Package 'MixedPsy'

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Description Tools for the analysis of psychophysical data. This package allows to estimate the Point of Subjective Equivalence (PSE) and the Just Noticeable Difference (JND), either from a psychometric function or from a Generalized Linear Mixed Model (GLMM). Additionally, the package allows plotting the fitted models and the response data, simulating psy chometric functions of different shapes, and simulating data sets. For a description of the use of GLMMs applied to psychophysical data, refer to Moscatelli et al. (2012), <doi:10.1167 12.11.26="">.</doi:10.1167>
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MixDelta

PSE/JND for Univariable GLMM Using Delta Methods

Description

Estimate the Point of Subjective Equivalence (PSE), the Just Noticeable Difference (JND) and the related Standard Errors for an univariate distribution by means of Delta Method.

Usage

```
MixDelta(xplode.obj, alpha = 0.05)
```

Arguments

xplode.obj an object of class xplode.obj (univariable GLMMs).

alpha significance level of the confidence interval. Default is 0.05.

Details

MixDelta estimates PSE and JND of a univariable psychometric function (object of class "glm"). The method only applies to univariable GLMMs having a *probit* link function. Use MixTreatment for multivariable GLMMs.

Value

MixDelta returns a list of length 1 including Estimate, Standard Error, Inferior and Superior Confidence Interval of PSE and JND. Confidence Intervals are computed as: $Estimate + / - z(1 - (\alpha/2)) * Std.Error$.

Note

The function assumes that the first model coefficient is the intercept and the second is the slope. The estimate of the JND assumes a *probit* link function.

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References

Moscatelli, A., Mezzetti, M., & Lacquaniti, F. (2012). Modeling psychophysical data at the population-level: The generalized linear mixed model. Journal of Vision, 12(11):26, 1-17. https://doi.org/10.1167/12.11.26 Casella, G., & Berger, R. L. (2002). Statistical inference (2nd ed.). Pacific Grove, CA: Duxbury Press

See Also

MixTreatment for univarible and multivariable GLMM. pseMer for bootstrap-based confidence intervals. xplode objects of class xplode.obj.

Examples

MixPlot

Plotting univariable GLMM

Description

Plot binomial data and the fitted GLMM (object of class xplode).

Usage

```
MixPlot(xplode.obj, pf = 1, p05line = F, x.range, x.ref, col = F,
    x.label = "Stimulus Intensity", y.label = "Predicted Response")
```

xplode.obj	an object of class xplode
pf	integer: for multivariable GLMM including one factorial predictor, the level number to be plotted
p05line	logical, should be an horizontal and a vertical line added? the horizontal line is fixed at $P(Y=1)=0.5$.
x.range	a vector of length two specifying the range for model predictions
x.ref	if $p05$ line = T, this is the position of the vertical line on the x axis
col	logical, if TRUE a different color will be used for different clusters/participants
x.label, y.label	
	label for the x and the y axes. If not specified, x.labels = ""Stimulus Intensity", y.label = "Predicted Response"

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Value

a data.frame object including the intercept and slope for each participant (algebraic sum of the fixed effects and the modes of the random effects) and the color number for the plot.

Note

The function is currently only working with GLMM including maximum three random effects (random intercept, random slope and covariance of the two)

See Also

```
xplode objects of class xplode.obj.
```

Examples

MixTreatment

PSE/JND for Multivariable GLMM Using Delta Methods

Description

Estimate the Point of Subjective Equivalence (PSE), the Just Noticeable Difference (JND) and the related Standard Errors for a multivariate distribution by means of Delta Method. The method applies to multivariable GLMM having a *probit* link function. The function is based on a recursive use of glmer and MixDelta

Usage

```
MixTreatment(xplode.obj, datafr)
```

```
xplode.obj an object of class xplode.obj. The fitted model (object of class "merMod") from xplode.obj includes one continuous predictor and one factorial predictor.

datafr the data frame fitted with the GLMM model
```

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Details

The function MixTreatment is based on a recursive use of glmer and PsychDelta to multivariable GLMM including continuous and factorial predictors. The same caveats of PsychDelta apply (e.g., confidence interval based on normality assumption).

Value

A list, whose length is equal to the levels of the factorial predictor, i. Each cell of the list is equal to the output of delta.psy.probit applied to a multivariable model whose baseline is level i of the factorial predictor.

References

Moscatelli, A., Mezzetti, M., & Lacquaniti, F. (2012). Modeling psychophysical data at the population-level: The generalized linear mixed model. Journal of Vision, 12(11):26, 1-17. https://doi.org/10.1167/12.11.26

See Also

glmer for Generalized Linear Mixed Models (including random effects).MixDelta for univariable model with delta method. pseMer for bootstrap-based confidence intervals.

Examples

```
library(lme4)
data(vibro_exp3)
formula.mod <- cbind(faster, slower) ~ speed * vibration + (1 + speed| subject)
mod <- glmer(formula = formula.mod, family = binomial(link = "probit"), data = vibro_exp3)
xplode.mod <- xplode(model = mod, name.cont = "speed", name.factor = "vibration")
MixTreatment(xplode.mod, vibro_exp3)</pre>
```

pseMer

PSE/JND for GLMM Using Bootstrap Methods

Description

Estimates the Point of Subjective Equivalence (PSE), the Just Noticeable Difference (JND) and the related Standard Errors by means of Bootstrap Method.

Usage

```
pseMer(mer.obj, B = 200, FUN = NULL, alpha = 0.05, ci.type = c("norm",
   "basic", "perc"), beep = F)
```

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Arguments

mer.obj An object of class "merMod".

B integer: the number of bootstrap samples.

FUN An optional, custom made function to specify the required parameters to be es-

timated. if NULL, pseMer() will estimate the PSE and the JND of a univariable

GLMM.

alpha Significance level of the confidence interval.

ci.type A vector of character strings representing the type of intervals required. The

value should be any subset of the values c("norm", "basic", "stud", "perc", "bca") or simply "all" which will compute all five types of intervals. "perc" should be

always included for the summary table.

beep Logical. If TRUE, a "ping" sound alerts that the simulation is complete.

Details

pseMer estimates PSE and JND (and additional user defined paremters) from a fitted GLMM model (class "merMod"). The "ping" sound is provided by beep function from the beepr package.

Value

pseMer returns a list of length 3 including a summary table (Estimate, Standard Error, Inferior and Superior Confidence Interval of the parameters) and the output of bootMer and boot.ci functions, for further analises. Confidence Intervals in the summary table are based on the percentile method.

Note

A first custom function was written in 2012 for the non-CRAN package MERpsychophisics, based on the algorithm in Moscatelli et al. (2012). The current function is a simple wrapper of lme4::bootMer() and boot::boot.ci() functions.

Increasing the nuber of bootstrap samples (B) makes the estimate more reliable. However, this will also increase the duration of the computation.

References

Moscatelli, A., Mezzetti, M., & Lacquaniti, F. (2012). Modeling psychophysical data at the population-level: The generalized linear mixed model. Journal of Vision, 12(11):26, 1-17. https://doi.org/10.1167/12.11.26

Bates, D., Mächler, M., Bolker, B., & Walker, S. (2015). Fitting Linear Mixed-Effects Models Using Ime4. Journal of Statistical Software, 67(1), 51. https://doi.org/10.18637/jss.v067.i01

See Also

bootMer from lme4 package and boot.ci from boot package.

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Examples

```
## Example 1: estimate pse/jnd of a univariable GLMM
library(lme4)
data(vibro_exp3)
formula.mod1 <- cbind(faster, slower) ~ speed + (1 + speed| subject)</pre>
mod1 <- glmer(formula = formula.mod1, family = binomial(link = "probit"),</pre>
              data = vibro_exp3[vibro_exp3$vibration == 0,])
BootEstim.1 <- pseMer(mod1, B = 100, ci.type = c("perc"))
## Example 2: specify custom parameters for bootstrap estimation of a
# multivariate model
formula.mod2 <- cbind(faster, slower) ~ speed * vibration + (1 + speed| subject)</pre>
mod2 <- glmer(formula = formula.mod2, family = binomial(link = "probit"),</pre>
               data = vibro_exp3)
fun2mod = function(mer.obj){
#allocate space: 4 parameters (jnd_0Hz, jnd_32Hz, pse_0Hz, pse_32Hz) j
jndpse = vector(mode = "numeric", length = 4)
names(jndpse) = c("jnd_0Hz","jnd_32Hz", "pse_0Hz", "pse_32Hz")
jndpse[1] = qnorm(0.75)/fixef(mer.obj)[2] #jnd_0Hz
jndpse[2] = qnorm(0.75)/(fixef(mer.obj)[2] + fixef(mer.obj)[4]) #jnd_32Hz
jndpse[3] = -fixef(mer.obj)[1]/fixef(mer.obj)[2] #pse_0Hz
jndpse[4] = -(fixef(mer.obj)[1] + fixef(mer.obj)[3])/(fixef(mer.obj)[2]
               + fixef(mer.obj)[4]) #pse_32Hz
return(jndpse)
}
BootEstim.2 = pseMer(mod2, B = 100, FUN = fun2mod)
```

PsychDelta

PSE/JND for univariable GLM Using Delta Method

Description

Estimate the Point of Subjective Equivalence (PSE), the Just Noticeable Difference (JND) and the related Standard Errors by means of Delta Method. The method only applies to univariable GLMs (psychometric functions) having a *probit* link function.

Usage

```
PsychDelta(model, alpha = 0.05)
```

```
model the fitted psychometric function. An object of class "glm". alpha significance level of the confidence interval.
```

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Details

PsychDelta estimates PSE and JND of a univariable psychometric function (object of class "glm").

Value

PsychDelta returns a matrix including Estimate, Standard Error, Inferior and Superior Confidence Interval of PSE and JND. Confidence Intervals are computed as: $Estimate + / - z(1 - (\alpha/2)) * Std.Error$.

Note

The function assumes that the first model coefficient is the intercept and the second is the slope. The estimate of the JND assumes a *probit* link function.

References

Faraggi, D., Izikson, P., & Reiser, B. (2003). Confidence intervals for the 50 per cent response dose. Statistics in medicine, 22(12), 1977-1988. https://doi.org/10.1002/sim.1368

Moscatelli, A., Mezzetti, M., & Lacquaniti, F. (2012). Modeling psychophysical data at the population-level: The generalized linear mixed model. Journal of Vision, 12(11):26, 1-17. https://doi.org/10.1167/12.11.26

See Also

glm for Generalized Linear Models (without random effects) and glmer for Generalized Linear Mixed Models (including random effects). MixDelta and MixTreatment for univariable and multivariable GLMM, respectively (object of class "merMod"). pseMer for bootstrap-based confidence intervals.

Examples

```
#load simulated data
data(simul_data)
#fit a glm (probit link)
model.glm = glm(formula = cbind(Longer, Total - Longer) ~ X,
family = binomial(link = "probit"), data = simul_data)
PsychDelta(model.glm)
```

PsychFunction

Fitting and Plotting Psychometric Functions

Description

Fit psychometric functions using either glm() or brglm(), estimate PSE, JND and the related confidence intervals, and draw the curve on an existing plot.

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Usage

```
PsychFunction(ps.formula, ps.link, ps.data, x.range = c(NA, NA), ps.x = NA,
  ps.lines = F, ps.col = "black", ps.lty = "dashed", ps.lwd = 1,
  br = F)
```

Arguments

ps.formula	an object of class "formula", such as cbind(yes, no) ~ X
ps.link	a link function for the binomial family of error distribution. See 'Details
ps.data	a data frame including the variables in the model
x.range	a vector of length two specifying the range for model predictions
ps.x	optionally, a data frame in which to look for variables with which to predict. See 'Details'
ps.lines	logical. If TRUE, model predictions and confidence intervals of the PSE will be added to an existing plot
ps.col	color of the lines to be plotted
ps.lty	line type
ps.lwd	line width
br	logical. If TRUE, brglm is used if fitted values are equal to 0 or 1

Details

If lines = TRUE, the function draws model predictions on an existing plot. Only for univariable glm of the type $F(Y) \sim X$, where X is a continuous predictor. If ps.x is empty, the new data frame is a vector of length = 1000, whose range is specified from x.range. Std. Errors and 95% confidence intervals of the PSE and JND are estimated via Delta Methods, see Faraggi et al. (2003).

Value

a list including the fitted glm (or brglm), the estimate of PSE and JND and a flag to indicate if brglm was called.

References

Faraggi, D., Izikson, P., & Reiser, B. (2003). Confidence intervals for the 50 per cent response dose. Statistics in medicine, 22(12), 1977-1988. https://doi.org/10.1002/sim.1368

Moscatelli, A., Mezzetti, M., & Lacquaniti, F. (2012). Modeling psychophysical data at the population-level: The generalized linear mixed model. Journal of Vision, 12(11):26, 1-17. https://doi.org/10.1167/12.11.26

See Also

glm for for Generalized Linear Models. PsychShape for plotting psychometric function of given PSE and JND

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Examples

PsychShape

Plotting Psychometric Functions given PSE and JND

Description

PsychShape() plots a psychometric function with known pse and jnd on an existing plot.

Usage

```
PsychShape(pse = 0, jnd, x.range = c(NA, NA), ps.link = "probit",
  ps.col = "black", ps.lwd = 1, ps.lty = "solid")
```

Arguments

pse, jnd	the pse and the jnd of the desired psychometric function
x.range	a vector of length two specifying the range of the function
ps.link	a link function for the binomial family of error distribution (see Details).
ps.col	color of the line to be plotted
ps.lwd	line width
ps.lty	line type

Details

PsychShape() can be used to visualize the predicted results of a psychophysical experiment or to plot a fitted psychometric function whose values of pse and jnd are known. Currently only working with probit and logit link function.

References

Moscatelli, A., Mezzetti, M., & Lacquaniti, F. (2012). Modeling psychophysical data at the population-level: The generalized linear mixed model. Journal of Vision, 12(11):26, 1-17. https://doi.org/10.1167/12.11.26 Knoblauch, K., & Maloney, L. T. (2012). Modeling psychophysical data in R (Vol. 32). Springer Science & Business Media.

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See Also

glm for for Generalized Linear Models. PsychFunction for estimation of PSE and JND.

Examples

```
y = c(0,1)

x = c(-40, 40)

plot(y \sim x, type = "n", bty = "n", lab = c(5,3,7))

PsychShape(pse = 0, jnd = 6, x.range = c(-40, 40), ps.col = "gray", ps.lwd = 3)

PsychShape(pse = 6, jnd = 6, x.range = c(-40, 40), ps.col = "black")

PsychShape(pse = 6, jnd = 6, x.range = c(-40, 40), ps.col = "red", ps.link = "logit", ps.lwd = 3)
```

PsySimulate

Simulate psychophysical data

Description

Given the arrays of fixed and random effects, as well as the covariance, and the characteristic of the simulated experiment (i.e.,) the function simulates a dataset in which for each subject the following information is provided: the slope and intercept value of the psychometric function, and the simulated responses to the stimulus levels that fit that function.

Usage

```
PsySimulate(fixeff = c(-7, 0.0875), raneff = c(2.4, -0.002, 2e-06), nsubjects = 8, pps = 9, ntrials = 40, xint = c(40, 120), constant = F)
```

Arguments

fixeff	Array of fixed effects. First item is the intercept, second item is the slope.	
raneff	Array of random effects. First item is the intercept, second item is the covar ance, third item is the slope.	
nsubjects	Number of subjects to simulate data for. Default is 8.	
pps	Number of stimulus levels. Default is 9.	
ntrials	Number of trials for each stimulus level. Default is 40.	
xint	Range of the stimulus interval. Default is $c(40,120)$	
constant	If set to FALSE, stimulus levels are randomly generated, uniformly distributed values within the selected interval. If constant = TRUE, the X interval is divided in intervals of constant length. Default is FALSE.	

Value

The simulated dataset

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Examples

```
#simulate dataset (one subject)
datafr.S1 <- PsySimulate(nsubject = 1, constant = TRUE)</pre>
```

simul_data

A simulated psychophysical dataset

Description

A dataset containing simulated data for 8 subjects. Created using PsySimulate(constant = T). The variables are as follows:

Usage

```
data(simul_data)
```

Format

A data frame with 72 rows and 6 variables:

X Samples of the X interval c(40,120)

Intercept Intercept of the psychometric function

Slope Slope of the psychometric function

Longer Number of trials in which response is judged "longer" than standard

Total Total number of trials for each sample of X interval

Subject Subject code (S1 to S8)

See Also

PsySimulate() for simulating dataframes.

vibro_exp3

Data from tactile discrimination task - EXP3

Description

A dataset containing the response and stimuli from a tactile discrimination task (nine participants). In a forced-choice experiment, participants were required to discriminate the motion speed of a moving surface by touching it. Simultaneously with the motion stimulus, a 32Hz masking vibration occurred in half of the trials.

Usage

```
data(vibro_exp3)
```

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Format

A data frame with 72 rows and 6 variables:

speed a numeric vector giving the motion speed in cm/s of the moving surface

vibration a numeric vector giving the vibration frequency in Hz of the masking stimulus. Either 32Hz or 0 (no vibration - control condition)

faster The proportion of trials in which the comparison stimulus was judged as faster than the reference

slower The proportion of trials in which the comparison stimulus was judged as slower than the reference

subject subject id

Source

Original data were published in Dallmann et al. (2015).

References

Dallmann, C. J., Ernst, M. O., & Moscatelli, A. (2015). The role of vibration in tactile speed perception. Journal of Neurophysiology, 114(6), 3131–3139. <doi:10.1152/jn.00621.2015>

xplode

Extract values from a fitted GLMM object

Description

Extract values from an object of class "merMod" (more specifically, from an object of subclass glmerMod).

Usage

```
xplode(model, name.cont = NA, name.factor = NA, names.response = NA,
  define.pf = list(pf1 = list(intercept = 1, slope = 2)))
```

model	The GLMM fitted with glmer. An object of class "merMod".
name.cont	A string providing the name of the continuous predictor, as in the formula object of the fitted model
name.factor	A string providing the name of name of the categorical predictor, as in the formula object of the fitted model
names.response	Optional. A string providing the name of name of the response variable, as in the formula object of the fitted model
define.pf	Optional. Specifies which parameter pertains to the intercept and which to the slope in the formula object.

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Details

For simplicity, several MixedPsy functions take as input an object of class xplode instead of an object of class "merMod". Most of these functions assume by default that the continuous predictor is entered first in the formula object. It is possible to use a different order, this requires to specify which parameter pertains to the intercept and which to the slope, by changing define.pf.

Value

An object of class "merMod" to be used with other MixedPsy functions.

See Also

merMod-class and glmer from package lme4 for objects of class "merMod". MixDelta, MixTreatment for use of objects of class xplode.obj.

Examples

```
library(lme4)
datafr = PsySimulate(nsubjects = 10)
mod1 = glmer(formula = cbind(Longer, Total - Longer) ~ X + (1 | Subject),
family = binomial(link = "probit"), data = datafr)
xplode.mod1 = xplode(model = mod1, name.cont = "X")
MixDelta(xplode.mod1)
```

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