

R documentation

of 'BANOVA.multi.mediation.Rd'

December 16, 2020

BANOVA.multi.mediation

Mediation analysis with multiple possibly correlated mediators

Description

BANOVA.multi.mediation is a function for analysis of multiple possibly correlated mediators. These mediators are assumed to have no causal influence on each other. Both single-level and multi-level models can be analyzed.

Usage

```
BANOVA.multi.mediation(sol_1, sol_2, xvar, mediators, individual = FALSE)
```

Arguments

sol_1	an object of class "BANOVA" returned by BANOVA.run function with a fitted model for an outcome variable regressed on a causal variable, a mediator, and, possibly, moderators and control variables. The outcome variable can follow Normal, T, Poisson, Bernoulli, Binomial, Truncated Normal and ordered Multinomial distributions.
sol_2	an object of class "BANOVA" returned by BANOVA.run function, which contains an outcome of the analysis for multiple Multivariate Normal mediators regressed on a causal variable and other possible moderators and control variables.
xvar	a character string that specifies the name of the causal variable used in both models.
mediators	a vector with character strings, which specifies the names of the mediator variables used in the models.
individual	logical indicator of whether to output effects for individual units in the analysis (TRUE or FALSE). This analysis requires a multilevel sol_1.

Details

The function extends BANOVA.mediation to the case with multiple possibly correlated mediators. For details about mediation analysis performed in BANOVA see the help page for the [BANOVA.mediation](#).

BANOVA.multi.mediation estimates and tests specific indirect effects of the causal variable conveyed through each mediator. Furthermore, the total indirect effect of the causal variables are computed as a sum of the specific indirect effects.

The function prints multiple tables with mediated effects. Tables with direct effects of the causal variable and mediators on the outcome variable, as well as direct effects of the causal variable on the mediators include a posterior mean and 95% credible intervals of the effects. Next, the function displays on the console tables with specific indirect effects and effect sizes of the mediators, followed by the TIE of the causal variable. These tables include the mean, 95% credible intervals, and two-sided Bayesian p-values.

Value

Returns an object of class "BANOVA.multi.mediation". The returned object is a list containing:

dir_effects	table or tables with the direct effect.
individual_direct	is returned if individual is set to TRUE and the causal variable is a within-subject variable. Contains a table or tables of the direct effect at the individual levels of the analysis
m1_effects	a list with tables of the effects of the mediator on the outcome
m2_effects	a list with tables of the effect of the causal variable on the mediator
indir_effects	tables of the indirect effect
individual_indirect	is returned if individual is set to TRUE and the mediator is a within-subject variable. Contains the table or tables with the indirect effect
effect_sizes	a list with effect sizes on individual mediators
total_indir_effects	table or tables with the total indirect effect of the causal variable
xvar	the name of the causal variable
mediators	the names of the mediating variables
individual	the value of the argument individual (TRUE or FALSE)

Author(s)

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Examples

```
# Use the colorad data set
data(colorad)
# Add a second mediator to the data set
colorad$blur_squared <- (colorad$blur)^2
# Prepare mediators to be analyzed in the Multivariate Normal model
mediators <- cbind(colorad$blur, colorad$blur_squared)
colnames(mediators) <- c("blur", "blur_squared")
colorad$mediators <- mediators

# Build and analyze the model for the outcome variable
```

```
model <- BANOVA.model('Binomial')
banova_binom_model <- BANOVA.build(model)
res_1 <- BANOVA.run(y ~ typic, ~ color + blur + blur_squared, fit = banova_binom_model,
                   data = colorad, id = 'id', num_trials = as.integer(16),
                   iter = 2000, thin = 1, chains = 2)
# Build and analyze the model for the mediators
model <- BANOVA.model('multiNormal')
banova_multi_norm_model <- BANOVA.build(model)
res_2 <- BANOVA.run(mediators ~ typic, ~ color, fit = banova_multi_norm_model,
                   data = colorad, id = 'id', iter = 2000, thin = 1, chains = 2)

# Calculate (moderated) effects of "typic" mediated by "blur" and "blur_squared"
results <- BANOVA.multi.mediation(res_1, res_2, xvar='typic', mediators=c("blur", "blur_squared"))
```

R documentation

of 'BANOVA.mediation.Rd'

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BANOVA.mediation

Mediation analysis based on BANOVA models

Description

BANOVA.mediation conducts mediation and moderated mediation analysis based on various BANOVA models with a single mediator and a casual variable.

Usage

```
BANOVA.mediation(sol_1, sol_2, xvar, mediator, individual = F,
  return_posterior_samples = F, multi_samples_beta1_raw_m = NULL)
## S3 method for class 'BANOVA.mediation'
print(x, ...)
```

Arguments

<code>sol_1</code>	an object of class "BANOVA" returned by BANOVA.run function with a fitted model for an outcome variable regressed on a causal variable, a mediator, and, possibly, moderators and control variables. The outcome variable can follow Normal, T, Poisson, Bernoulli, Binomial, and ordered Multinomial distributions.
<code>sol_2</code>	an object of class "BANOVA" returned by BANOVA.run function with a fitted model for a Normal outcome variable regressed on a causal variable, and, possibly, moderators and control variables.
<code>xvar</code>	a character string which specifies the name of the causal variable used in both models.
<code>mediator</code>	a character string which specifies the name of the mediator variable used in the model.
<code>individual</code>	logical indicator of whether to output effects for individual units in the analysis (TRUE or FALSE). This analysis requires a multilevel <code>sol_1</code> .
<code>return_posterior_samples</code>	logical indicator of whether posterior samples of mediated effects should be returned (TRUE or FALSE).
<code>x</code>	a BANOVA.mediation object
<code>...</code>	additional arguments, currently ignored
<code>multi_samples_beta1_raw_m</code>	argument for an internal use in the package. Please do not change.

Details

This function conducts a correlational mediation and moderated mediation analysis (Baron and Kenny 1986; Zao, Lynch and Chen 2010; Zhang, Wedel and Pieters 2008) based on BANOVA models. Based on the samples from posterior distributions, the function calculates the direct effect and indirect effect for which posterior means and 95% credible intervals are reported. The effect size of the indirect effect is computed as a generalized partial eta-squared. For details about this metric, see the publication of Wedel and Dong (2019).

When the algorithm is calculating the effects of a given causal variable it first identifies all moderators which are interacting with the investigated variable. Based on each interaction, moderated coefficients are computed and summarized in a table. If the causal variable is a part of an interaction term with three or more variables, separate results are computed for each of the moderators and all of their possible combinations. This results in multiple tables with the effects. If a continuous variable is involved in the interaction the effects are evaluated at its mean value, which is zero by default. This is equivalent to omitting the continuous variable from the interaction.

The function combines the effects of the mediator on the dependent variable with the effect of the causal variable on the mediator in a multiplicative manner to obtain the indirect effect of the treatment. If multiple tables with moderated effects of the mediator or the causal variable on mediator are obtained in the previous steps of the analysis, the indirect effects are computed for each combination of these table.

For models with a Normal outcome variable, it is possible to interpret the effects as causal by explicitly encoding the causal variable with dummy coding and including an interaction between the causal and mediating variables in the model. For further details, see the publication of MacKinnon et al. (2020).

Value

BANOVA.mediation returns an object of class "BANOVA.mediation". The returned object is a list containing:

<code>dir_effects</code>	tables of the direct effect
<code>individual_direct</code>	a table of the direct effect at the individual level if individual = T and the causal variable is a within-subject variable
<code>m1_effects</code>	tables of the effect of the mediator on the outcome
<code>m2_effects</code>	tables of the effect of the causal variable on the mediator
<code>indir_effects</code>	tables of the indirect effect
<code>individual_indirect</code>	the table of the indirect effect at the individual level if individual = T and the mediator is a within-subject variable
<code>effect_size</code>	a table with the effect size of the mediator
<code>xvar</code>	the name of the causal variable
<code>mediator</code>	the name of the mediating variable
<code>individual</code>	the value of the argument individual (TRUE or FALSE)

References

Baron, R.M., and Kenny, D.A. (1986) *Moderator Mediator Variables Distinction in Social Psychological Research: Conceptual, Strategic, and Statistical Considerations*, Journal of Personality and Social Psychology, Vol. 51, No. 6, pp. 1173-82.

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- Ying, Y. and MacKinnon, D.P. (2009) *Bayesian Mediation Analysis*, Psychological Methods, Vol. 14, No.4, pp. 301-322.
- Zhao, X., John G.L., and Chen, Q. (2010) *Reconsidering Baron and Kenny: Myths and Truths About Mediation Analysis*, Journal of Consumer Research, Vol.37, No.2, pp. 197-206.
- Wedel, M., and Dong, C. (2019) *BANOVA: Bayesian Analysis of Variance for Consumer Research*, Journal of Consumer Psychology, Vol. 30, No. 1, pp. 3-23.
- MacKinnon, D.P., Valente, M.J., and Gonzalez, O. (2020) *The correspondence between causal and traditional mediation analysis: the link is the mediator by treatment interaction*, Prevention Science, Vol. 21, No. 2, pp. 147-157.

Examples

```
data(condstudy_sub)

# use BANOVA.run based on 'Stan'
model <- BANOVA.model('Normal')
banova_model <- BANOVA.build(model)
res_1 <- BANOVA.run(att~cond+pict, ~type, fit = banova_model, data = condstudy_sub,
                    id = 'id', iter = 500, thin = 1, chains = 2)
res_2 <- BANOVA.run(pict~cond, ~type, fit = banova_model, data = condstudy_sub,
                    id = 'id', iter = 500, thin = 1, chains = 2)

# (moderated) mediation
sol <- BANOVA.mediation(res_1, res_2, xvar='cond', mediator='pict')
print(sol)
print(sol$dir_effects)
```