

Package ‘nestedLogit’

May 28, 2026

Title Nested Dichotomy Logistic Regression Models

Version 0.4.2

Date 2026-05-27

Description Provides functions for specifying and fitting nested dichotomy logistic regression models for a multi-category response and methods for summarising and plotting those models. Nested dichotomies are statistically independent, and hence provide an additive decomposition of tests for the overall ‘polytomous’ response. When the dichotomies make sense substantively, this method can be a simpler alternative to the standard ‘multinomial’ logistic model which compares response categories to a reference level. See: J. Fox (2016), ‘Applied Regression Analysis and Generalized Linear Models’, 3rd Ed., ISBN 1452205663.

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URL <https://github.com/friendly/nestedLogit>,
<https://friendly.github.io/nestedLogit/>

BugReports <https://github.com/friendly/nestedLogit/issues>

Depends R (>= 4.1.0)

Imports broom, car, dplyr, effects, graphics, grDevices, stats,
stringr, tibble, scales

Suggests AER, carData, equatiomatic, DescTools, geomtextpath, ggplot2,
ggeffects, here, insight, lobstr, knitr, nnet, parameters,
performance, rmarkdown, see, spelling, testthat, tidyr, MASS,
VGAM, mlogit, vcd

VignetteBuilder knitr, rmarkdown

Encoding UTF-8

Language en-US

LazyData TRUE

Config/roxygen2/version 8.0.0

NeedsCompilation no

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Repository CRAN

Date/Publication 2026-05-28 14:10:01 UTC

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as.data.frame.predictNestedLogit
Convert a Predicted Objects to a data.frame

Description

These functions provide simple ways to convert the results of `predict.nestedLogit` to a data frame in a consistent format for plotting and other actions.

Usage

```
## S3 method for class 'predictNestedLogit'
as.data.frame(x, row.names = NULL, optional = FALSE, ...)
```

Arguments

x	a "predictNestedLogit" object
row.names	row.names for result (for conformity with generic; not currently used)
optional	logical. If TRUE, setting row names and converting column names (to syntactic names: see make.names is optional
...	other arguments (unused)

Value

- For `predict(..., model="nested")` (the default), returns a data frame containing the values of predictors along with the columns `response`, `p`, `se.p`, `logit`, `se.logit`.
- For `predict(..., model="dichotomies")`, returns a data frame containing the values of predictors along with the columns `response`, `logit`, and `se.logit`.

Examples

```
data("Women1f", package = "carData")
comparisons <- logits(work=dichotomy("not.work", c("parttime", "fulltime")),
  full=dichotomy("parttime", "fulltime"))

wlf.nested <- nestedLogit(partic ~ hincome + children,
  dichotomies = comparisons,
  data=Women1f)
# get predicted values for a grid of `hincome` and `children`
new <- expand.grid(hincome=seq(0, 45, length=10),
  children=c("absent", "present"))

pred.nested <- predict(wlf.nested, new)
plotdata <- as.data.frame(pred.nested)
str(plotdata)

# Predicted logit values for the dichotomies
pred.dichot <- predict(wlf.nested, new, model = "dichotomies")
plotlogit <- as.data.frame(pred.dichot)
str(plotlogit)
```

as.tree

Display the Tree Structure of Nested Dichotomies

Description

Display the nested structure of a "dichotomies" or "continuationDichotomies" object as a 2-D ASCII tree diagram showing how the response categories are split at each level of the nesting.

Usage

```
as.tree(x, ...)

## S3 method for class 'dichotomies'
as.tree(x, response = NULL, lobstr = FALSE, ...)

## S3 method for class 'continuationDichotomies'
as.tree(x, response = NULL, lobstr = FALSE, ...)
```

Arguments

x	A "dichotomies" or "continuationDichotomies" object.
...	additional arguments (currently unused).
response	Optional character string giving the name of the response variable, used as the root label of the tree. If NULL (default), the root is labeled "(response)".
lobstr	Logical. If FALSE (default), renders a 2-D ASCII tree with / and \ branch connectors, with each node centered above its two children. If TRUE, builds a nested list representation of the tree and renders it via tree , which must be installed.

Details

The flat list of dichotomies is reconstructed into a binary tree by matching each dichotomy's domain (the union of its two sides) to the multi-level groups produced by earlier splits. Branch labels are taken from the named arguments to [dichotomy](#) when present, and are otherwise generated automatically as {level1, level2, ...}.

Value

Invisibly returns x; called for its side effect of printing.

See Also

[logits](#), [continuationLogits](#), [print.dichotomies](#) Other conversions: [as.matrix.dichotomies](#), [as.character.dichotomies](#)

Examples

```
## Womenlf: named group on one branch
comparisons <- logits(work = dichotomy("not.work",
                                     working = c("parttime", "fulltime")),
                    full = dichotomy("parttime", "fulltime"))
as.tree(comparisons, response = "partic")

## GSS: continuation logits for ordered education levels
cont <- continuationLogits(c("<highschool", "highschool",
                           "college", "graduate"))
as.tree(cont, response = "degree")

## gators data: Food choice
```

```
# create dichotomies
gators.dichots <- logits(d1=dichotomy("Other", c("Fish", "Invertebrates")),
                        d2=dichotomy("Fish", "Invertebrates"))
as.tree(gators.dichots, response = "Food")
as.tree(gators.dichots, response = "Food", lobstr = TRUE)
```

broomMethods

Broom Related Methods

Description

These functions give compact summaries of a "nestedLogit" object

`glance` Construct a single row summaries for the dichotomies "nestedLogit" model.

`tidy` Summarizes the terms in "nestedLogit" model.

Usage

```
## S3 method for class 'nestedLogit'
glance(x, ...)

## S3 method for class 'nestedLogit'
tidy(x, ...)
```

Arguments

`x` an object of class "nestedLogit".

`...` arguments to be passed down.

Value

- `glance` returns a tibble containing one row of fit statistics for each dichotomy, labeled response. See [glance](#) for details.
- `tidy` returns a tibble containing coefficient estimates and test statistics for the combinations of response and term. See [tidy](#) for details.

See Also

[nestedMethods](#), [glance](#), [tidy](#)

Examples

```

data("Women1f", package = "carData")
m <- nestedLogit(partic ~ hincome + children,
                 dichotomies = logits(work=dichotomy("not.work",
                                                    working=c("parttime", "fulltime")),
                                     full=dichotomy("parttime", "fulltime")),
                 data=Women1f)

# one-line summaries
broom::glance(m)
# coefficients and tests
broom::tidy(m)

```

Effect.nestedLogit *Effect Displays for Nested Logit Models*

Description

Computes effects (in the sense of the **effects** package—see, in particular, [Effect](#))—for "nestedLogit" models, which then can be used with other functions in the **effects** package, for example, [predictorEffects](#) and to produce effect plots.

Usage

```

## S3 method for class 'nestedLogit'
Effect(
  focal.predictors,
  mod,
  confidence.level = 0.95,
  fixed.predictors = NULL,
  ...
)

```

Arguments

focal.predictors a character vector of the names of one or more of the predictors in the model, for which the effect display should be computed.

mod a "nestedLogit" model object.

confidence.level for point-wise confidence bands around the effects (the default is 0.95).

fixed.predictors controls the values at which other predictors are fixed; see [Effect](#) for details; if NULL (the default), numeric predictors are set to their means, factors to their distribution in the data.

... optional arguments to be passed to the [Effect](#) method for binary logit models (fit by the [glm](#) function).

Value

an object of class "effpoly" (see [Effect](#)).

Author(s)

John Fox

References

John Fox and Sanford Weisberg (2019). *An R Companion to Applied Regression*, 3rd Edition. Sage, Thousand Oaks, CA.

John Fox, Sanford Weisberg (2018). Visualizing Fit and Lack of Fit in Complex Regression Models with Predictor Effect Plots and Partial Residuals. *Journal of Statistical Software*, 87(9), 1-27.

See Also

[Effect](#), [plot.effpoly](#), [predictorEffects](#)

Examples

```
data("Womenlf", package = "carData")
comparisons <- logits(work=dichotomy("not.work",
                                     working=c("parttime", "fulltime")),
                    full=dichotomy("parttime", "fulltime"))
m <- nestedLogit(partic ~ hincome + children,
                dichotomies = comparisons,
                data=Womenlf)
peff.women <- effects::predictorEffects(m)
plot(peff.women)
plot(peff.women, axes=list(y=list(style="stacked")))
summary(peff.women)

dichots <- logits(AB_CD = dichotomy(c("A", "B"), c("C", "D")),
                A_B = dichotomy("A", "B"),
                C_D = dichotomy("C", "D"))
m.health <- nestedLogit(product4 ~ age + gender*household + position_level,
                    dichotomies = dichots, data = HealthInsurance)
eff.gen.hh <- effects::Effect(c("gender", "household"), m.health,
                            xlevels=list(household=0:7))
eff.gen.hh
plot(eff.gen.hh, axes=list(x=list(rug=FALSE)))
plot(eff.gen.hh, axes=list(x=list(rug=FALSE),
                            y=list(style="stacked")))
```

 extract_eq.nestedLogit

Extract Equations for a "nestedLogit" Model

Description

Provides an `equatiomatic::extract_eq()` method for "nestedLogit" objects to render the models as LaTeX equations in documents.

It extracts the LaTeX equation for each of the binary logit sub-models comprising a nested logit model, substituting the dichotomy name for the internal `..y` placeholder used by `nestedLogit()`.

Usage

```
extract_eq.nestedLogit(model, submodel = NULL, ...)
```

Arguments

<code>model</code>	a "nestedLogit" object.
<code>submodel</code>	character; name of a single dichotomy (e.g. "work"). When supplied, a single "equation" object is returned and renders directly in a knitr chunk. When omitted, equations for all dichotomies are returned as a named list of class "nestedLogit_equations".
<code>...</code>	additional arguments passed to <code>equatiomatic::extract_eq()</code> . For example, use <code>use_coefs = TRUE</code> to display fitted coefficients rather than symbols; <code>wrap = TRUE</code> to split the RHS across multiple lines; <code>greek_colors</code> , <code>var_colors</code> , etc. to color the symbols.

Details

Internally, `nestedLogit()` fits each sub-model on a temporary response column named `..y`, so `equatiomatic::extract_eq()` would render the response as `..y` rather than the dichotomy name. This method replaces `..y` with the name of each dichotomy.

Because `_` is the subscript operator in LaTeX, any underscores in dichotomy names are replaced by `.` in the displayed equation (e.g., a dichotomy named `fish_inv` appears as `fish.inv`). This does not affect the names of the returned list, which retain the original R names.

Value

When `submodel` is `NULL`, an object of class `c("nestedLogit_equations", "list")` containing one "equation" per dichotomy, named by the dichotomy. Individual equations can be extracted with `$name` or by passing `submodel`. When `submodel` is a dichotomy name, the single corresponding "equation" object is returned and renders directly in a knitr chunk.

See Also

[equatiomatic::extract_eq\(\)](#)

Examples

```

data("Womenlf", package = "carData")
wlf.nested <- nestedLogit(partic ~ hincome + children,
  dichotomies = logits(work = dichotomy("not.work",
    working = c("parttime", "fulltime")),
    full = dichotomy("parttime", "fulltime")),
  data = Womenlf)
if (requireNamespace("equatiomatic", quietly = TRUE)) {
  equatiomatic::extract_eq(wlf.nested)
}

```

gators

*Alligator Food Choice***Description**

Agresti (1996, p. 207) gives this data on 59 alligators sampled from a lake in Florida. It has the length of the alligator in meters and the primary food type found in the alligator's stomach. The food type was classified into three categories: "Fish", "Invertebrates", and "Other".

Of interest is whether or not the length of an alligator is associated with the primary food type. Does knowing the length of an alligator give us some indication about its primary food type? If so, how is length associated with the choice of food type?

Usage

```
data("gators", package = "nestedLogit")
```

Format

A data frame with 59 rows and 2 columns.

food Primary food type found in the alligator's stomach, a factor with levels "Other", "Fish", and "Invertebrates".

length Length of the alligator in meters, a numeric vector.

Source

Agresti, A. (1996). *An Introduction to Categorical Data Analysis*. Wiley.

References

An example using this from <https://data.library.virginia.edu/getting-started-with-multinomial-logit-models/>

See Also

[nestedLogit](#).

Examples

```

data(gators)
table(gators$food)
# average length of gators by food
with(gators, tapply(length, food, mean))

# create dichotomies
dichot <- logits(d1=dichotomy("Other", c("Fish", "Invertebrates")),
               d2=dichotomy("Fish", "Invertebrates"))

gators.nested <- nestedLogit(food ~ length,
                             dichotomies = dichot,
                             data = gators)

car::Anova(gators.nested)

# use the plot method
plot(gators.nested, x.var = "length")

```

GSS

Data From the U.S. General Social Survey 1972-2016

Description

This data set is drawn from the U.S. General Social Survey (GSS) for years between 1972 and 2016.

Usage

```
data("GSS", package = "nestedLogit")
```

Format

A data frame with 44091 rows and 3 columns.

parentdeg A factor representing parents' attained level of education (highest "degree" obtained), recording the higher of mother's and father's education, with levels "<highschool", "highschool", "college", and "graduate".

degree The respondent's level of education, a factor with the same levels as parentdeg.

year The year of the survey, between 1972 and 2016.

Source

General Social Survey, NORC, The University of Chicago <https://www.norc.org/Research/Projects/Pages/general-social-survey.aspx>.

See Also

[nestedLogit](#).

Examples

```

round(100*with(GSS, prop.table(table(degree, parentdeg), 2)))
m.GSS <- nestedLogit(degree ~ parentdeg*year,
                    continuationLogits(c("<highschool", "highschool",
                                         "college", "graduate")),
                    data=GSS)

car::Anova(m.GSS)
summary(m.GSS)

# plot fitted probabilities
plot(m.GSS, x.var = "year",
      others = list(parentdeg = "<highschool"),
      lty = 1,
      label = TRUE)
plot(m.GSS, x.var = "year",
      others = list(parentdeg = "graduate"),
      lty = 1,
      label = TRUE)

```

HealthInsurance

Choice of Health Insurance Product

Description

A company recently introduced a new health insurance provider for its employees. At the beginning of the year the employees had to choose one of three (or four) different health plan products from this provider to best suit their needs.

This dataset was modified from its original source (McNulty, 2022) for the present purposes by adding a fourth choice, sampled randomly from the original three.

Usage

```
data("HealthInsurance", package = "nestedLogit")
```

Format

A data frame with 1448 rows and 7 columns.

product Choice among three products, a factor with levels "A", "B", and "C".

product4 Choice among four products, a factor with levels "A", "B", "C", and "D".

age The age of the individual, in years.

household The number of people living with the individual in the same household.

position_level Position level in the company at the time the choice was made, where 1 is the lowest level and 5 is the highest, a numeric vector.

gender The gender of the individual, a factor with levels "Female" and "Male".

absent The number of days the individual was absent from work in the year prior to the choice,

Source

Originally taken from McNulty, K. (2022). *Handbook of Regression Modeling in People Analytics*, https://peopleanalytics-regression-book.org/data/health_insurance.csv.

See Also

[nestedLogit](#).

Examples

```
lbinary <- logits(AB_CD = dichotomy(c("A", "B"), c("C", "D")),
                A_B   = dichotomy("A", "B"),
                C_D   = dichotomy("C", "D"))
as.matrix(lbinary)
health.nested <- nestedLogit(product4 ~ age + gender * household + position_level,
                             dichotomies = lbinary, data = HealthInsurance)
coef(car::Anova(health.nested))
```

models

Extract Binary Logit Models from a nestedLogit Object

Description

Nested logit models represent an overall models for a polytomous response (>2 categories) by a set of binary logit models corresponding to nested dichotomies among the response categories. `models` is used to extract "glm" objects representing binary logit models from a "nestedLogit" object.

Usage

```
models(model, select, as.list = FALSE)

## S3 method for class 'nestedLogit'
models(model, select, as.list = FALSE)
```

Arguments

<code>model</code>	a "nestedLogit" model.
<code>select</code>	a numeric or character vector giving the number(s) or names(s) of one or more binary logit models to be extracted from <code>model</code> ; if absent, a list of all of the binary logits models in <code>model</code> is returned.
<code>as.list</code>	if TRUE (the default is FALSE) and one binary logit model is selected, return the "glm" object in a one-element named list; otherwise a single model is returned directly as a "glm" object; when more than one binary logit model is selected, the corresponding "glm" objects are <i>always</i> returned as a named list.

Value

model returns either a single "glm" object (see [glm](#)) or a list of "glm" objects, each representing a binary logit model.

Examples

```
data("Women1f", package = "carData")
comparisons <- logits(work=dichotomy("not.work",
                                     working=c("parttime", "fulltime")),
                    full=dichotomy("parttime", "fulltime"))
m <- nestedLogit(partic ~ hincome + children,
                dichotomies = comparisons,
                data=Women1f)

# extract both submodels, as a list
models(m, c("work", "full"))

# extract the binomial logit model for working vs. non-working
m_work <- models(m, "work")

# use that to plot residuals
plot(density(residuals(m_work)))

# or plot that model -- gives the 'regression quartet' for a glm()
op <- par(mfrow = c(2,2))
plot(m_work)
par(op)
```

 nestedHypotheses

Hypothesis-Testing and Related Methods for "nestedLogit" Objects

Description

Various methods for testing hypotheses about nested logit models.

Anova Calculates type-II or type-III analysis-of-variance tables for "nestedLogit" objects; see [Anova](#) in the **car** package.

anova Computes sequential analysis of variance (or deviance) tables for one or more fitted "nestedLogit" objects; see [anova](#).

linearHypothesis Computes Wald tests for linear hypotheses; see [linearHypothesis](#) in the **car** package.

logLik Returns the log-likelihood and degrees of freedom for the nested-dichotomies model. (and through it [AIC](#) and [BIC](#) model-comparison statistics).

Usage

```

## S3 method for class 'nestedLogit'
Anova(mod, ...)

## S3 method for class 'Anova.nestedLogit'
print(x, ...)

## S3 method for class 'nestedLogit'
linearHypothesis(model, ...)

## S3 method for class 'nestedLogit'
anova(object, object2, ...)

## S3 method for class 'anova.nestedLogit'
print(x, ...)

## S3 method for class 'nestedLogit'
logLik(object, ...)

```

Arguments

... arguments to be passed down. In the case of `linearHypothesis`, the second argument is typically the `hypothesis.matrix`. See the Details section of [linearHypothesis](#). In the case of `anova`, additional sequential "nestedLogit" models.

`x`, `object`, `object2`, `mod`, `model`
in most cases, an object of class "nestedLogit".

Value

- The `Anova` and `anova` methods return objects of class "Anova.nestedLogit" and "anova.nestedLogit", respectively, each of which contains a list of "anova" objects (see [anova](#)) and is usually printed.
- The `linearHypothesis` method is called for its side effect, printing the result of linear hypothesis tests, and invisibly returns NULL.
- The `logLik` method returns an object of class "logLik" (see [logLik](#)).

Author(s)

John Fox

See Also

[Anova](#), [anova](#), [linearHypothesis](#), [logLik](#), [AIC](#), [BIC](#)

Examples

```
# define continuation dichotomies for level of education
cont.dichots <- continuationLogits(c("<highschool",
                                     "highschool",
                                     "college",
                                     "graduate"))

# fit a nested model for the GSS data examining education degree in relation to parent & year
m <- nestedLogit(degree ~ parentdeg + year,
                 cont.dichots,
                 data=GSS)

# Anova and anova tests
car::Anova(m) # type-II (partial) tests

anova(update(m, . ~ . - year), m) # model comparison

# Wald test
car::linearHypothesis(m, c("parentdeghighschool", "parentdegcollege",
                           "parentdeggraduate"))

# log-likelihood, AIC, and BIC
logLik(m)
AIC(m)
BIC(m)
```

 nestedLogit

Binary Logit Models for Nested Dichotomies

Description

Fit a related set of binary logit models via the `glm` function to nested dichotomies, comprising a model for the polytomy. A polytomous response with m categories can be analyzed using $m - 1$ binary logit comparisons. When these comparisons are nested, the $m - 1$ sub-models are statistically independent. Therefore, the likelihood chi-square statistics for the sub-models are additive and give overall tests for a model for the polytomy. This method was introduced by Fienberg (1980), and subsequently illustrated by Fox (2016) and Friendly & Meyer (2016).

`dichotomy` and `logits` are helper functions to construct the dichotomies.

`continuationLogits` constructs a set of $m - 1$ logit comparisons, called continuation logits, for an ordered response. With $m = 4$ levels, say, A, B, C, D, considered low to high: The first contrasts B, C, D against A. The second ignores A and contrasts C, D against B. The second ignores A, B and contrasts D against C.

Usage

```
nestedLogit(formula, dichotomies, data, subset = NULL, contrasts = NULL, ...)

logits(...)
```

```
dichotomy(...)
```

```
continuationLogits(levels, names, prefix = "above_")
```

Arguments

formula	a model formula with the polytomous response on the left-hand side and the usual linear-model-like specification on the right-hand side.
dichotomies	specification of the logits for the nested dichotomies, constructed by the logits and dichotomy functions, or continuationLogits. Alternatively, the dichotomies can be specified as a nested (i.e., recursive) list, the elements of which can be given optional names. See Details.
data	a data frame with the data for the model; unlike in most statistical modeling functions, the data argument is required. Cases with NAs in any of the variables appearing in the model formula will be removed with a Note message.
subset	a character string specifying an expression to fit the model to a subset of the data; the default, NULL, uses the full data set.
contrasts	an optional list of contrast specification for specific factors in the model; see lm for details.
...	for nestedLogit, optional named arguments to be passed to glm ; for logits, definitions of the nested logits—with each named argument specifying a dichotomy; for dichotomy, two character vectors giving the levels defining the dichotomy; the vectors can optionally be named.
levels	A character vector of set of levels of the variables or a number specifying the numbers of levels (in which case, uppercase letters will be use for the levels).
names	Names to be assigned to the dichotomies; if absent, names will be generated from the levels.
prefix	a character string (default: "above_") used as a prefix to the names of the continuation dichotomies.

Details

A *dichotomy* for a categorical variable is a comparison of one subset of levels against another subset. A set of dichotomies is *nested*, if after an initial dichotomy, all subsequent ones are *within* the groups of levels lumped together in earlier ones. Nested dichotomies correspond to a binary tree of the successive divisions.

For example, for a 3-level response, a first dichotomy could be {A}, {B, C} and then the second one would be just {B}, {C}. Note that in the second dichotomy, observations with response A are treated as NA.

The function `dichotomy` constructs a *single* dichotomy in the required form, which is a list of length 2 containing two character vectors giving the levels defining the dichotomy. The function `logits` is used to create the set of dichotomies for a response factor. Alternatively, the nested dichotomies can be specified more compactly as a nested (i.e., recursive) list with optionally named elements; for example, `list(air="plane", ground=list(public=list("train", "bus"), private="car"))`.

The function `continuationLogits` provides a convenient way to generate all dichotomies for an ordered response. For an ordered response with $m = 4$ levels, say, A, B, C, D, considered low

to high: The dichotomy first contrasts B, C, D against A. The second ignores A and contrasts C, D against B. The second ignores A, B and contrasts D against C.

Value

nestedLogit returns an object of class "nestedLogit" containing the following elements:

- `models`, a named list of (normally) $m - 1$ "glm" objects, each a binary logit model for one of the $m - 1$ nested dichotomies representing the m -level response.
- `formula`, the model formula for the nested logit models.
- `dichotomies`, the "dichotomies" object defining the nested dichotomies for the model.
- `data.name`, the name of the data set to which the model is fit, of class "name".
- `data`, the data set to which the model is fit.
- `subset`, a character representation of the subset argument or "NULL" if the argument isn't specified.
- `contrasts`, the contrasts argument or NULL if the argument isn't specified.
- `contrasts.print` a character representation of the contrasts argument or "NULL" if the argument isn't specified.

`logits` and `continuationLogits` return objects of class "dichotomies" and `c("continuationDichotomies" "dichotomies")`, respectively, which are two-elements lists, each element containing a list of two character vectors representing a dichotomy. `dichotomy` returns a list of two character vectors representing a dichotomy.

Author(s)

John Fox

References

- S. Fienberg (1980). *The Analysis of Cross-Classified Categorical Data*, 2nd Edition, MIT Press, Section 6.6.
- J. Fox (2016), *Applied Linear Regression and Generalized Linear Models*, 3rd Edition, Sage, Section 14.2.2.
- J. Fox and S. Weisberg (2011), *An R Companion to Applied Regression*, 2nd Edition, Sage, Section 5.8.
- M. Friendly and D. Meyers (2016), *Discrete Data Analysis with R*, CRC Press, Section 8.2.

See Also

[nestedMethods](#)

Examples

```

data("Women1f", package = "carData")

#' Use `logits()` and `dichotomy()` to specify the comparisons of interest
comparisons <- logits(work=dichotomy("not.work",
                                     working=c("parttime", "fulltime")),
                    full=dichotomy("parttime", "fulltime"))
print(comparisons)

m <- nestedLogit(partic ~ hincome + children,
                dichotomies = comparisons,
                data=Women1f)
print(summary(m))
print(car::Anova(m))
coef(m)

# equivalent;
nestedLogit(partic ~ hincome + children,
            dichotomies = list("not.work",
                               working=list("parttime", "fulltime")),
            data=Women1f)

# get predicted values
new <- expand.grid(hincome=seq(0, 45, length=10),
                  children=c("absent", "present"))
pred.nested <- predict(m, new)

# plot
op <- par(mfcol=c(1, 2), mar=c(4, 4, 3, 1) + 0.1)
plot(m, "hincome", list(children="absent"),
     xlab="Husband's Income", legend=FALSE)
plot(m, "hincome", list(children="present"),
     xlab="Husband's Income")
par(op)

continuationLogits(c("none", "gradeschool", "highschool", "college"))
continuationLogits(4)

```

Description

Various methods for processing "nestedLogit" and related objects. Most of these are the standard methods for a model-fitting function.

`coef`, `vcov` Return the coefficients and their variance-covariance matrix respectively.

`update` Re-fit a "nestedLogit" model with a change in any of the formula, dichotomies, data, subset, or contrasts, arguments.

`summary` Summarize a "nestedLogit" model, giving the summary for each binary logit model in the nested dichotomies.

`print` Print the model or a summary of the model.

`as.matrix`, `as.character`, `as.dichotomies` Coerce dichotomy-related objects to matrices, character vectors, and dichotomies objects.

Usage

```
## S3 method for class 'nestedLogit'
print(x, ...)

## S3 method for class 'nestedLogit'
summary(object, ...)

## S3 method for class 'summary.nestedLogit'
print(x, ...)

## S3 method for class 'dichotomies'
print(x, ...)

## S3 method for class 'nestedLogit'
coef(object, as.matrix = TRUE, ...)

## S3 method for class 'nestedLogit'
vcov(object, as.matrix = FALSE, ...)

## S3 method for class 'nestedLogit'
update(object, formula, dichotomies, data, subset, contrasts, ...)

## S3 method for class 'dichotomies'
as.matrix(x, ...)

## S3 method for class 'dichotomies'
as.character(x, ...)

## S3 method for class 'continuationDichotomies'
as.matrix(x, ...)

as.dichotomies(x, ...)

## S3 method for class 'matrix'
as.dichotomies(x, ...)
```

Arguments

`x`, `object` in most cases, an object of class "nestedLogit".

...	arguments to be passed down.
as.matrix	if TRUE (the default for coef) return coefficients as a matrix with one column for each nested dichotomy, or coefficient covariances as a matrix with one row and column for each combination of dichotomies and coefficients; if FALSE (the default for vcov), return a list of coefficients or coefficient covariances with one element for each dichotomy.
formula	optional updated model formula.
dichotomies	optional updated dichotomies object.
data	optional updated data argument
subset	optional updated subset argument.
contrasts	optional updated contrasts argument.

Value

- The coef and vcov methods return either matrices or lists of regression coefficients and their covariances, respectively.
- The update method returns an object of class "nestedLogit" (see [nestedLogit](#)) derived from the original nested-logit model.
- The summary method returns an object of class "summary.nestedLogit", which is a list of summaries of the [glm](#) objects that comprise the nested-dichotomies model; the object is normally printed.
- The methods for as.matrix, as.character, and as.dichotomies coerce various objects to matrices, character vectors, and dichotomies objects.
- The various print methods invisibly return their x arguments.

Author(s)

John Fox and Michael Friendly

See Also

[nestedLogit](#), [predict.nestedLogit](#), [plot.nestedLogit](#), [glance.nestedLogit](#), [tidy.nestedLogit](#)

Examples

```
# define continuation dichotomies for level of education
cont.dichots <- continuationLogits(c("<highschool",
                                   "highschool",
                                   "college",
                                   "graduate"))

# Show dichotomies in various forms
print(cont.dichots)
as.matrix(cont.dichots)
as.character(cont.dichots)

# fit a nested model for the GSS data examining education degree in relation to parent & year
```

```

m <- nestedLogit(degree ~ parentdeg + year,
                 cont.dichots,
                 data=GSS)

coef(m) # coefficient estimates
sqrt(diag(vcov(m, as.matrix=TRUE))) # standard errors
print(m)
summary(m)

```

plot.nestedLogit *Plotting Nested Logit Models*

Description

A `plot` method for "nestedLogit" objects produced by the `nestedLogit` function. Fitted probabilities under the model, or the corresponding logits are plotted for each level of the polytomous response variable, with one of the explanatory variables on the horizontal axis and other explanatory variables fixed to particular values. By default, a 95% pointwise confidence envelope is added to the plot.

Usage

```

## S3 method for class 'nestedLogit'
plot(
  x,
  x.var,
  others,
  n.x.values = 100L,
  scale = c("prob", "logit"),
  xlab = x.var,
  ylab = NULL,
  main,
  cex.main = 1,
  digits.main = getOption("digits") - 2L,
  font.main = 1L,
  pch = 1L:length(response.levels),
  lwd = 3,
  lty = 1L:length(response.levels),
  col = (scales::hue_pal())(length(response.levels)),
  legend = TRUE,
  legend.inset = 0.01,
  legend.location = "topleft",
  legend.bty = "n",
  conf.level = 0.95,
  conf.alpha = 0.25,
  label = FALSE,
  label.x = "max",

```

```

    label.cex = 1.25,
    label.col = col,
    ...
)

```

Arguments

<code>x</code>	an object of "nestedLogit" produced by <code>nestedLogit</code> .
<code>x.var</code>	quoted name of the variable to appear on the x-axis; if omitted, the first predictor in the model is used.
<code>others</code>	a named list of values for the other variables in the model, that is, other than <code>x.var</code> ; if any other predictor is omitted, it is set to an arbitrary value—the mean for a numeric predictor or the first level or value of a factor, character, or logical predictor; only one value may be specified for each variable in <code>others</code> .
<code>n.x.values</code>	the number of evenly spaced values of <code>x.var</code> at which to evaluate fitted probabilities to be plotted (default 100).
<code>scale</code>	character string; "prob" (the default) plots fitted probabilities on the y-axis; "logit" plots fitted log odds (logits), i.e., $\log(p/(1-p))$.
<code>xlab</code>	label for the x-axis (defaults to the value of <code>x.var</code>).
<code>ylab</code>	label for the y-axis (defaults to "Fitted Probability" when <code>scale = "prob"</code> and "Fitted Log Odds" when <code>scale = "logit"</code>).
<code>main</code>	main title for the graph (if missing, constructed from the variables and values in <code>others</code>).
<code>cex.main</code>	size of main title (see <code>par</code>).
<code>digits.main</code>	number of digits to retain when rounding values for the main title.
<code>font.main</code>	font for main title (see <code>par</code>).
<code>pch</code>	plotting characters (see <code>par</code>).
<code>lwd</code>	line width (see <code>par</code>).
<code>lty</code>	line types (see <code>par</code>).
<code>col</code>	line colors for the response levels (see <code>par</code>).
<code>legend</code>	if TRUE (the default), add a legend for the response levels to the graph. Ignored when <code>label = TRUE</code> .
<code>legend.inset</code>	default 0.01 (see <code>legend</code>).
<code>legend.location</code>	position of the legend (default "topleft", see <code>legend</code>).
<code>legend.bty</code>	the type of box to be drawn around the legend. The allowed values are "o" (the default) and "n".
<code>conf.level</code>	the level for pointwise confidence envelopes around the predicted values; the default is 0.95. If NULL, the confidence envelopes are suppressed.
<code>conf.alpha</code>	the opacity of the confidence envelopes; the default is 0.3.
<code>label</code>	if TRUE, label the curves directly instead of using a legend. Default is FALSE.

label.x	where to place the label for each curve. Either a single string, "min" or "max" (applied to all curves), or a character vector of length equal to the number of response categories with each element being "min" or "max", e.g. label.x = c("min", "max", "max"). "min" places the label at the left end of the curve; "max" (the default) places it at the right end.
label.cex	character expansion factor for direct labels; default 1.25.
label.col	colors for direct labels; defaults to col, so labels match their curves. Supply a vector of length equal to the number of response categories to use different colors for the labels.
...	arguments to be passed to matplot .

Value

NULL Used for its side-effect of producing a plot

Author(s)

John Fox, Michael Friendly

See Also

[nestedLogit](#), [matplot](#)

Examples

```
data("Womenlf", package = "carData")
m <- nestedLogit(partic ~ hincome + children,
                 logits(work=dichotomy("not.work", c("parttime", "fulltime")),
                       full=dichotomy("parttime", "fulltime")),
                 data=Womenlf)
plot(m, legend.location="top")

op <- par(mfcol=c(1, 2), mar=c(4, 4, 3, 1) + 0.1)
plot(m, "hincome", list(children="absent"),
     xlab="Husband's Income", legend=FALSE)
plot(m, "hincome", list(children="present"),
     xlab="Husband's Income")
par(op)

# Plot on the logit (log-odds) scale
plot(m, "hincome", list(children="absent"), scale = "logit",
     xlab = "Husband's Income")

# Gators example: direct curve labels instead of a legend
data("gators", package = "nestedLogit")
gators.nested <- nestedLogit(food ~ length,
                             logits(d1 = dichotomy("Other", c("Fish", "Invertebrates")),
                                   d2 = dichotomy("Fish", "Invertebrates")),
                             data = gators)

# All labels at the right end (default)
```

```

plot(gators.nested, x.var = "length", label = TRUE,
     xlab = "Alligator length (m)")

# Mixed placement: Other and Invertebrates labeled at left, Fish at right
# (food levels are: "Other", "Fish", "Invertebrates")
plot(gators.nested, x.var = "length", label = TRUE,
     label.x = c("min", "max", "min"),
     xlab = "Alligator length (m)")

```

predict.nestedLogit *Predicted Probabilities and Logits for "nestedLogit" Models*

Description

The predict and fitted methods compute predicted values from a fitted "nestedLogit" model. The confint method computes point-wise confidence limits for predicted response-category probabilities or logits.

predict, fitted Compute predicted response-category probabilities (or predicted logits for each binary logit model in the dichotomies) from a fitted "nestedLogit" model.

confint Compute point-wise confidence limits for predicted response-category probabilities or logits.

print **methods** Print predicted probabilities, logits, and their standard errors, with control over how many rows to display.

Usage

```

## S3 method for class 'nestedLogit'
predict(object, newdata, model = c("nested", "dichotomies"), ...)

## S3 method for class 'predictNestedLogit'
print(x, n = min(10L, nrow(x$p)), ...)

## S3 method for class 'predictNestedLogit'
confint(
  object,
  parm = c("prob", "logit"),
  level = 0.95,
  conf.limits.logit = TRUE,
  ...
)

## S3 method for class 'predictDichotomies'
print(x, n = 10L, ...)

## S3 method for class 'nestedLogit'
fitted(object, model = c("nested", "dichotomies"), ...)

```

Arguments

<code>object</code>	a fitted object of class "nestedLogit"; for <code>confint</code> , an object of class "predictNestedLogit".
<code>newdata</code>	a data frame containing combinations of values of the predictors at which fitted probabilities (or other quantities) are to be computed. If missing, the original data are used.
<code>model</code>	either "nested" (the default), in which case fitted probabilities under the nested logit model are returned, or "dichotomies", in which case <code>predict.glm</code> is invoked for each binary logit model fit to the nested dichotomies and a named list of the results is returned.
<code>...</code>	arguments to be passed down.
<code>x</code>	an object of class "predictNestedLogit" or "predictDichotomies".
<code>n</code>	an integer or "all" to control how many rows are printed for each of the predicted values.
<code>parm</code>	one of "prob" or "logit", indicating whether to generate confidence intervals for probabilities or logits of the responses.
<code>level</code>	confidence level, a number between 0 and 1; default is 0.95.
<code>conf.limits.logit</code>	logical; when <code>parm = "prob"</code> , if TRUE (the default), confidence limits are computed on the logit scale and back-transformed to probabilities; if FALSE, Wald-type limits are computed directly on the probability scale.

Details

The `predict` method provides predicted values for two representations of the model. `model = "nested"` (the default) gives the fitted probabilities for each of the response categories, along with the corresponding logits and standard errors of each. `model = "dichotomies"` gives the fitted log odds for each of the binary logit models in the nested dichotomies.

The `fitted` method (with no `newdata`) is equivalent to `predict` applied to the original data used to fit the model.

For the `confint` method with `parm = "prob"`, setting `conf.limits.logit = TRUE` (the default) computes confidence limits on the logit scale and back-transforms them to probabilities, which ensures that the limits lie in $[0, 1]$. Setting `conf.limits.logit = FALSE` computes Wald-type confidence intervals directly on the probability scale, which may extend outside $[0, 1]$.

Value

- The `predict` and `fitted` methods return an object of class "predictNestedLogit" (when `model = "nested"`) or "predictDichotomies" (when `model = "dichotomies"`). A "predictNestedLogit" object is a list containing:
 - `p` a data frame of predicted probabilities, with one column per response category.
 - `logit` a data frame of predicted logits.
 - `se.p` a data frame of standard errors of predicted probabilities.
 - `se.logit` a data frame of standard errors of predicted logits.
 - `.data` the `newdata` data frame, if supplied.

A "predictDichotomies" object is a named list of data frames, one per dichotomy, each produced by `predict.glm`.

- The `confint` method returns a data frame of point estimates and confidence limits.
- The various print methods invisibly return their `x` arguments.

Author(s)

John Fox and Michael Friendly

See Also

[nestedLogit](#), [nestedMethods](#), [plot.nestedLogit](#), [as.data.frame.predictNestedLogit](#)

Examples

```
# define continuation dichotomies for level of education
cont.dichots <- continuationLogits(c("<highschool",
                                   "highschool",
                                   "college",
                                   "graduate"))

# fit a nested model for the GSS data
m <- nestedLogit(degree ~ parentdeg + year,
                 cont.dichots,
                 data=GSS)

# predicted probabilities for first few cases
predict(m)

# predicted probabilities at specific values of predictors
new <- expand.grid(parentdeg=c("<highschool", "highschool",
                              "college", "graduate"),
                  year=c(1972, 2016))

fit <- predict(m, newdata=new)
cbind(new, fit)

# use fitted() -- equivalent to predict() on the original data
f <- fitted(m)

# predicted logits for each dichotomy
pred.dichot <- predict(m, newdata=new, model="dichotomies")
pred.dichot

# confidence intervals for predicted probabilities
fit.ci <- confint(fit)
head(fit.ci)

# confidence intervals for predicted logits
fit.ci.logit <- confint(fit, parm="logit")
head(fit.ci.logit)
```

Description

Computes pseudo-R² and related fit measures for a "nestedLogit" object and related models for a polytomous response. For the "nestedLogit" case, the result shows one row per binary logit sub-model (dichotomy) and an additional "Combined" row for the overall polytomous model.

Usage

```
RSQ(x, ...)  
  
## S3 method for class 'nestedLogit'  
RSQ(  
  x,  
  which = c("McFadden", "CoxSnell", "Nagelkerke"),  
  include = "AIC",  
  digits = 3L,  
  ...  
)  
  
## S3 method for class 'RSQ.nestedLogit'  
print(x, digits = attr(x, "digits"), ...)  
  
## S3 method for class 'multinom'  
RSQ(  
  x,  
  which = c("McFadden", "CoxSnell", "Nagelkerke"),  
  include = "AIC",  
  digits = 3L,  
  ...  
)  
  
## S3 method for class 'RSQ.multinom'  
print(x, digits = attr(x, "digits"), ...)  
  
## S3 method for class 'polr'  
RSQ(  
  x,  
  which = c("McFadden", "CoxSnell", "Nagelkerke"),  
  include = "AIC",  
  digits = 3L,  
  ...  
)  
  
## S3 method for class 'RSQ.polr'
```

```
print(x, digits = attr(x, "digits"), ...)
```

Arguments

<code>x</code>	a "nestedLogit" object.
<code>...</code>	currently unused.
<code>which</code>	character vector naming the pseudo- R^2 measures to compute. Any subset of <code>c("McFadden", "McFaddenAdj", "CoxSnell", "Nagelkerke", "Tjur")</code> , or "ALL" to include all of them. Default: <code>c("McFadden", "CoxSnell", "Nagelkerke")</code> .
<code>include</code>	character vector of additional columns to append to the result. Any subset of <code>c("AIC", "BIC", "n")</code> , where "n" adds the number of observations used for each row, or "ALL" to include all of them. Default: "AIC".
<code>digits</code>	integer; number of decimal places used when printing (default 3L).

Details

RSQ is implemented as an S3 generic with methods for "nestedLogit", as well as `nnet::multinom()`, and `MASS::polr()` objects, which are other methods for modeling a polytomous response variable.

In contrast to standard, Gaussian linear models, where R^2 has a uniformly simple interpretation as "variance accounted for" by the model, and with different, yet *equivalent* computational formulas, there is no single commonly accepted measure for logistic regression models for a binary response or a dichotomy among outcomes.

The following measures are available via the `which` argument:

"McFadden" $1 - L/L_0$, where L is the fitted model log-likelihood and L_0 that of the null (intercept-only) model (McFadden, 1979). Values of 0.1–0.3 indicate a reasonable fit in logistic regression.

"McFaddenAdj" $1 - (L - k)/L_0$, where k is the number of non-intercept parameters; penalises model complexity (Hosmer & Lemeshow, 2000).

"CoxSnell" $1 - \exp(2(L_0 - L)/n)$; bounded strictly below 1 for discrete outcomes (Cox & Snell, 1989).

"Nagelkerke" Cox-Snell divided by its theoretical maximum, rescaling to $[0, 1]$ (Nagelkerke, 1991).

"Tjur" Mean fitted value for $y = 1$ minus mean fitted value for $y = 0$; the coefficient of discrimination (Tjur, 2009). Per-dichotomy only (NA in the Combined row).

For the **Combined** row the log-likelihood is the sum of the sub-model log-likelihoods (exploiting the independence of the nested dichotomies), and n is `nrow(x$data)` — the full sample size of the polytomous model — not the sum of per-dichotomy observation counts, which would double-count observations that appear in more than one sub-model.

A wider range of pseudo- R^2 measures for logistic-type models (`glm`, `polr`, `multinom`, `vglm`) is available in `DescTools::PseudoR2()`, including the Efron (1978) and McKelvey & Zavoina (1975) measures not implemented here. For an accessible overview see <https://statisticalhorizons.com/r2logistic/>.

Value

An object of class `c("RSQ.nestedLogit", "data.frame")` with one row per dichotomy plus a final "Combined" row, and columns response (the sub-model name), the requested pseudo-R² measures, and any additional statistics requested via `include`. The formula, object name, and digits are stored as attributes and used by the print method.

Author(s)

Michael Friendly

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See Also

`nestedLogit()`, `broom::glance()`, `DescTools::PseudoR2()`, `nnet::multinom()`, `MASS::polr()`

Examples

```
data("Womenlf", package = "carData")
wlf.nested <- nestedLogit(partic ~ hincome + children,
  logits(work = dichotomy("not.work", c("parttime", "fulltime")),
    full = dichotomy("parttime", "fulltime")),
  data = Womenlf)

# Default: McFadden, CoxSnell, Nagelkerke + AIC
RSQ(wlf.nested)

# All measures and all extra columns
RSQ(wlf.nested, which = "ALL", include = "ALL")

# Multinomial logit for comparison
if (requireNamespace("nnet", quietly = TRUE)) {
  wlf.multi <- nnet::multinom(partic ~ hincome + children, data = Womenlf,
```

```
                                trace = FALSE)
  RSQ(wlf.multi)
}

# Proportional-odds model for comparison
if (requireNamespace("MASS", quietly = TRUE)) {
  wlf.polr <- MASS::polr(partic ~ hincome + children, data = Women1f)
  RSQ(wlf.polr)
}
```

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