# Package 'BinaryEPPM'

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Type Package

**Title** Mean and Scale-Factor Modeling of Under- And Over-Dispersed Binary Data

Version 3.0

Imports Formula, expm, numDeriv, stats, lmtest, grDevices, graphics

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**Depends** R (>= 3.5.0)

Description Under- and over-dispersed binary data are modeled using an extended Poisson process model (EPPM) appropriate for binary data. A feature of the model is that the under-dispersion relative to the binomial distribution only needs to be greater than zero, but the over-dispersion is restricted compared to other distributional models such as the beta and correlated binomials. Because of this, the examples focus on under-dispersed data and how, in combination with the beta or correlated distributions, flexible models can be fitted to data displaying both under- and over-dispersion. Using Generalized Linear Model (GLM) terminology, the functions utilize linear predictors for the probability of success and scale-factor with various link functions for p, and log link for scale-factor, to fit a variety of models relevant to areas such as bioassay.

Details of the EPPM are in Faddy and Smith (2012) <doi:10.1002/bimj.201100214> and Smith and Faddy (2019) <doi:10.18637/jss.v090.i08>.

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Suggests R.rsp

VignetteBuilder R.rsp

NeedsCompilation no

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BinaryEPPM-package	Mean and Scale-Factor Modeling of Under- And Over-Dispersed Bi-
	nary Data

#### **Description**

Under- and over-dispersed binary data are modeled using an extended Poisson process model (EPPM) appropriate for binary data. A feature of the model is that the under-dispersion relative to the binomial distribution only needs to be greater than zero, but the over-dispersion is restricted compared to other distributional models such as the beta and correlated binomials. Because of this, the examples focus on under-dispersed data and how, in combination with the beta or correlated distributions, flexible models can be fitted to data displaying both under- and over-dispersion. Using Generalized Linear Model (GLM) terminology, the functions utilize linear predictors for the probability of success and scale-factor with various link functions for p, and log link for scale-factor, to fit a variety of models relevant to areas such as bioassay. Details of the EPPM are in Faddy and Smith (2012) and Smith and Faddy (2019). Two important changes from version 2.3 are the change to scale-factor rather than variance modeling, and the inclusion of a vignette.

#### **Details**

Index of help topics:

BBprob	Calculation of vector of probabilities for	LITE
	محالي المنافعة المناف	

beta binomial distribution.

in litters of varying sizes for the Berkshire

breed of pigs.

BinaryEPPM Fitting of EPPM models to binary data.

BinaryEPPM-package Mean and Scale-Factor Modeling of Under- And

Over-Dispersed Binary Data

CBprob Calculation of vector of probabilities for the

correlated binomial distribution.

EPPMprob Calculation of vector of probabilities for a

extended Poisson process model (EPPM).

GBprob Calculation of vector of probabilities for the

EPPM binomial distribution.

KupperHaseman.case Kupper and Haseman example data

LL.Regression.Binary Function called by optim to calculate the log

likelihood from the probabilities and hence perform the fitting of regression models to the

binary data.

LL.gradient Function used to calculate the first

derivatives of the log likelihood with respect

to the model parameters.

Model.BCBinProb Probabilities for beta and correlated binomial

distributions given p's and scale-factors.

Model.Binary Function for obtaining output from

distributional models.

Model.GB Probabilities for binomial and EPPM extended

binomial distributions given p's and b.

Model.JMVGB Probabilities for EPPM extended binomial

distributions given p's and scale-factors.

in litters of varying sizes for the Parkes

breed of pigs.

in litters of varying sizes for the Yorkshire

breed of pigs.

coef.BinaryEPPM Extraction of model coefficients for BinaryEPPM

Objects

cooks.distance.BinaryEPPM

Cook's distance for BinaryEPPM Objects

doubexp Double exponential Link Function
doubrecip Double reciprocal Link Function

fitted.BinaryEPPM Extraction of fitted values from BinaryEPPM

**Objects** 

hatvalues.BinaryEPPM Extraction of hat matrix values from BinaryEPPM

Objects

logLik.BinaryEPPM Extract Log-Likelihood loglog Log-log Link Function

negcomplog Negative complementary log-log Link Function plot.BinaryEPPM Diagnostic Plots for BinaryEPPM Objects

powerlogit Power Logit Link Function

predict.BinaryEPPM Prediction Method for BinaryEPPM Objects

print.BinaryEPPM Printing of BinaryEPPM Objects

print.summaryBinaryEPPM

Printing of summaryBinaryEPPM Objects

residuals.BinaryEPPM Residuals for BinaryEPPM Objects

ropespores.case Dilution series for the presence of rope

spores.

ropespores.grouped Dilution series for the presence of rope

spores.

summary.BinaryEPPM Summary of BinaryEPPM Objects

vcov.BinaryEPPM Variance/Covariance Matrix for Coefficients waldtest.BinaryEPPM Wald Test of Nested Models for BinaryEPPM

**Objects** 

wordcount.case Number of occurences of an article in five-word

and ten-word samples from two authors.

wordcount.grouped Number of occurences of an article in five-word

and ten-word samples from two authors.

#### Author(s)

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#### References

Cribari-Neto F, Zeileis A. (2010). Beta Regression in R. *Journal of Statistical Software*, **34**(2), 1-24. doi:10.18637/jss.v034.i02.

Faddy M, Smith D. (2012). Extended Poisson Process Modeling and Analysis of Grouped Binary Data. *Biometrical Journal*, **54**, 426-435. doi:10.1002/bimj.201100214.

Grun B, Kosmidis I, Zeileis A. (2012). Extended Beta Regression in R: Shaken, Stirred, Mixed, and Partitioned. *Journal of Statistical Software*, **48**(11), 1-25. doi:10.18637/jss.v048.i11.

Smith D, Faddy M. (2019). Mean and Variance Modeling of Under-Dispersed and Over-Dispersed Grouped Binary Data. *Journal of Statistical Software*, **90**(8), 1-20. doi:10.18637/jss.v090.i08.

Zeileis A, Croissant Y. (2010). Extended Model Formulas in R: Multiple Parts and Multiple Responses. *Journal of Statistical Software*, **34**(1), 1-13. doi:10.18637/jss.v034.i01.

#### See Also

CountsEPPM betareg

## **Examples**

**BBprob** 

Calculation of vector of probabilities for the beta binomial distribution.

## Description

Given a vector of parameters and a scalar of the number of trials the function returns a vector of probabilities.

#### Usage

```
BBprob(twoparameter, nt)
```

#### **Arguments**

twoparameter A vector of the parameters of the beta binomial distribution.

nt The number of trials.

#### Value

Vector of probabilities

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#### Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Smith D (1982). Algorithm AS189. Maximum Likelihood Estimation of the Parameters of the Beta Binomial Distribution. Applied Statistics, 32, 196-204.

Williams D (1996). "Overdispersion in Logistic Linear Models." In B Mrgan (ed.), Statistics in Toxicology, pp75-84. Oxford Science Publications.

## **Examples**

```
twoparameter <- c(0.96477815, 0.7561417) names(twoparameter) <- c('p', 'theta') nt <- 37 BBprob(twoparameter,nt)
```

Berkshires.litters

The data are of the number of male piglets born in litters of varying sizes for the Berkshire breed of pigs.

#### **Description**

The data are arranged as a list of binomial frequency distributions where the listing is by litter size which is included both as a variate (vsize) and as a factor (fsize)

#### Usage

```
data("Berkshires.litters")
```

## **Format**

The format is: List of 3 \$ fsize : Factor w/ 7 levels " size 5"," size 6",..: 1 2 3 4 5 6 7 \$ vsize : int [1:7] 5 6 7 8 9 10 11 \$ number.success:List of 7 ..\$ : num [1:6] 8 29 72 65 40 3 ..\$ : num [1:7] 5 22 89 129 74 35 4 ..\$ : num [1:8] 1 25 62 131 136 89 26 5 ..\$ : num [1:9] 1 15 79 179 219 149 71 33 4 ..\$ : num [1:10] 2 6 47 117 172 181 117 40 9 2 ..\$ : num [1:11] 2 1 23 65 131 145 120 61 20 3 ... ..\$ : num [1:12] 0 3 9 22 53 94 72 54 20 4 ...

#### Source

Brooks, R.J., James, W.H., Gray, E. (1993). Modelling Sub-Binomial Variation in the Frequency of Sex Combinations in Litters of Pigs. Biometrics 47, 403-417.

```
data("Berkshires.litters")
```

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BinaryEPPM

Fitting of EPPM models to binary data.

#### **Description**

Fits regression models to under- and over-dispersed binary data using extended Poisson process models

## Usage

```
BinaryEPPM(formula, data, subset = NULL, na.action = NULL,
    weights = NULL, model.type = "p only",
    model.name = "EPPM extended binomial", link = "cloglog",
    initial = NULL, method = "Nelder-Mead",
    pseudo.r.squared.type = "square of correlation", control = NULL)
```

#### **Arguments**

formula

Formulae for the probability of a success p and scale-factor. The object used is from the package Formula of Zeileis and Croissant (2010) which allows multiple parts and multiple responses. "formula" should consist of a left hand side (lhs) of single response variable and a right hand side (rhs) of one or two sets of variables for the linear predictors for the mean and (if two sets) the variance. This is as used for the R function "glm" and also, for example, as for the package "betareg" (Cribari-Neto and Zeileis, 2010). The function identifies from the argument data whether a data frame (as for use of "glm") or a list has been input. The list should be exactly the same as for a data frame except that the response variable is a list of vectors of frequency distributions rather than two vectors of paired counts of number responding (r) out of number tested as for the data frame. The subordinate functions fit models where the response variables are "p.obs", or "scalef.obs" according to the model type being fitted. The values for these response variables are not input as part of "data", they are calculated within the function from a list of grouped binary data input. If the "model.type" is "p only", "formula" consists of a lhs of the response variable and a rhs of the terms of the linear predictor for the mean model. If the "model.type" is "p and scale-factor" there are two sets of terms in the rhs of "formula" i.e., "p.obs" and "scalef.obs" together with the two sets of terms for the linear predictors of p and scale-factor.

data

"data" should be either a data frame (as for use of "glm") or a list. The list should be exactly the same as for a data frame except that the response variable is a list of vectors of frequency distributions rather than a vector of single counts as for the data frame. Only one list is allowed within "data" as it is identified as the dependent variable. If other lists are in "data", for example for use as weights, they should be removed from "data" prior to calling this function. The extracted list can be called using the "weights" argument to this function. Within the function a working list "listcounts" and data frames with components such as "p.obs", "scalef.obs", "covariates", "offset.mean", "offset.variance" are set up.

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The component "covariates" is a data frame of vectors of covariates in the model. The component "listcounts" is a list of vectors of frequency distributions, or the single pairs of r/n in grouped form if "data" is a data frame.

subset Subsetting commands.

model.type

model.name

link

na.action Action taken for NAs in data.
weights Vector of list of lists of weights.

Takes one of two values i.e. "p only" or "p and scale-factor". The "p only" value fits a linear predictor function to the parameter a in equation (3) of Faddy and Smith (2012). If the model type being fitted is binomial, modeling a is the same as modeling the mean. For the negative binomial the mean is  $b \exp(a)-1$ , b also being as in equation (3) of Faddy and Smith (2012). The "p and scale-factor" value fits linear predictor functions to both the probability of a success p and the

scale-factor.

If model.type is "p only" the model being fitted is one of the four "binomial", "EPPM extended binomial", "beta binomial", "correlated binomial". If model.type is "p and scale-factor" the model being fitted is either "EPPM extended binomial" i.e. as equations (4) and (6) of Faddy and Smith (2012) or one of the two

"beta binomial", "correlated binomial".

Takes one of nine values i.e., 'logit', 'probit', 'cloglog', 'cauchit', 'log', 'loglog', 'double exponential', 'double reciprocal', 'power logit'. The default is 'cloglog'. The 'power logit' has an attribute of 'power' for which the default is 1 i.e., a logit

link.

initial This is a vector of initial values for the parameters. If this vector is NULL

then initial values based on a fitting binomial models using "glm" are calculated

within the function.

method Takes one of the two values "Nelder-Mead" or "BFGS" these being arguments

of optim.

pseudo.r.squared.type

Takes one of the three values "square of correlation", "R square" or "max-rescaled R square". The "default" is as used in Cribari-Neto and Zeileis (2010) and is the square of the correlation between the observed and predicted values on the GLM linear predictor scale. The other two are as described in Cox and

Snell (1989), and Nagelkerke (1991) and apply to logistic regression.

control "control" is a list of control parameters as used in "optim". If this list is NULL

the defaults for "optim" are set as "control <- list(fnscale=-1, trace=0, maxit=1000)". The control parameters that can be changed by inputting a variable length list are "fnscale, trace, maxit, abstol, reltol, alpha, beta, gamma". Details of "optim"

and its control parameters are available in the online R help manuals.

#### Value

An object of class "BinaryEPMM" is returned. A list of object items follows.

data.type The type of the data i.e., data frame or list
list.data Data as a list of lists of frequency distributions

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call The call of the function formula The formula argument

model.type The type of model being fitted

model.name The model being fitted

link The link function

covariates.matrix.p

The design matrix for the probability of a success

covariates.matrix.scalef

The design matrix for the scalefactor

offset.p The offset vector for the probability of a success

offset.scalef The offset vector for the scalefactor

coefficients Estimates of model parameters

loglikelihood Loglikelihood

n The number of observations

The number of observations

The number of observations

df.null The degrees of freedom of the null model
df.residual The degrees of freedom of the residual

vnmax Vector of maximums of grouped count data vectors in list.counts

weights Vector or list of weights

converged Whether the iterative process converged, TRUE or FALSE

iterations Number of iterations taken

method Method for optim either Nelder-Mead or BFGS

pseudo.r.squared

Pseudo R\*\*2 value

start Starting values for iterative process
optim Estimates of model parameters
control Control parameters for optim

fitted.values Fitted values for probability of success

y Dependent variable terms Terms in model fitted

#### Author(s)

David M. Smith <dmccsmith@verizon.net>

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#### References

Cox DR, Snell EJ. (1989). Analysis of Binary Data. Second Edition. Chapman & Hall.

Cribari-Neto F, Zeileis A. (2010). Beta Regression in R. *Journal of Statistical Software*, **34**(2), 1-24. doi:10.18637/jss.v034.i02.

Grun B, Kosmidis I, Zeileis A. (2012). Extended Beta Regression in R: Shaken, Stirred, Mixed, and Partitioned. *Journal of Statistical Software*, **48**(11), 1-25. doi:10.18637/jss.v048.i11.

Faddy M, Smith D. (2012). Extended Poisson Process Modeling and Analysis of Grouped Binary Data. *Biometrical Journal*, **54**, 426-435. doi:10.1002/bimj.201100214.

Nagelkerke NJD. (1991). A Note on a General Definition of the Coefficient of Determination. *Biometrika*, **78**, 691-692.

Smith D, Faddy M. (2019). Mean and Variance Modeling of Under-Dispersed and Over-Dispersed Grouped Binary Data. *Journal of Statistical Software*, **90**(8), 1-20. doi:10.18637/jss.v090.i08.

Zeileis A, Croissant Y. (2010). Extended Model Formulas in R: Multiple Parts and Multiple Responses. *Journal of Statistical Software*, **34**(1), 1-13. doi:10.18637/jss.v034.i01.

#### See Also

CountsEPPM betareg

## **Examples**

**CBprob** 

Calculation of vector of probabilities for the correlated binomial distribution.

#### **Description**

Given a vector of parameters and a scalar of the number of trials the function returns a vector of probabilities.

#### Usage

```
CBprob(twoparameter, nt)
```

#### **Arguments**

twoparameter A vector of the parameters of the correlated binomial distribution.

nt The number of trials.

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#### Value

Vector of probabilities

## Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Kupper L, Haseman J (1978). The Use of a Correlated Binomial Model for the Analysis of Toxicological Experiments. Biometrics, 34(1), 69-76.

## **Examples**

```
twoparameter <- c(0.971242852, 0.001465007) names(twoparameter) <- c('p', 'rho') nt <- 37 CBprob(twoparameter,nt)
```

coef.BinaryEPPM

Extraction of model coefficients for BinaryEPPM Objects

## Description

Extract the regression model coefficients from models of class "BinaryEPMM".

#### Usage

```
## S3 method for class 'BinaryEPPM'
coef(object, prtpar = c("full", "p", "scale.factor"), ...)
```

## **Arguments**

object fitted model object of class "BinaryEPPM".

prtpar character indicating coefficients of the fitted model to be output: all coefficients

("full"), coefficients of the model for probability of success ("p"), coefficients of

the model for scale-factor ("scale.factor")

... some methods for this generic function require additional arguments.

## **Details**

One of a set of standard extractor functions for fitted model objects of class "BinaryEPPM.

#### Value

Vector of coefficients of fitted regression model.

#### Author(s)

David M. Smith <dmccsmith@verizon.net>

#### See Also

betareg

## **Examples**

cooks.distance.BinaryEPPM

Cook's distance for BinaryEPPM Objects

## Description

Calculates Cook's distances for BinaryEPPM objects.

## Usage

```
## S3 method for class 'BinaryEPPM'
cooks.distance(model, ...)
```

#### **Arguments**

model fitted model object of class "BinaryEPPM".
... some methods for this generic function require additional arguments.

## **Details**

Cook's distances as in GLMs.

## Value

A vector of Cook's distances.

#### Author(s)

David M. Smith <dmccsmith@verizon.net>

doubexp 13

#### References

Cribari-Neto F, Zeileis A. (2010). Beta Regression in R. *Journal of Statistical Software*, **34**(2), 1-24. doi:10.18637/jss.v034.i02.

#### See Also

betareg

#### **Examples**

doubexp

Double exponential Link Function

## **Description**

Computes the double exponential link function, including its inverse.

#### Usage

doubexp()

## Value

The double exponential transformation of theta.

#### Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Ford I, Torsney B, Wu C (1992). "The Use of a Canonical Form in the Construction of Locally Optimal Designs for Non-linear Problems." Journal of the Royal Statistical Society B, 54, 569-583. doi:10.1111/j.25176161.1992.tb01897.x

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doubrecip

Double reciprocal Link Function

#### **Description**

Computes the double reciprocal link function, including its inverse.

## Usage

doubrecip()

#### Value

The double reciprocal transformation of theta.

## Author(s)

David M. Smith <a href="mailto:smith@verizon.net">dmccsmith@verizon.net</a>

#### References

Ford I, Torsney B, Wu C (1992). "The Use of a Canonical Form in the Construction of Locally Optimal Designs for Non-linear Problems." Journal of the Royal Statistical Society B, 54, 569-583. doi:10.1111/j.25176161.1992.tb01897.x

**EPPMprob** 

Calculation of vector of probabilities for a extended Poisson process model (EPPM).

## **Description**

Calculates a vector of probabilities given a vector of rates using the matrix exponential function from Maechler, Dutang, Goulet, Bates, Firth (2023).

## Usage

EPPMprob(vlambda)

## **Arguments**

vlambda

a vector of rates of an extended Poisson process.

#### **Details**

This is a similar function to that in Smith and Faddy (2014).

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## Value

The value returned is a vector of probabilities.

#### Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Maechler M, Dutang C, Goulet V, Bates D, Firth D. (2023). expm: Matrix Exponential. R package version 0.999-8, https://CRAN.R-project.org/package=expm.

Smith D, Faddy M (2014). CountsEPPM: Mean and Variance Modeling of Count Data. R package version 2.0, https://CRAN.R-project.org/package=CountsEPPM.

fitted.BinaryEPPM

Extraction of fitted values from BinaryEPPM Objects

#### **Description**

This function is generic. Extract the fitted values from models of class "BinaryEPMM".

## Usage

```
## S3 method for class 'BinaryEPPM'
fitted(object, ...)
```

#### **Arguments**

object fitted model object of class "BinaryEPPM". . . . . currently not used.

#### **Details**

This function is included so that function lrtest from package lmtest can be used.

## Value

An vector of class "numeric" of the fitted values from the object of class "BinaryEPMM".

## Author(s)

David M. Smith <dmccsmith@verizon.net>

#### See Also

fitted

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GBprob	Calculation of vector of probabilities for the EPPM binomial distribution.

## **Description**

Given a vector of parameters and a scalar of the number of trials the function returns a vector of probabilities. The name GBprob is used to avoid confusion with EPPMprob which is the function calculating the probabilities given the constructed vector vector of lambdas.

## Usage

```
GBprob(twoparameter, nt)
```

## Arguments

twoparameter A vector of the parameters of the EPPM binomial distribution.

nt The number of trials.

#### Value

Vector of probabilities

## Author(s)

David M. Smith <a href="mailto:smith@verizon.net">dmccsmith@verizon.net</a>

#### References

Faddy M, Smith D. (2012). Extended Poisson Process Modeling and Analysis of Grouped Binary Data. *Biometrical Journal*, **54**, 426-435. doi:10.1002/bimj.201100214.

```
twoparameter <- c(0.971242852, 0.001465007) names(twoparameter) <- c('p', 'b') nt <- 37 GBprob(twoparameter,nt)
```

hatvalues.BinaryEPPM

hatvalues.BinaryEPPM Extraction of hat matrix values from BinaryEPPM Objects

## **Description**

Extract the values of the hat matrix from models of class "BinaryEPMM".

#### Usage

```
## S3 method for class 'BinaryEPPM'
hatvalues(model, ...)
```

#### **Arguments**

model fitted model object of class "BinaryEPPM".

... some methods for this generic function require additional arguments.

#### Value

The calculated hat values for the fitted model. These are used to calculate Cook's distances.

#### Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Cribari-Neto F, Zeileis A. (2010). Beta Regression in R. *Journal of Statistical Software*, **34**(2), 1-24. doi:10.18637/jss.v034.i02.

## See Also

betareg

18 LL.gradient

KupperHaseman.case

Kupper and Haseman example data

#### **Description**

Data of the number of deaths out of number of implants for pregnant female mice for two groups each of size 10.

#### Usage

```
data("KupperHaseman.case")
```

#### **Format**

A data frame with 20 observations on the following 3 variables.

Group a factor with levels Control Treated

Number. Deaths a numeric vector

Number. Implants a numeric vector

#### Source

Kupper L, Haseman J. (1978). The Use of a Correlated Binomial Model for the Analysis of Toxicological Experiments. Biometrics, 34(1), 69-76.

## **Examples**

```
data("KupperHaseman.case")
```

LL.gradient

Function used to calculate the first derivatives of the log likelihood with respect to the model parameters.

#### **Description**

Function used to calculate the first derivatives of the log likelihood with respect to the model parameters. These are numerical derivatives calculated using the numerical derivative functions of Gilbert and Varadhan (2015).

## Usage

```
LL.gradient(parameter, model.type, model.name, link, ntrials, nsuccess, covariates.matrix.p, covariates.matrix.scalef, offset.p, offset.scalef, weights, grad.method)
```

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#### **Arguments**

parameter A vector of the parameters of the model which is set to initial estimates on

function call.

model.type Takes one of two values i.e. 'p only' or 'p and scale-factor'. The 'p only' value

fits linear predictor functions to the probability of a success 'p' as in Faddy and Smith (2012). The 'p and scale-factor' value fits linear predictor functions to

both the 'p' and the scale-factor. The default is 'p and scale-factor'.

model.name If model.type is 'p only' the model being fitted is one of the four 'binomial',

'EPPM extended binomial', 'beta binomial' or 'correlated binomial'. If model.type is 'p and scale-factor' the model being fitted is one of the three 'EPPM extended binomial', 'beta binomial' or 'correlated binomial'. Information about these models is given in Faddy and Smith (2012). The default is 'EPPM extended

binomial'.

link Takes one of nine values i.e., 'logit', 'probit', 'cloglog', 'cauchit', 'log', 'loglog',

'double exponential', 'double reciprocal', 'power logit'. The default is 'cloglog'. The 'power logit' has an attribute of 'power' for which the default is 1 i.e., a logit

link.

ntrials A vector length 'n+1' representing the number of trials 'n' i.e., a vector with all

elements equal to 'n'.

nsuccess A vector representing the frequency distribution of the binomial distribution for

fixed number of trials 'n'.

covariates.matrix.p

A matrix of covariates for the mean where rows are the number of values in list.binary and columns the covariates. This matrix is extracted from the formulae in function BinaryEPPM. However, in the accompanying example it is shown how it can be constructed independently of function BinaryEPPM.

covariates.matrix.scalef

A matrix of covariates for the variance where rows are the number of values in list.binary and columns the covariates. The default is a vector of ones. This matrix is extracted from the formulae in function BinaryEPPM. However, in the accompanying example it is shown how it can be constructed independently of

function BinaryEPPM.

offset.p An offset vector for the probability of success p. The default is a vector of ones.

offset.scalef An offset vector for the scale-factor. The default is a vector of ones.

weights A vector or list of weights for the modeling of probability of success. The default

is a vector of ones.

grad.method Numerical method used to calculate gradients when the optimization method for

optim is BFGS either simple or Richardson. This is the grad.method attribute of

argument method of BinaryEPPM. The default is simple.

#### Value

A vector of numerical first derivatives.

#### Author(s)

David M. Smith <a href="mailto:smith@verizon.net">dmccsmith@verizon.net</a>

#### References

Gilbert P, Varadhan R. (2015). numDeriv: Accurate Numerical Derivatives. R Package version 2014.2-1, https://CRAN.R-project.org/package=numDeriv.

#### **Examples**

```
link <- 'cloglog'
attr(link, which="p") <- make.link(link)
nsuccess <- list(c(rep(0,5),352,479,530,291,101,17))
ntrials <- list(c(rep(10,11)))
parameter <- c(0.06363398,-0.47085362)
LL.gradient(parameter, model.type = "p and scale-factor",
    model.name = "EPPM extended binomial", link = link, ntrials = ntrials, nsuccess = nsuccess,
    covariates.matrix.p = matrix(c(1), nrow=1),
    covariates.matrix.scalef = matrix(c(1), nrow=1),
    offset.p = c(0), offset.scalef = c(0), weights = list(c(rep(1,11))),
    grad.method = "Richardson")</pre>
```

LL.Regression.Binary

Function called by optim to calculate the log likelihood from the probabilities and hence perform the fitting of regression models to the binary data.

## Description

Fits specified regression models to the data.

#### Usage

```
LL.Regression.Binary(parameter,model.type,model.name,link,ntrials,nsuccess, covariates.matrix.p,covariates.matrix.scalef, offset.p,offset.scalef,weights,grad.method)
```

## **Arguments**

parameter A vector of the parameters of the model which is set to initial estimates on

function call.

model.type Takes one of two values i.e. 'p only' or 'p and scale-factor'. The 'p only' value

fits linear predictor functions to the probability of a success 'p' as in Faddy and Smith (2012). The 'p and scale-factor' value fits linear predictor functions to

both the 'p' and the scale-factor. The default is 'p and scale-factor'.

model.name If model.type is 'p only' the model being fitted is one of the four 'binomial',

'EPPM extended binomial', 'beta binomial' or 'correlated binomial'. If model.type is 'p and scale-factor' the model being fitted is one of the three 'EPPM extended binomial', 'beta binomial' or 'correlated binomial'. Information about these models is given in Faddy and Smith (2012). The default is 'EPPM extended

binomial'.

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link Takes one of nine values i.e., 'logit', 'probit', 'cloglog', 'cauchit', 'log', 'loglog',

'double exponential', 'double reciprocal', 'power logit'. The default is 'cloglog'. The 'power logit' has an attribute of 'power' for which the default is 1 i.e., a logit

link.

ntrials A vector length 'n+1' representing the number of trials 'n' i.e., a vector with all

elements equal to 'n'.

nsuccess A vector representing the frequency distribution of the binomial distribution for

fixed number of trials 'n'.

covariates.matrix.p

A matrix of covariates for the mean where rows are the number of values in list.binary and columns the covariates. This matrix is extracted from the formulae in function BinaryEPPM. However, in the accompanying example it is shown how it can be constructed independently of function BinaryEPPM.

covariates.matrix.scalef

A matrix of covariates for the variance where rows are the number of values in list.binary and columns the covariates. The default is a vector of ones. This matrix is extracted from the formulae in function BinaryEPPM. However, in the accompanying example it is shown how it can be constructed independently of

function BinaryEPPM.

offset.p An offset vector for the probability of success p. The default is a vector of ones.

offset.scalef An offset vector for the scale-factor. The default is a vector of ones.

weights A vector or list of weights for the modeling of probability of success. The default

is a vector of ones.

grad.method Numerical method used to calculate gradients either simple or Richardson. The

default is simple.

#### Value

The log likelihood is returned.

## Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Faddy M, Smith D. (2012). Extended Poisson Process Modeling and Analysis of Grouped Binary Data. *Biometrical Journal*, **54**, 426-435. doi:10.1002/bimj.201100214.

22 logLik.BinaryEPPM

```
covariates.matrix.p = matrix(c(1), nrow=1),
covariates.matrix.scalef = matrix(c(1), nrow=1),
offset.p = c(0), offset.scalef = c(0),
weights = list(c(rep(1,11))))
```

logLik.BinaryEPPM

Extract Log-Likelihood

## **Description**

This function is generic. It is a method for extracting the log-likelihood for objects of class "BinaryEPPM".

## Usage

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

## Arguments

object fitted model object of class "BinaryEPPM".

... some methods for this generic function require additional arguments

## **Details**

logLik is most commonly used for a model fitted by maximum likelihood as is done here.

## Value

The log likelihood value for the fitted model object.

## Author(s)

David M. Smith <dmccsmith@verizon.net>

## See Also

betareg

loglog 23

loglog

Log-log Link Function

## **Description**

Computes the loglog link function, including its inverse.

## Usage

loglog()

## **Details**

Same link function as in Cribari-Neto and Zeileis (2010).

#### Value

The loglog of theta where the logarithms are to base e.

## Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Cribari-Neto F, Zeileis A. (2010). Beta Regression in R. *Journal of Statistical Software*, **34**(2), 1-24. doi:10.18637/jss.v034.i02.

Model.BCBinProb

Probabilities for beta and correlated binomial distributions given p's and scale-factors.

## **Description**

Calculates the probabilities for beta and correlated binomials given values for p's and scale-factors.

## Usage

```
Model.BCBinProb(parameter, model.type, model.name, link, ntrials, covariates.matrix.p,
covariates.matrix.scalef = matrix(c(rep(1, nrow(covariates.matrix.p))), ncol = 1),
offset.p = c(rep(0, length(ntrials))), offset.scalef = c(rep(0, length(ntrials))))
```

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## Arguments

parameter A vector of the parameters of the model which is set to initial estimates on

function call.

model.type Takes one of two values i.e. 'p only' or 'p and scale-factor'. The 'p only' value

fits a linear predictor function to the parameter p which is the 'm(1)' in equation (6) of Faddy and Smith (2012) divided by 'N'. The 'p and scale-factor' value

fits linear predictor functions to both p and the scale-factor.

model.name The model being fitted is one of the two 'beta binomial' or 'correlated binomial'.

link Takes one of nine values i.e., 'logit', 'probit', 'cloglog', 'cauchit', 'log', 'loglog',

'double exponential', 'double reciprocal', 'power logit'. The default is 'cloglog'. The 'power logit' has an attribute of 'power' for which the default is 1 i.e., a logit

link.

ntrials This is a scalar representing the denominator i.e., the length of the probability

mass function returned is this scalar + 1.

covariates.matrix.p

A matrix of covariates for p where rows are the number of values in listbinary and columns the covariates. This matrix is extracted from the formulae in function BinaryEPPM. However, in the accompanying example it is shown how it

can be constructed independently of function BinaryEPPM.

covariates.matrix.scalef

A matrix of covariates for the scale-factor where rows are the number of values in listbinary and columns the covariates. The default is a vector of ones. This matrix is extracted from the formulae in function BinaryEPPM. However, in the accompanying example it is shown how it can be constructed independently of

function BinaryEPPM.

offset.p An offset vector for p. The default is a vector of ones.

offset.scalef An offset vector for the scale-factor. The default is a vector of ones.

#### Value

List of arguments input together with a list of probabilities vectors and a data frame of values of p, theta (beta binomial) or rho (correlated binomial) and the limits for theta or rho.

model The model is either 'beta binomial' or 'correlated binomial'.

link The link is either 'logit' or 'cloglog'.

parameter A vector of the parameters of the model which is set to initial estimates on

function call.

probabilities A list of the vectors of probabilities of the model.

probabilities A data frame of values of p, theta (beta binomial) or rho (correlated binomial)

and the limits for theta or rho.

#### Author(s)

David M. Smith <a href="mailto:smith@verizon.net">dmccsmith@verizon.net</a>

Model.Binary 25

#### References

Hughes G, Madden L (1995). Some methods allowing for aggregated patterns of disease incidence in the analysis of data from designed experiments. Plant Pathology, 44, 927-943.

Kupper L, Haseman J (1978). The use of a correlated binomial model for the analysis of toxicological epxeriments. Biometrics, 34(1), 69-76.

## **Examples**

Model.Binary

Function for obtaining output from distributional models.

#### **Description**

Produces output of model, parameters and probabilities from the various models.

## Usage

Model.Binary(parameter, model.type, model.name, link, ntrials, covariates.matrix.p, covariates.matrix.scalef, offset.p, offset.scalef)

#### **Arguments**

parameter A vector of the parameters of the model which is set to initial estimates on function call.

model.type Takes one of two values i.e. 'p only' or 'p and scale-factor'. The 'p only' value

fits a linear predictor function to the parameter p which is the 'm(1)' in equation (6) of Faddy and Smith (2012) divided by 'N'. The 'p and scale-factor' value

fits linear predictor functions to both p and the scale-factor.

model.name If model.type is 'p only' the model being fitted is one of the six 'binomial',

'over-dispersed-one', 'over-dispersed-two', 'EPPM binomial', 'beta binomial' or 'correlated binomial'. If model type is 'p and scale-factor' the model being fitted is one of the three 'EPPM binomial', 'beta binomial' or 'correlated binomial'.

mial'.

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link Takes one of nine values i.e., 'logit', 'probit', 'cloglog', 'cauchit', 'log', 'loglog',

'double exponential', 'double reciprocal', 'power logit'. The default is 'cloglog'. The 'power logit' has an attribute of 'power' for which the default is 1 i.e., a logit

link.

ntrials This is a scalar representing the denominator i.e., the length of the probability

mass function returned is this scalar + 1.

covariates.matrix.p

A matrix of covariates for p where rows are the number of values in listbinary and columns the covariates. This matrix is extracted from the formulae in function BinaryEPPM. However, in the accompanying example it is shown how it can be constructed independently of function BinaryEPPM.

covariates.matrix.scalef

A matrix of covariates for the scale-factor where rows are the number of values in listbinary and columns the covariates. The default is a vector of ones. This matrix is extracted from the formulae in function BinaryEPPM. However, in the accompanying example it is shown how it can be constructed independently of function BinaryEPPM.

offset.p An offset vector for p. The default is a vector of ones.

offset.scalef An offset vector for the scale-factor. The default is a vector of ones.

#### Value

The output from either Model.BCBinProb, Model.GB, Model.Binary, Model.JMVGB, or Model.ODB.

#### Author(s)

David M. Smith <dmccsmith@verizon.net>

## References

Faddy M, Smith D. (2012). Extended Poisson Process Modeling and Analysis of Grouped Binary Data. *Biometrical Journal*, **54**, 426-435. doi:10.1002/bimj.201100214.

Model.GB 27

Model.GB	Probabilities for binomial and EPPM extended binomial distributions given p's and b.

## **Description**

Calculates the probabilities for binomial and EPPM extended binomial given values for p's and b.

## Usage

## Arguments

٤	guments	
	parameter	A vector of the parameters of the model which is set to initial estimates on function call.
	model.name	The model being fitted is one of the two 'binomial' or 'EPPM extended binomial'.
	link	Takes one of nine values i.e., 'logit', 'probit', 'cloglog', 'cauchit', 'log', 'loglog', 'double exponential', 'double reciprocal', 'power logit'. The default is 'cloglog'. The 'power logit' has an attribute of 'power' for which the default is 1 i.e., a logit link.
	ntrials	This is a scalar representing the denominator i.e., the length of the probability mass function returned is this scalar $+ 1$ .
covariates.matrix.p		
		A matrix of covariates for p where rows are the number of values in listbinary and columns the covariates. This matrix is extracted from the formulae in function BinaryEPPM. However, in the accompanying example it is shown how it can be constructed independently of function BinaryEPPM.

## Value

offset.p

parameter

List of arguments input together with a list of probabilities vectors and a data frame of values of a and b of Equation (5) of Faddy and Smith (2012).

An offset vector for p. The default is a vector of ones.

model The model is either 'binomial' or 'EPPM extended binomial'.

link The link is either 'logit' or 'cloglog'.

A vector of the parameters of the model which is set to initial estimates on

function call.

probabilities A list of the vectors of probabilities of the model.

Dparameters A data frame of values of a and b of Equation (5) of Faddy and Smith (2012).

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#### Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Faddy M, Smith D. (2012). Extended Poisson Process Modeling and Analysis of Grouped Binary Data. *Biometrical Journal*, **54**, 426-435. doi:10.1002/bimj.201100214.

#### **Examples**

Model.JMVGB

Probabilities for EPPM extended binomial distributions given p's and scale-factors.

## Description

Calculates the probabilities for binomial and generalized binomial given values for p's and scalefactors.

#### Usage

## **Arguments**

parameter A vector of the parameters of the model which is set to initial estimates on

function call.

model.name The model being fitted is one of the two 'binomial' or 'EPPM extended bino-

mial'.

link Takes one of nine values i.e., 'logit', 'probit', 'cloglog', 'cauchit', 'log', 'loglog',

'double exponential', 'double reciprocal', 'power logit'. The default is 'cloglog'. The 'power logit' has an attribute of 'power' for which the default is 1 i.e., a logit

link.

Model.JMVGB 29

ntrials

This is a scalar representing the denominator i.e., the length of the probability mass function returned is this scalar + 1.

covariates.matrix.p

A matrix of covariates for p where rows are the number of values in listbinary and columns the covariates. This matrix is extracted from the formulae in function BinaryEPPM. However, in the accompanying example it is shown how it can be constructed independently of function BinaryEPPM.

covariates.matrix.scalef

A matrix of covariates for the scale-factor where rows are the number of values in listbinary and columns the covariates. The default is a vector of ones. This matrix is extracted from the formulae in function BinaryEPPM. However, in the accompanying example it is shown how it can be constructed independently of function BinaryEPPM.

offset.p An offset vector for p. The default is a vector of ones.

offset.scalef An offset vector for the scale-factor. The default is a vector of ones.

#### Value

List of arguments input together with a list of probabilities vectors and a data frame of values of a and b of Equation (5) of Faddy and Smith (2012).

model The model is either 'binomial' or 'EPPM extended binomial'.

link The link is either 'logit' or 'cloglog'.

parameter A vector of the parameters of the model which is set to initial estimates on

function call.

probabilities A list of the vectors of probabilities of the model.

Dparameters A data frame of values of a and b of Equation (5) of Faddy and Smith (2012).

#### Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Faddy M, Smith D. (2012). Extended Poisson Process Modeling and Analysis of Grouped Binary Data. *Biometrical Journal*, **54**, 426-435. doi:10.1002/bimj.201100214.

30 Parkes.litters

negcomplog

Negative complementary log-log Link Function

## **Description**

Computes the negative complementary log-log link function, including its inverse.

#### Usage

```
negcomplog()
```

#### Value

The negative complementary log-log of theta.

#### Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Tibshirani RJ, Ciampi A (1983). "A Family of Proportional- and Additive-Hazards Models for Survival Data". Biometrics 39(1), 141-147.

Parkes.litters

The data are of the number of male piglets born in litters of varying sizes for the Parkes breed of pigs.

## Description

The data are arranged as a list of binomial frequency distributions where the listing is by litter size which is included both as a variate (vsize) and as a factor (fsize)

## Usage

```
data("Parkes.litters")
```

#### **Format**

The format is: List of 3 \$ fsize : Factor w/ 7 levels " size 5"," size 6",..: 1 2 3 4 5 6 7 \$ vsize : int [1:7] 5 6 7 8 9 10 11 \$ number.success:List of 7 ...\$ : num [1:6] 2 20 41 35 14 4 ...\$ : num [1:7] 3 16 53 78 53 18 0 ...\$ : num [1:8] 0 21 63 117 104 46 21 2 ...\$ : num [1:9] 1 8 37 81 162 77 30 5 1 ...\$ : num [1:10] 0 2 23 72 101 83 46 12 7 0 ...\$ : num [1:11] 0 7 8 19 79 82 48 24 10 0 ... ...\$ : num [1:12] 0 1 3 15 15 33 13 12 8 1 ...

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#### **Source**

Brooks, R.J., James, W.H., Gray, E. (1993). Modelling Sub-Binomial Variation in the Frequency of Sex Combinations in Litters of Pigs. Biometrics 47, 403-417.

#### **Examples**

```
data("Parkes.litters")
```

plot.BinaryEPPM

Diagnostic Plots for BinaryEPPM Objects

## **Description**

This function is generic. Various types of standard diagnostic plots can be produced, involving various types of residuals, influence measures etc. It is a minorly modified version of the generic plot function of **betareg** with details of the displays given in Cribari-Neto and Zeileis (2010). The same six displays and arguments list as in Cribari-Neto and Zeileis (2010) are used. The six displays are "Residuals vs indices of obs", "Cook's distance plot", "Leverage vs predicted values", "Residuals vs linear predictor", "Normal Q-Q plot of residuals", "Predicted vs observed values".

#### Usage

```
## S3 method for class 'BinaryEPPM'
plot(x, which = 1:4,
    caption = c("Residuals vs indices of obs.", "Cook's distance plot",
    "Leverage vs predicted values", "Residuals vs linear predictor",
    "Normal Q-Q plot of residuals", "Predicted vs observed values"),
    sub.caption = " ", main = "",
    ask = prod(par("mfcol"), 1) < length(which) && dev.interactive(),
    ..., type = "spearson")</pre>
```

#### **Arguments**

Χ	fitted model object of class "BinaryEPPM".
which	numeric. If a subset of plots is required, specify a subset of the numbers 1:6.
caption	character. Captions to appear above the plots.
sub.caption	character. Common title-above figures if there are multiple.
main	character. Title to each plot in addition to the above caption.
ask	logical. If true, the user is asked before each plot.
	other parameters to be passed through to plotting functions.
type	character indicating type of residual to be used, see residuals.BinaryEPPM.

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#### **Details**

The plot method for BinaryEPPM objects produces various plots of diagnostic plots similar to those produced by **betareg**. See Cribari-Neto and Zeileis (2010) for further details of the displays of **betareg**.

#### Value

No return value.

## Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Cribari-Neto F, Zeileis A. (2010). Beta Regression in R. *Journal of Statistical Software*, **34**(2), 1-24. doi:10.18637/jss.v034.i02.

#### See Also

```
plot.betareg
```

## **Examples**

powerlogit

Power Logit Link Function

## Description

Computes the power logit link function, including its inverse.

#### Usage

```
powerlogit(power = 1)
```

#### **Arguments**

power

power value for the power logit link function.

#### Value

The power logit transformation of theta. All logarithms are natural ones, i.e., to base e.

predict.BinaryEPPM 33

#### Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Gaudard MA, Karson MJ, Linder E, Tse Sk (1993). Efficient Designs for Estimation in the Power Logistic Quantal Response Model." Statistica Sinica, 3, 233-243.

predict.BinaryEPPM

Prediction Method for BinaryEPPM Objects

## **Description**

Extract various types of predictions from BinaryEPPM regression models.

## Usage

## Arguments

•	•	
	object	fitted model object of class "BinaryEPPM".
	newdata	optionally, a data frame in which to look for variables with which to predict. If omitted, the original observations are used.
	type	character indicating type of predictions: fitted means of responses ("response"), linear predictors ("linear.predictor.p", "linear.predictor.scale.factor"), fitted value of probability of success ("p"), fitted value of scale-factor ("scale.factor"), fitted value of mean ("mean"), scale factor limits ("scale.factor.limits"), fitted value of variance ("variance"), fitted probability distribution ("distribution"), parameters of fitted distributions ("distribution.parameters")
	na.action	function determining what should be done with missing values in <i>newdata</i> . The default is to predict NA.
		some methods for this generic function require additional arguments.

## Value

A vector or list of the predicted values from the fitted model object.

## Author(s)

David M. Smith <dmccsmith@verizon.net>

34 print.BinaryEPPM

#### References

Cribari-Neto F, Zeileis A. (2010). Beta Regression in R. *Journal of Statistical Software*, **34**(2), 1-24. doi:10.18637/jss.v034.i02.

#### See Also

```
predict.betareg
```

## **Examples**

print.BinaryEPPM

Printing of BinaryEPPM Objects

## **Description**

Prints objects of class "BinaryEPPM".

#### Usage

```
## S3 method for class 'BinaryEPPM'
print(x, digits = max(3, getOption("digits") - 3), ...)
```

#### **Arguments**

x fitted model object of class "BinaryEPPM".

digits digits of printed output.
... not currently used.

## Value

An object of class "BinaryEPPM" is constructed. This object has the following attributes.

data.type Indicator of the type of data either 0 "data.frame" or 1 "list".

list.data Regardless of the "data.type", the data in list form.

call The "call" to the function "BinaryEPPM".

formula The model formula in "call".

model.type The model type in "call".

model.name The model name in "call".

link The link function in "call".

print.BinaryEPPM 35

covariates.matrix.p

The matrix of covariates for the model for p.

covariates.matrix.scalef

The matrix of covariates for the model for scale-factor.

offset.p The vector of offsets for the model for p.

offset.scalef The vector of offsets for the model for scale-factor.

coefficients The coefficients of the fitted model.

loglik The log-likelihood of the fitted model.

vcov The variance-covariance matrix of the fitted model.

n The number of observations. Relabelled duplication of "nobs" needed when

calling function "lrtest".

nobs The number of observations.

df.null The degrees of freedom of the null model.

df.residual The degrees of freedom of the residual model.

vnmax Vector of number of "trials" in each observation.

weights Vector of weights for observation.

converged Indicator of convergence.

method Method used to calculate pseudo.r.squared.

pseudo.r.squared

The value of the coefficient of determination r squared.

start Initial estimates.
optim Final model fit.

control Control parameters for optimization function "optim".

fitted.values The fitted values.

y The dependent variable in the model.

terms The terms in the model.

#### Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Cribari-Neto F, Zeileis A. (2010). Beta Regression in R. *Journal of Statistical Software*, **34**(2), 1-24. doi:10.18637/jss.v034.i02.

## See Also

betareg

```
print.summaryBinaryEPPM
```

Printing of summaryBinaryEPPM Objects

## Description

Prints the objects of class "summaryBinaryEPPM".

## Usage

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

## **Arguments**

```
x object output by summary.BinaryEPPM.
... not currently used.
```

## Value

No return value.

#### Author(s)

David M. Smith <a href="mailto:smith@verizon.net">dmccsmith@verizon.net</a>

## References

Cribari-Neto F, Zeileis A. (2010). Beta Regression in R. *Journal of Statistical Software*, **34**(2), 1-24. doi:10.18637/jss.v034.i02.

## See Also

betareg

residuals.BinaryEPPM

residuals.BinaryEPPM Residuals for BinaryEPPM Objects

## **Description**

This function is generic. Extract various types of residuals from objects of class "BinaryEPPM".

## Usage

## **Arguments**

object Fitted model object of class "BinaryEPPM".

type Type of residuals wanted i.e., standardized Pearson "spearson", deviance "de-

viance", Pearson "pearson",response "response", likelihood "likelihood", stan-

dardized deviance "sdeviance".

... Some methods for this generic function require additional arguments.

#### **Details**

Residuals as Cribari-Neto and Zeileis (2010).

#### Value

An vector of class "numeric" of residuals of a specified type from the object of class "BinaryEPMM".

## Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Cribari-Neto F, Zeileis A. (2010). Beta Regression in R. *Journal of Statistical Software*, **34**(2), 1-24. doi:10.18637/jss.v034.i02.

## See Also

```
residuals.betareg
```

38 ropespores.grouped

ropespores.case

Dilution series for the presence of rope spores.

## Description

Dilution series where at each dilution of a suspension of potato flour a number of samples were examined for the presence of rope spores. These data are in data frame form.

## Usage

```
data("ropespores.case")
```

#### **Format**

A data frame with 10 observations on the following 5 variables.

```
vdilution a numeric vector
fdilution a factor with levels 0.25 0.5 1 2 4 8 16 32 64 128
logdilution a numeric vector
number.spores a numeric vector
number.tested a numeric vector
```

#### **Source**

Finney, D.J. (1971). Statistical Methods in Biological Assay. Griffin, London, 2nd edition.

## **Examples**

```
data("ropespores.case")
```

ropespores.grouped

Dilution series for the presence of rope spores.

## **Description**

Dilution series where at each dilution of a suspension of potato flour a number of samples were examined for the presence of rope spores. These data are in list form.

## Usage

```
data("ropespores.grouped")
```

#### **Format**

#### Source

Finney, D.J. (1971). Statistical Methods in Biological Assay. Griffin, London, 2nd edition.

## **Examples**

```
data("ropespores.grouped")
```

summary.BinaryEPPM

Summary of BinaryEPPM Objects

#### **Description**

This function is generic. Summary of objects of class "BinaryEPPM".

#### Usage

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

#### **Arguments**

object Fitted model object of class "BinaryEPPM".

... some methods for this generic function require additional arguments.

## **Details**

Similar output to that of summary.glm "summary.glm" and summary.betareg Cribari-Neto and Zeileis (2010).

#### Value

An object of class "summaryBinaryEPPM" is constructed. This object has the following attributes.

data.type Indicator of the type of data either 0 "data.frame" or 1 "list".

call The "call" to the function "BinaryEPPM".

formula The model formula in "call".

model.type The model type in "call".

model.name The model name in "call".

link The link function in "call".

offset.p The vector of offsets for the model for p.

offset.scalef The vector of offsets for the model for scale-factor.

coeff.table.p The coefficients of the fitted model for p.

coeff.table.scalef

The coefficients of the fitted model for scale-factor.

loglik The log-likelihood of the fitted model.

n The number of observations. Relabelled duplication of "nobs" needed when

calling function "lrtest".

nobs The number of observations.

df.null The degrees of freedom of the null model.

df.residual The degrees of freedom of the residual model.

vnmax Vector of number of "trials" in each observation.

weights Vector of weights for observation.

converged Indicator of convergence.

method Method used to calculate pseudo.r.squared.

pseudo.r.squared

The value of the coefficient of determination r squared.

optim Final model fit.

control Control parameters for optimization function "optim".

fitted.values The fitted values.

y The dependent variable in the model.

terms The terms in the model.

npar The number of parameters in the model.

#### Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Cribari-Neto F, Zeileis A. (2010). Beta Regression in R. *Journal of Statistical Software*, **34**(2), 1-24. doi:10.18637/jss.v034.i02.

#### See Also

summary.betareg print.summaryBinaryEPPM

vcov.BinaryEPPM 41

VCOV	Rina	rvFPPM

Variance/Covariance Matrix for Coefficients

## **Description**

Variance/covariance matrix for coefficients of fitted model.

## Usage

```
## S3 method for class 'BinaryEPPM'
vcov(object, model = c("full", "p", "scale.factor"), ...)
```

## **Arguments**

object fitted model object of class "BinaryEPPM".

model character indicating variance/covariance matrix for all coefficients to be output:

all coefficients ("full"), variance/covariance matrix for coefficients of probability of success ("p"), variance/covariance matrix for coefficients of scale-factor

("scale.factor")

... other parameters to be passed through to plotting functions.

#### Value

The variance/covariance matrix of the parameters of the fitted model object.

#### Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Cribari-Neto F, Zeileis A. (2010). Beta Regression in R. *Journal of Statistical Software*, **34**(2), 1-24. doi:10.18637/jss.v034.i02.

#### See Also

betareg

waldtest.BinaryEPPM

Wald Test of Nested Models for BinaryEPPM Objects

## **Description**

waldtest is a generic function for comparisons of nested (generalized) linear models via Wald tests.

## Usage

```
## S3 method for class 'BinaryEPPM'
waldtest(object, ..., vcov = NULL,
    test = c("Chisq", "F"))
```

## **Arguments**

object an object of class "BinaryEPPM".

... further object specifications passed to methods. See below for details.

vcov a function for estimating the covariance matrix of the regression coefficients. If

only two models are compared it can also be the covariance matrix of the more

general model.

test character specifying whether to compute the large sample Chi-squared statistic

(with asymptotic Chi-squared distribution) or the finite sample F statistic (with

approximate F distribution).

#### **Details**

waldtest is a generic function for comparisons of nested (generalized)linear models via Wald tests. It does not have the same functionality as the versions of **betareg** and **lmtest** with a reduced list of arguments. With these caveats, more details can be obtained from the **Details** pages of those packages.

#### Value

An object of class "anova" which contains the residual degrees of freedom, the difference in degrees of freedom, Wald statistic (either "Chisq" or "F") and corresponding p value.

## Author(s)

David M. Smith <dmccsmith@verizon.net>

#### References

Cribari-Neto F, Zeileis A. (2010). Beta Regression in R. *Journal of Statistical Software*, **34**(2), 1-24. doi:10.18637/jss.v034.i02.

Zeileis A, Hothorn T. (2002). Diagnostic Checking in Regression Relationships. *R News*, **2**(3), 7-10. https://CRAN.R-project.org/doc/Rnews/.

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## See Also

```
waldtest betareg
```

#### **Examples**

```
data("ropespores.case")
output.fn <- BinaryEPPM(data = ropespores.case,
  number.spores / number.tested ~ 1 + offset(logdilution),
  model.type = 'p only', model.name = 'binomial')
output.fn.one <- BinaryEPPM(data = ropespores.case,
  number.spores / number.tested ~ 1 + offset(logdilution),
  model.type = 'p only', model.name = 'beta binomial')
waldtest.BinaryEPPM(output.fn, output.fn.one, test = c("Chisq", "F"),
  vcov = vcov)</pre>
```

wordcount.case

Number of occurences of an article in five-word and ten-word samples from two authors.

## Description

The data are the number of occurences of an article in five-word and ten-word samples from Macaulay's 'Essay on Milton' and G.K. Chesterton's essay 'About the workers'.

## Usage

```
data("wordcount.case")
```

#### **Format**

A data frame with 340 observations on the following 5 variables.

```
author a factor with levels Macaulay Chesterton
fsize a factor with levels 5 10
vsize a numeric vector
number.words a numeric vector
number.tested a numeric vector
```

#### Source

Bailey, B.J.R. (1990). A model for Function Word Counts. Appl. Statist. 39(1), 107-114.

#### References

Sellers, K.F., Swift, A.W., Weems, K.S. (2017). A flexible distribution class for count data. Journal of Statistical Distributions and Applications 41(12), 2616-2626.

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#### **Examples**

```
data(wordcount.case)
```

wordcount.grouped

Number of occurences of an article in five-word and ten-word samples from two authors.

## **Description**

The data are the number of occurences of an article in five-word and ten-word samples from Macaulay's 'Essay on Milton' and G.K. Chesterton's essay 'About the workers'.

## Usage

```
data("wordcount.grouped")
```

#### **Format**

The format is: List of 4 \$ author : Factor w/ 2 levels " Macaulay", "Chesterton": 1 1 2 2 \$ fsize : Factor w/ 2 levels "5", "10": 1 2 1 2 \$ vsize : num [1:4] 5 10 5 10 \$ number.words:List of 4 ..\$ : num [1:6] 45 49 6 0 0 0 ..\$ : num [1:11] 27 44 26 3 0 0 0 0 0 0 ... ..\$ : num [1:6] 32 35 3 0 0 0 ..\$ : num [1:11] 14 38 16 2 0 0 0 0 0 0 ...

#### **Source**

Bailey, B.J.R. (1990). A model for Function Word Counts. Appl. Statist. 39(1), 107-114.

#### References

Sellers, K.F., Swift, A.W., Weems, K.S. (2017). A flexible distribution class for count data. Journal of Statistical Distributions and Applications 41(12), 2616-2626.

## **Examples**

data(wordcount.grouped)

Yorkshires.litters 45

Yorkshires.litters The data are of the number of male piglets sizes for the Yorkshire breed of pigs.	born in litters of varying
--	----------------------------

## Description

The data are arranged as a list of binomial frequency distributions where the listing is by litter size which is included both as a variate (vsize) and as a factor (fsize)

## Usage

```
data("Yorkshires.litters")
```

#### **Format**

The format is: List of 3 \$ fsize : Factor w/ 9 levels " size 5"," size 6",...: 1 2 3 4 5 6 7 8 9 \$ vsize : int [1:9] 5 6 7 8 9 10 11 12 13 \$ number.success:List of 9 ...\$ : num [1:6] 3 22 30 37 13 5 ...\$ : num [1:7] 7 18 44 62 27 17 4 ...\$ : num [1:8] 2 14 25 63 69 41 12 5 ...\$ : num [1:9] 2 15 32 70 127 90 45 18 1 ...\$ : num [1:10] 0 8 33 63 106 115 62 30 11 1 ...\$ : num [1:11] 0 3 20 49 79 119 91 59 23 4 ... ...\$ : num [1:12] 0 0 7 20 60 94 100 47 31 9 ... ...\$ : num [1:13] 0 1 6 16 29 52 66 43 34 22 ... ...\$ : num [1:14] 0 2 2 2 14 19 44 45 22 13 ...

#### **Source**

Brooks, R.J., James, W.H., Gray, E. (1993). Modelling Sub-Binomial Variation in the Frequency of Sex Combinations in Litters of Pigs. Biometrics 47, 403-417.

```
data("Yorkshires.litters")
```

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