenditure diary (MAED).” *European Journal of Transport and Infrastructure Research*, **18**, 4–35. doi: [10.3929/ethz-b000181072](https://doi.org/10.3929/ethz-b000181072).
- Hoessinger R, Aschauer F, Jara-Diaz S, Jokubauskaite S, Schmid B, Peer S, Axhausen KW, Gerike R (2020). “A joint time-assignment and expenditure-allocation model: value of leisure and value of time assigned to travel for specific population segments.” *Transportation*, **47**, 1439–1475. doi: [10.1007/s1111601910022w](https://doi.org/10.1007/s1111601910022w).
- Jokubauskaite S, Hoessinger R, Aschauer F, Gerike R, Jara-Diaz S, Peer S, Schmid B, Axhausen KW, Leisch F (2019). “Advanced continuous-discrete model for joint time-use expenditure and mode choice estimation.” *Transportation Research Part B Methodological*, **129**, 397–421. doi: [10.1016/j.trb.2019.09.010](https://doi.org/10.1016/j.trb.2019.09.010), <https://www.sciencedirect.com/science/article/pii/S0191261518308245>.
- Schmid B, Jokubauskaite S, Aschauer F, Peer S, Hoessinger R, Gerike R, Jara-Diaz SR, Axhausen KW (2019). “A pooled RP/SP mode, route and destination choice model to investigate mode and user-type effects in the value of travel time savings.” *Transportation Research Part A Policy and Practice*, **124**, 262–294. doi: [10.1016/j.tra.2019.03.001](https://doi.org/10.1016/j.tra.2019.03.001), <https://www.sciencedirect.com/science/article/pii/S0965856418301721>.

## Examples

```
data(MAEDtimeExpenditure)
```

MAEDtravel

*Trip dataset*

## Description

Data gathered in Austria in 2015 according to Mobility-Activity-Expenditure-Dairy (MAED), which reported all trips, activities (time use) and expenditures of 737 persons over a whole week.

## Usage

```
data(MAEDtravel)
```

## Format

A dataframe containing:

**PeID** individual index

**PeGenF** gender of the individual

**PeAge** age in years

**PeEduc** education level

**PeEmploy** employment state

**HhCh** type of household: with children or without children

**WeID** trip index

**choice** chosen mode: 1 - walk, 2 - bike, 3 - car, 4 -public transport (PT)

**dist** trip distance, km

**avl\_1** availability dummy for mode 1, walk

**avl\_2** availability dummy for mode 2, bike

**avl\_3** availability dummy for mode 3, car

**avl\_4** availability dummy for mode 4, PT

**chc\_1** choice dummy for mode 1, walk

**chc\_2** choice dummy for mode 2, bike

**chc\_3** choice dummy for mode 3, car

**chc\_4** choice dummy for mode 4, PT

**cost\_3** cost of car mode

**cost\_4** cost of PT mode

**dur\_1** trip duration with mode 1, minutes

**dur\_2** trip duration with mode 2, minutes

**dur\_3** trip duration with mode 3, minutes

**vdur\_4** in vehicle time mode 4, minutes  
**acc\_4** time to stop or from stop for mode 4, minutes  
**HhCarPark** dummy, car parking at home available  
**JobCarPark** dummy, car parking at workplace available  
**PbAvl\_3** dummy, ar parking restrictions (time and/or cost) in-force at the destination of the trip  
**servIdx\_4** public transport service interval in minutes  
**stopUs1R1\_4** necessary number of changes to reach the destination with public transport  
**leis** trip purpose leisure, effect coding  
**work** trip purpose work, effect coding  
**oth** trip purpose other, effect coding  
**int\_1** inertia for mode 1  
**int\_2** inertia for mode 2  
**int\_3** inertia for mode 3  
**int\_4** inertia for mode 4

## Details

For more on data collection and description see (Aschauer et al. 2019) and (Aschauer et al. 2018). A variant of this dataset was used in: (Schmid et al. 2019), (Jokubauskaite et al. 2019) and (Hoessinger et al. 2020).

To get the full dataset please contact r.hoessinger@boku.ac.at.

Transport modes available: walk, bike, car, public transport (PT). The inertia variable (int\_i) is a dummy, which is equal to one if the mode chosen by a person for a trip at the start of the current tour is the same as the one chosen in the previous tour made for the same purpose, and zero otherwise. Variables for trip purpose (leis, work, oth) were created using the effect coding.

## References

- Aschauer F, Hoessinger R, Axhausen KW, Schmid B, Gerike R (2018). “Implications of survey methods on travel and non-travel activities. A comparison of the Austrian national travel survey and an innovative mobility-activity-expenditure diary (MAED).” *European Journal of Transport and Infrastructure Research*, **18**, 4–35. doi: [10.3929/ethzb000181072](https://doi.org/10.3929/ethzb000181072).
- Aschauer F, Roesel I, Hoessinger R, Kreis BH, Gerike R (2019). “Time use, mobility, expenditure: An innovative survey design for understanding individual trade-off processes.” *Transportation*, **46**, 307–339. doi: [10.1007/s1111601899619](https://doi.org/10.1007/s1111601899619).
- Hoessinger R, Aschauer F, Jara-Diaz S, Jokubauskaite S, Schmid B, Peer S, Axhausen KW, Gerike R (2020). “A joint time-assignment and expenditure-allocation model: value of leisure and value of time assigned to travel for specific population segments.” *Transportation*, **47**, 1439–1475. doi: [10.1007/s1111601910022w](https://doi.org/10.1007/s1111601910022w).
- Jokubauskaite S, Hoessinger R, Aschauer F, Gerike R, Jara-Diaz S, Peer S, Schmid B, Axhausen KW, Leisch F (2019). “Advanced continuous-discrete model for joint time-use expenditure and mode choice estimation.” *Transportation Research Part B Methodological*, **129**, 397–421. doi: [10.1016/j.trb.2019.09.010](https://doi.org/10.1016/j.trb.2019.09.010), <https://www.sciencedirect.com/science/article/pii/S0191261518308245>.

Schmid B, Jokubauskaite S, Aschauer F, Peer S, Hoessinger R, Gerike R, Jara-Diaz SR, Axhausen KW (2019). “A pooled RP/SP mode, route and destination choice model to investigate mode and user-type effects in the value of travel time savings.” *Transportation Research Part A Policy and Practice*, **124**, 262–294. doi: 10.1016/j.tra.2019.03.001, <https://www.sciencedirect.com/science/article/pii/S0965856418301721>.

## Examples

```
data(MAEDtravel)
```

**maxle**

*maxle returns expression of log-likelihood (LL) of joint normal distribution.*

## Description

`maxle` returns expression of log-likelihood (LL) of joint normal distribution.

## Usage

```
maxle(cheqs0, fixed_term = TRUE)
```

## Arguments

- |                         |                                                                                       |
|-------------------------|---------------------------------------------------------------------------------------|
| <code>cheqs0</code>     | Strings defining equations of errors. Systems of Nonlinear Regressions (SNR) variant. |
| <code>fixed_term</code> | if TRUE fixed term $-(k/2)\log(2\pi)$ ( $k$ number of equations) is included          |

## Value

List with LL expressions of joint normal distribution, first element is string with expression for derivative calculations, the second - string for evaluation.

## Examples

```
# normal distribution
eq_c <- c("Tw ~ (((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) -(1 + (PH))) +
sqrt(((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) -(1 + (PH)))^2 - 4 * (1 + (PH) +
(tw)) *(-(PH) * (ta - Tc + 2) + (1 - (tw) * (ta - Tc + 2)) * (2/w - Ec/w)))/(2 * (1 + (PH) +
(tw)))",
"Tf1 ~ (th1) * (ta - (((((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) -(1 + (PH))) +
sqrt(((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) -(1 + (PH)))^2 - 4 * (1 + (PH) +
(tw)) *(-(PH) * (ta - Tc + 2) + (1 - (tw) * (ta - Tc + 2)) * (2/w - Ec/w)))/(2 * (1 + (PH) +
(tw)))) -Tc + 2) - 1",
"Ef1 ~ (ph1)/(PH) * (w * (((((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) -
(1 + (PH))) + sqrt(((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) -(1 + (PH)))^2 - 4 *
(1 + (PH) + (tw)) *(-(PH) * (ta - Tc + 2) + (1 - (tw) * (ta - Tc + 2)) * (2/w - Ec/w)))/(2 *
(1 + (PH) + (tw)))) -Ec + 2) - 1")
parl <- c("tw", "PH", "th1", "ph1")
```

```
para_cont <- get_par(parl, eq_c)
cheqs0 <- para_cont$cheqs0
res <- maxle(cheqs0=cheqs0)
```

**maxle\_p**

*maxle\_p returns expression of partitioned log-likelihood.  
 $f(y_1, y_2, \dots, y_n) = f(y_1)f(y_2|y_1)f(y_3|y_2y_1)\dots f(y_n|y_1..y_{(n-1)})$*

## Description

`maxle_p` returns expression of partitioned log-likelihood.  $f(y_1, y_2, \dots, y_n) = f(y_1)f(y_2|y_1)f(y_3|y_2y_1)\dots f(y_n|y_1..y_{(n-1)})$

## Usage

```
maxle_p(cheqs0, fixed_term = TRUE, version2 = TRUE)
```

## Arguments

<code>cheqs0</code>	Strings defining equations of errors.
<code>fixed_term</code>	if <code>TRUE</code> fixed term $-(k/2)*\log(2*\pi)$ ( $k$ number of equations) is included
<code>version2</code>	another formulation of log-likelihood

## Value

List. First element is expression of joint distribution for derivatives, second for evaluation, third latex, fourth marginal distributions for each variable.

## Examples

```
# joint normal distribution
eq_c <- c("Tw ~ (((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) -(1 + (PH))) +
sqrt(((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) -(1 + (PH)))^2 - 4 * (1 + (PH) +
(tw)) *(-(PH) * (ta - Tc + 2) + (1 - (tw) * (ta - Tc + 2)) * (2/w - Ec/w)))/(2 * (1 + (PH) +
(tw)))",
"Tw1 ~ (th1) * (ta - (((((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) -(1 + (PH))) +
sqrt(((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) -(1 + (PH)))^2 - 4 * (1 + (PH) +
(tw)) *(-(PH) * (ta - Tc + 2) + (1 - (tw) * (ta - Tc + 2)) * (2/w - Ec/w)))/(2 * (1 + (PH) +
(tw)))) -Tc + 2) - 1",
"Ef1 ~ (ph1)/(PH) * (w * (((((PH) + (tw)) * (ta - Tc + 2) + (1 +(tw)) * (Ec/w - 2/w) -(1 +
(PH))) + sqrt(((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) -(1 + (PH)))^2 - 4 *
(1 + (PH) + (tw)) *(-(PH) * (ta - Tc + 2) + (1 - (tw) * (ta - Tc + 2)) * (2/w - Ec/w)))/(2 *
(1 + (PH) + (tw)))) -Ec + 2) - 1")
parl <- c("tw", "PH", "th1", "ph1")
para_cont <- get_par(parl, eq_c)
cheqs0 <- para_cont$cheqs0
res <- maxle_p(cheqs0=cheqs0)
```

---

<code>mlsem</code>	<i>mlsem</i> returns expression of log-likelihood for joint normal distribution, for maximum likelihood (ML), Simultaneous Equations Models (SEM) variant.
--------------------	------------------------------------------------------------------------------------------------------------------------------------------------------------

---

**Description**

`mlsem` returns expression of log-likelihood for joint normal distribution, for maximum likelihood (ML), Simultaneous Equations Models (SEM) variant.

**Usage**

```
mlsem(cheqs0, fixed_term = TRUE)
```

**Arguments**

- |                         |                                                                              |
|-------------------------|------------------------------------------------------------------------------|
| <code>cheqs0</code>     | Strings defining equations of errors.                                        |
| <code>fixed_term</code> | if TRUE fixed term $-(k/2)\log(2\pi)$ ( $k$ number of equations) is included |

**Value**

List with LL expressions of joint normal distribution, first element is string with expression for derivative calculations, the second - string for evaluation.

**Examples**

```
# normal distribution
eq_c <- c("Tw ~ (((PH) + (tw)) * (ta - Tc) + Ec/w * (1 + (tw)) + sqrt((Ec/w * (1 + (tw)) +
(ta - Tc) * ((PH) + (tw)))^2 - 4 * Ec/w * (ta - Tc) * (tw) * (1 + (PH) + (tw))))/(2 * (1 + (PH))
+ (tw)))",
"Tf1 ~ (th1) * (Tw - Tc)",
"Ef1 ~ (ph1)/(PH) * (w * Tw - Ec)"
parl <- c("tw", "PH", "th1", "ph1")
para_cont <- get_par(parl, eq_c)
cheqs0 <- para_cont$cheqs0
res <- mlsem(cheqs0, fixed_term=FALSE)
```

**MNlogitf**

*MNlogitf or MNdogitf* returns log-likelihood(LL) expression for discrete equations of "logit" or "dogit" model.

**Description**

`MNlogitf` or `MNdogitf` returns log-likelihood(LL) expression for discrete equations of "logit" or "dogit" model.

**Usage**

```
MNlogitf(ffor, transform = FALSE, separatenmm = FALSE, weight_disc = FALSE)

MNdogitf(
  ffor,
  transform = FALSE,
  separatenmm = FALSE,
  weight_disc = FALSE,
  ppardogit = NULL
)
```

**Arguments**

ffor	Discrete choice equations.
transform	if TRUE, quantile transformation (normal) is applied.
separatenmm	if TRUE, equation specific LL is calculated
weight_disc	if TRUE, equations will include equation specific weights.
ppardogit	"dogit" parameters only used in MNdogitf

**Value**

formula	simplified LL for each equation or joint
probs	probability expression unsimplified
expr	LL string simplified
formulat	unsimplified log(P), only if transform is TRUE
probt	unsimplified P with quantile transformation qnorm(ifelse(P)), only if transform is TRUE
expreteso	unsimplified pnorm((qnorm(P)-mean)/sd), only if transform is TRUE
probte	qnorm(P)
tval	If quantile transformation is applied
sume	Denominator of logit probability

**Functions**

- **MNlogitf**: returns log-likelihood(LL) expression for discrete equations of "logit" model.
- **MNdogitf**: returns log-likelihood(LL) expression for discrete equations of "dogit" model.

**Examples**

```
eq_d <- c("ASC1 * 1 + B11_dur * dur_1" , "ASC2 * 1 + B12_dur * dur_2",
"ASC3 * 1 + B13_dur * dur_3 + B20_cost * cost_3 + B53_parkman * PbAvl_3",
"ASC4 * 1 + B14_dur * dur_4 + B20_cost * cost_4 + B34_serv * servIdx_4 + B44_stop * stopUs1R1_4")
par1 <- c(paste0("ASC", 1:4), paste0("B1", 1:4, "_dur"), "B20_cost", "B53_parkman", "B34_serv",
"B44_stop")
disc_par <- get_par(par1, eq_d)
```

```
ffor <- disc_par$cheqs0
res_l <- MNlogitf(ffor, separatenmm=FALSE, transform=FALSE)
res_d <- MNdogitf(ffor, separatenmm=FALSE, transform=FALSE)
```

nmm

*Maximum likelihood estimation of nonlinear multivariate models (NMM).*

## Description

nmm, nmm\_sigma and summary are the main functions used for the estimation of NMM.

- nmm - Maximum likelihood estimation of nonlinear multivariate models (NMM)
- nmm\_sigma - Optimizes the covariance matrix
- summary - returns summary of nmm object with "normal", "robust" or "clustered" standard errors. With option new\_coef one can supply new coefficients and test their significance.

## Usage

```
nmm(
  data,
  eq_type = c("joint", "cont", "disc"),
  eq_d = NULL,
  eq_c = NULL,
  par_c = NULL,
  par_d = NULL,
  start_v = NULL,
  check_hess = TRUE,
  corr1 = TRUE,
  weight_paths = TRUE,
  weight_paths_cont = FALSE,
  data_weight = NULL,
  estimate = TRUE,
  fixed_term = FALSE,
  best_method = FALSE,
  DEoptim_run = FALSE,
  hessian = "joint_hess",
  print_out = FALSE,
  diff_hessian = FALSE,
  bayesian_random = FALSE,
  DEoptim_run_main = FALSE,
  deconst = 2,
  numerical_deriv = FALSE,
  best_method4start = FALSE,
  eqsys = "sur",
  miterlim = 10000,
  opt_method = "BFGS",
```

```

try_last_DOptim = TRUE,
transform = NULL,
MNtypef = "logit",
nmm_object = NULL
)

## S3 method for class 'nmm'
summary(object, type = "normal", new_coef = NULL, ...)

nmm_sigma(
object,
methodopt = "BFGS",
try_1good = TRUE,
try_DOptim = FALSE,
try_diff_method = FALSE,
trace = FALSE,
estimate = FALSE
)

```

## Arguments

data	data.frame used in the optimization.
eq_type	Possible options "joint", "cont", "disc".
eq_d	Discrete equations.
eq_c	Continuous equations.
par_c	Parameters from continuous equations.
par_d	Parameters from discrete equations.
start_v	Starting values for optimization. If NULL, starting values are found by get_start function.
check_hess	If TRUE, check the Hessian.
corr1	If TRUE, correlation between blocks (continuous and discrete).
weight_paths	If TRUE, weight according to the number of choices made by individual i will be added to the whole system.
weight_paths_cont	If TRUE, if only to continuous part should be weighted.
data_weight	Data weight matrix.
estimate	If TRUE, estimation is performed.
fixed_term	If TRUE, includes fixed term to continuous equation block.
best_method	If TRUE, all optimizers are checked.
DOptim_run	If TRUE, runs DOptim in generation of starting values.
hessian	String, name of the Hessian function.
print_out	If TRUE, prints out log-likelihood for each equation.
diff_hessian	If TRUE, for changing hessian and gradient with weights.

<b>bayesian_random</b>	If TRUE, than par[1] is changed to par[,1] to be used for optimization of random parameters in Bayesian estimation.
<b>DEoptim_run_main</b>	If TRUE, run DEoptim in the main optimization.
<b>deconst</b>	absolute value of lower and upper bound in DEoptim optimization.
<b>numerical_deriv</b>	If TRUE, uses numerical derivative instead of the analytical.
<b>best_method4start</b>	If TRUE, all optimizers are checked for starting values.
<b>eqsys</b>	"sur" or "sem".
<b>miterlim</b>	Number many iterations passed to maxLik function
<b>opt_method</b>	optimization method for maxLik.
<b>try_last_DEoptim</b>	If TRUE, in case of error in maxLik should DEoptim be run.
<b>transform</b>	if TRUE, quantile transformation is applied to discrete equations.
<b>MNtypef</b>	estimate "logit", or "dogit"
<b>nmm_object</b>	nmm object created by nmm function.
<b>object</b>	nmm object, for summary and nmm_sigma
<b>type</b>	Type of standard errors c("robust", "clustered", "normal"), for summary
<b>new_coef</b>	New coefficients that will be tested, for summary
<b>...</b>	additional arguments affecting the summary produced, for summary
<b>methodopt</b>	optimizer from maxLik package, for nmm_sigma
<b>try_1good</b>	If TRUE, stops then first good values are found, for nmm_sigma
<b>try_DEoptim</b>	If TRUE, uses DEoptim for optimization, then maxLik optimizers produce errors, for nmm_sigma
<b>try_diff_method</b>	If TRUE, stops then first good values with Hessian check are found, for nmm_sigma
<b>trace</b>	If TRUE, trace of DEoptim is printed, for nmm_sigma

### Value

<b>nmm</b>	returns nmm object with estimated parameters, functions, and data.
<b>nmm_sigma</b>	returns estimated parameters, functions, data.
<b>summary</b>	returns summary of nmm object.

### Examples

```
# estimation of System of Nonlinear Equations based on example from 'systemfit'
library(systemfit)
data( ppine , package="systemfit")
hg.formula <- hg ~ exp( h0 + h1*log(tht) + h2*tht^2 + h3*elev)
dg.formula <- dg ~ exp( d0 + d1*log(dbh) + d2*hg + d3*cr)
```

```

labels <- list( "height.growth", "diameter.growth" )
model <- list( hg.formula, dg.formula )
start.values <- c(h0=-0.5, h1=0.5, h2=-0.001, h3=0.0001,
                  d0=-0.5, d1=0.009, d2=0.25, d3=0.005)
model.sur <- nlsystemfit( "SUR", model, start.values, data=ppine, eqnlabels=labels )
eq_c <- as.character(c(hg.formula, dg.formula))
parl <- c(paste0("h", 0:3), paste0("d", 0:3))
res <- nmm(ppine, eq_c=eq_c, par_c=parl, start_v = start.values,
eq_type = "cont", best_method = FALSE, numerical_deriv=TRUE)
summary(res)
res_sigma_cont <- nmm_sigma(res, estimate=TRUE) # Estimation of the Variance-Covariance matrix
summary(res_sigma_cont)

#example discrete choice
library(mlogit)
data("Fishing", package = "mlogit")
Fish <- mlogit.data(Fishing, varying = c(2:9), shape = "wide", choice = "mode")
## a pure "conditional" model
mres <- summary(mlogit(mode ~ price + catch, data = Fish))
data <- prepare_data(Fish %>% data.frame %>% dplyr::select(-idx),
choice="alt", dummy="mode", PeID="chid", mode_spec_var = c("price", "catch"),
type="long")
eq_d <- c("a1 + p1 * price_1 + p2 * catch_2", "a2 + p1 * price_2 + p2 * catch_2",
          "a3 + p1 * price_3 + p2 * catch_3", "a4 + p1 * price_4 + p2 * catch_4")
par_d <- c(paste0("a", 1:4), paste0("p", 1:2))
res <- nmm(data, eq_d=eq_d, par_d = par_d, eq_type="disc", fixed_term=FALSE,
best_method=FALSE)
summary(res)

# joint estimation mockup example
data(dataM)
dataMp <- dataM %>% data.frame %>% prepare_data(. , choice="DR_Course",
PeID = "Student")
eq_c <- c("PlcmtScore ~ exp(a0 + a1 * PSATM + a2 * Rank + a3 * Size)",
"ACTM ~ exp(c0 + c1 * GPAadj)")
par_c <- c(paste0("a", 0:3), paste0("c", 0:1))
eq_d <- c("ASC1" ,
"ASC2 + b1_2 * SATM + b2_2 * PlcmtScore",
"ASC3 + b1_3 * SATM + b2_3 * PlcmtScore")
par_d <- c(paste0("ASC", 1:3), paste0("b", rep(1:2, rep(2,2)), "_", 2:3))

nmm_joint_res <- nmm(dataMp, eq_type = "joint", eq_d = eq_d,
par_d = par_d, eq_c = eq_c, par_c = par_c,
start_v = c(a0=3.394, a1=0.001, a2=-0.001, a3=0, c0=3.583, c1=-0.008,
ASC2=-1.452, ASC3=3.047, b1_2=0.145, b1_3=0.102, b2_2=-0.133, b2_3=-0.168))
summary(nmm_joint_res)

```

## Description

`prepare_data` prepare data for the estimation.

## Usage

```
prepare_data(
  data,
  choice = "",
  dummy = "",
  PeID = "",
  WeID = "",
  type = "",
  mode_spec_var = "",
  avl = TRUE,
  chc = TRUE,
  wc = TRUE,
  wd = TRUE,
  nc = 0,
  weights = NULL,
  weight_paths = FALSE,
  weight_paths_cont = FALSE,
  mode_factors = NULL
)
```

## Arguments

<code>data</code>	<code>data.frame</code>
<code>choice</code>	Name of variable with modes.
<code>dummy</code>	Name of variable indicating, if the mode was chosen.
<code>PeID</code>	Name of variable with individual identification numbers.
<code>WeID</code>	Name of variable with trip identification.
<code>type</code>	Type of data. If "long", then modifications are done.
<code>mode_spec_var</code>	Used if format "long", mode specific variables.
<code>avl</code>	if TRUE, includes dummies for mode availability.
<code>chc</code>	if TRUE, includes dummies for choice of mode.
<code>wc</code>	if TRUE, creates weights 1 for continuous equations.
<code>wd</code>	if TRUE, creates weights 1 for discrete equations.
<code>nc</code>	Integer, number of continuous equations.
<code>weights</code>	Data matrix with weights, column names have to be \$wc_i\$(continuous), \$wd_i\$(discrete).
<code>weight_paths</code>	if TRUE, weight according to number of trips per person, discrete part.
<code>weight_paths_cont</code>	if TRUE, weight continuous part.
<code>mode_factors</code>	if choice is not factor or numeric, this is important to supply.

**Value**

`data.frame` used for modeling.

**Examples**

```
data("TravelMode", package = "AER")
mode_spec_var <- c("wait", "vcost", "travel", "gcost")
res <- prepare_data(TravelMode, choice="mode", dummy="choice", PeID="individual", WeID="",
type="long", mode_spec_var = mode_spec_var, nc=3)
```

pseudoR

*pseudo R<sup>2</sup>***Description**

Calculates pseudo R<sup>2</sup> for discrete choice part.

**Usage**

```
pseudoR(
  x,
  which = c("all", "McFadden", "adjMcFadden", "Cox&Snell", "Nagelkerke"),
  only_total = FALSE
)
```

**Arguments**

<code>x</code>	Fitted <code>nmm</code> model.
<code>which</code>	which pseudo R <sup>2</sup> to calculate, options are: "all", "McFadden", "adjMcFadden", "Cox&Snell", "Nagelkerke".
<code>only_total</code>	If TRUE, compute R <sup>2</sup> only for the whole sample.

**Value**

`matrix` with goodness of fit measures

**Examples**

```
library(mlogit)
data("Fishing", package = "mlogit")
Fish <- mlogit.data(Fishing, varying = c(2:9), shape = "wide", choice = "mode")
## a pure "conditional" model
mres <- summary(mlogit(mode ~ price + catch, data = Fish))
data <- prepare_data(Fish %>% data.frame %>% dplyr::select(-idx),
choice="alt", dummy="mode", PeID="chid", mode_spec_var = c("price", "catch"),
type="long")
eq_d <- c("a1 + p1 * price_1 + p2 * catch_2", "a2 + p1 * price_2 + p2 * catch_2",
         "a3 + p1 * price_3 + p2 * catch_3", "a4 + p1 * price_4 + p2 * catch_4")
```

```

par_d <- c(paste0("a", 1:4), paste0("p", 1:2))
res <- nmm(data, eq_d=eq_d, par_d = par_d, eq_type="disc", fixed_term=FALSE,
best_method=FALSE)
pseudoR(res, which = c("McFadden"))
ncf <- c(mres$coefficients)
names(ncf) <- par_d[-1]
mress <- in2nmm(res, new_coef = ncf)
pseudoR(mress, which = c("McFadden"))
pseudoR(mress)

```

replace\_par

replace\_par replaces text with other text.

## Description

replace\_par replaces text with other text.

## Usage

```
replace_par(iter, repdat, biogu)
```

## Arguments

iter	Integer, which line of repdat is used, if 1 iteratively all will be replaced, if >1 only the i-th parameter.
repdat	data.frame with columns "old" and "new"
biogu	string which will be modified

## Value

Modified string.

## Examples

```

eq_c <- c("Tw ~ (((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) -(1 + (PH))) +
sqrt(((PH) + (tw)) * (ta - Tc + 2) + (1 +(tw)) * (Ec/w - 2/w) - (1 + (PH)))^2 - 4 *
(1 + (PH) + (tw)) *(-(PH) * (ta - Tc + 2) + (1 - (tw) * (ta - Tc + 2)) * (2/w -Ec/w)))/(2 *
(1 + (PH) + (tw))),",
"Tf1 ~ (th1) * (ta - (((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) - (1 + (PH))) +
sqrt(((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) - (1 + (PH)))^2 - 4 *(1 + (PH) +
(tw)) * (- (PH) * (ta - Tc + 2) + (1 - (tw) * (ta - Tc + 2)) * (2/w - Ec/w)))/(2 * (1 + (PH) +
(tw))) -Tc + 2) - 1",
"Ef1 ~ (ph1)/(PH) * (w * (((((PH) + (tw)) * (ta - Tc + 2) + (1 +(tw)) * (Ec/w - 2/w) -
(1 + (PH))) + sqrt(((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) - (1 + (PH)))^2 -4 *
(1 + (PH) + (tw)) * (- (PH) * (ta - Tc + 2) + (1 - (tw) * (ta - Tc + 2)) * (2/w - Ec/w)))/(2 *
(1 + (PH) + (tw))) -Ec + 2) - 1")
parl <- c("tw", "PH", "th1", "ph1")
parll <- paste0("par[", 1:length(parl), "]")
repdat <- data.frame(old=parl, new=parll)
replace_par(1, repdat, eq_c)

```

---

replace_par_wrap	replace_par_wrap <i>replace text with other text, wrapper of replace_par.</i>
------------------	-------------------------------------------------------------------------------

---

## Description

`replace_par_wrap` replace text with other text, wrapper of `replace_par`.

## Usage

```
replace_par_wrap(repdat, obj)
```

## Arguments

<code>repdat</code>	data.frame with columns "old" and "new"
<code>obj</code>	character string or vector of strings that will be modified.

## Value

Modified string/vector.

## Examples

```
eq_c <- c("Tw ~ (((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) -(1 + (PH))) + sqrt(((PH) + (tw)) * (ta - Tc + 2) + (1 +(tw)) * (Ec/w - 2/w) - (1 + (PH)))^2 - 4 * (1 + (PH) + (tw)) *(-PH) * (ta - Tc + 2) + (1 - (tw) * (ta - Tc + 2)) * (2/w -Ec/w))))/(2 * (1 + (PH) + (tw))),  
"Tf1 ~ (th1) * (ta - (((((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) - (1 + (PH))) + sqrt(((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) - (1 + (PH)))^2 - 4 *(1 + (PH) + (tw)) * (-PH) * (ta - Tc + 2) + (1 - (tw) * (ta - Tc + 2)) * (2/w - Ec/w))))/(2 * (1 + (PH) + (tw))) -Tc + 2) - 1",  
"Ef1 ~ (ph1)/(PH) * (w * (((((PH) + (tw)) * (ta - Tc + 2) + (1 +(tw)) * (Ec/w - 2/w) - (1 + (PH))) + sqrt(((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) - (1 + (PH)))^2 - 4 *(1 + (PH) + (tw)) * (-PH) * (ta - Tc + 2) + (1 - (tw) * (ta - Tc + 2)) * (2/w - Ec/w))))/(2 * (1 + (PH) + (tw))) -Ec + 2) - 1")  
parl <- c("tw", "PH", "th1", "ph1")  
parll <- paste0("par[", 1:length(parl), "]")  
repdat <- data.frame(old=parl, new=parll)  
replace_par(2, repdat, eq_c)  
replace_par_wrap(repdat, eq_c)
```

---

stats_function	<i>Helper functions</i>
----------------	-------------------------

---

### Description

Produce function that calculate estimates of endogenous variables from the continuous block and probabilities from discrete part.

### Usage

```
stats_function(eq_c = NULL, eq_d = NULL, par_c = NULL, par_d = NULL, fixed = 0)
```

### Arguments

eq_c	continuous equations
eq_d	discrete equations
par_c	parameters from cont. eq
par_d	parameters from disc. eq
fixed	index of fixed parameter

### Value

Returns functions.

### Examples

```
eq_d <- c("ASC1 * 1 + B11_dur * dur_1" , "ASC2 * 1 + B12_dur * dur_2",
"ASC3 * 1 + B13_dur * dur_3")
eq_c <- c("Tw ~ tw*w + ph1*Tc", "Tf1 ~ (1+w)^tw + ph1^3*Tc")
parl <- c("tw", "ph1")
par_d <- c(paste0("ASC", 1:3), paste0("B1", 1:3, "_dur"))
stfunc <- stats_function(eq_c, eq_d, parl, par_d, fixed=3)
data <- matrix(runif(1000, min=0.001, max=50), ncol=8)
data <- data.frame(data)
names(data) <- c("dur_1", "dur_2", "dur_3", "w", "Tc", "avl_1", "avl_2", "avl_3")
parv <- c(0.5, 1, 1.5, 2, 1, -0.3, 0.2, -0.8)
methodopt <- "BHHH"
separatenmm <- TRUE
env <- environment()
fnames <- c("prob_func", "cont_func")
sapply(1:(length(stfunc)-1), function(x)assign(fnames[x], stfunc[[x]], envir=env))
eval(parse(text=paste0("environment(", fnames[1:(length(stfunc)-1)], ") <- env")))
probs <- prob_func(parv)
apply(probs, 2, mean)
cont <- cont_func(parv)
apply(cont, 2, mean)
```

---

**string2formula**

*string2formula add square brackets to expressions. Reverse of formula2string. Convert par[2] <- par2, sigma[2] <- sigma\_2\_, sigma[2,3] <- sigma\_2x2*

---

### Description

**string2formula** add square brackets to expressions. Reverse of formula2string. Convert par[2] <- par2, sigma[2] <- sigma\_2\_, sigma[2,3] <- sigma\_2x2

### Usage

**string2formula(x)**

### Arguments

**x** String without square brackets.

### Value

String with square brackets.

### Examples

```
xx <- "par[1]*3+par[2]*par+rho1[1,2]+sigma1[2]"
xm <- formula2string(xx)
string2formula(xm)
```

---

**wxMaxima**

*wxMaxima does symbolic computation in 'Maxima' Requires installation of Maxima software.*

---

### Description

**wxMaxima** does symbolic computation in 'Maxima' Requires installation of Maxima software.

### Usage

**wxMaxima(obj, out, tex = FALSE)**

### Arguments

**obj** Lines without printed output, that will be evaluated in wxMaxima.

**out** Lines with printed output, is end results of the function.

**tex** if TRUE TeX expression of the printed output.

## Value

List of character strings.

## Examples

```
#components(determinant, Sigma inverse, argument for exponent) for joint normal distribution
eq_c <- c("Tw ~ (((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) -(1 + (PH))) +
sqrt(((PH) + (tw)) * (ta - Tc + 2) + (1 +(tw)) * (Ec/w - 2/w) - (1 + (PH)))^2 - 4 * (1 + (PH) +
(tw)) *(- (PH) * (ta - Tc + 2) + (1 - (tw) * (ta - Tc + 2)) * (2/w - Ec/w)))/(2 * (1 + (PH) +
(tw)))",
"Th1 ~ (th1) * (ta - (((((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) - (1 + (PH))) +
sqrt(((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) - (1 + (PH)))^2 - 4 * (1 + (PH) +
(tw)) * (- (PH) * (ta - Tc + 2) + (1 - (tw) * (ta - Tc + 2)) * (2/w - Ec/w)))/(2 * (1 + (PH) +
(tw)))) -Tc + 2) - 1",
"Ef1 ~ (ph1)/(PH) * (w * (((((PH) + (tw)) * (ta - Tc + 2) + (1 +(tw)) * (Ec/w - 2/w) -
(1 + (PH))) + sqrt(((PH) + (tw)) * (ta - Tc + 2) + (1 + (tw)) * (Ec/w - 2/w) - (1 + (PH)))^2 - 4 *
(1 + (PH) + (tw)) * (- (PH) * (ta - Tc + 2) + (1 - (tw) * (ta - Tc + 2)) * (2/w - Ec/w)))/(2 *
(1 + (PH) + (tw))) -Ec + 2) - 1")
parl <- c("tw", "PH", "th1", "ph1")
para_cont <- get_par(parl, eq_c)
cheqs0 <- para_cont$cheqs0
npar <- get_npar(cheqs0)
neq <- length(cheqs0)
sdv <- rep(NA, neq)
mv <- rep(NA, neq)
sigma <- expand.grid(1:neq, 1:neq)%>%apply(., 1, function(x)paste0(x, collapse=','))%>%
paste0('sigma', '[', . , ']')%>%sort
sigma <- matrix(sigma, neq, neq, byrow = TRUE)
sigma[lower.tri(sigma)] <- sort(sigma[upper.tri(sigma)])
#create Y
y <- paste0('eps[, 1:neq, ]')
ypy <- paste0('[', y, ']', collapse = ',')
ypy <- paste0('Y : matrix(', ypy, ')')
#sigma
spy <- paste0(sapply(1:neq, function(i) paste0('[', paste0(sigma[i,], collapse = ' ', '), ']')),
collapse = ',')
spy <- paste0('Sigma : matrix(', spy, ')')
dets <- 'dets : ratsimp(determinant(Sigma))'
invs <- 'invs : ratsimp(invert(Sigma))'
expt <- 'expt : ratsimp(transpose(Y) . invs . Y)'
obj <- c(ypy, spy, dets, invs, expt)
#grind function in Maxima returns an object that can be mathematically evaluated
out <- c('print(new)', 'grind(dets)', 'print(new)', 'grind(invs)', 'print(new)', 'grind(expt)')
## Not run:
rez <- wxMaxima(obj, out)

## End(Not run)
```

# Index

\* datasets  
  dat4cond\_mean\_cov\_expr, 7  
  dataM, 8  
  datmaxle, 9  
  datmlsem, 10  
  expr\_ll\_norm, 11  
  expr\_ll\_norm\_v2, 12  
  MAEDtimeExpenditure, 22  
  MAEDtravel, 24  
  
  add\_variable, 3  
  addInter, 2  
  AICc, 4  
  
  BIC.nmm (AICc), 4  
  
  cond\_expr, 5  
  cont\_stats, 6  
  convert\_attr2exp, 7  
  
  dat4cond\_mean\_cov\_expr, 7  
  dataM, 8  
  datmaxle, 9  
  datmlsem, 10  
  diagnostics, 10  
  
  expr\_ll\_norm, 11  
  expr\_ll\_norm\_v2, 12  
  extract\_attr\_deriv, 12  
  
  f\_create, 14  
  formula2string, 13  
  
  get\_npar, 15  
  get\_par, 16  
  get\_start, 16  
  grad\_hess\_eval, 19  
  
  in2nmm, 20  
  
  logLik.nmm, 21  
  
  MAEDtimeExpenditure, 22  
  MAEDtravel, 24  
  maxle, 26  
  maxle\_p, 27  
  mlsem, 28  
  MNdogitf (MNlogitf), 28  
  MNlogitf, 28  
  
  nmm, 30  
  nmm\_sigma (nmm), 30  
  
  prepare\_data, 33  
  pseudoR, 35  
  
  replace\_par, 36  
  replace\_par\_wrap, 37  
  
  stats\_function, 38  
  string2formula, 39  
  summary.nmm (nmm), 30  
  
  wxMaxima, 39