Package 'MedSurvey'

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Title Mediation Analysis for Complex Surveys
Version 1.1.0
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Description It is a computer tool to conduct mediation analysis for complex surveys using multi-stage sampling. Specifically, the mediation analysis method using balanced repeated replications was proposed by Mai, Ha, and Soulakova (2019) <doi:10.1080 10705511.2018.1559065="">. The development of 'MedSurvey' was sponsored by American Lebanese Syrian Associated Charities (ALSAC). However, the contents of MedSurvey do not necessarily represent the policy of the ALSAC</doi:10.1080>
Depends R (>= 2.50), stats, Matrix, survey, lavaan, parallel
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Encoding UTF-8
LazyData true
<pre>URL https://CRAN.R-project.org/package=MedSurvey</pre>
https://github.com/YujiaoMai/MedSurvey
https://sjbiostat.shinyapps.io/MedSurvey/
RoxygenNote 6.1.1
R topics documented:
chisq.BRR

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Description

This function is used to adjust model fit statistics for complex surveys with balanced repeated replications (*Oberski*, 2014; *Satorra & Muthen*, 1995). It saves time to only obtain the model fit statistics during the model selection stage.

Usage

```
chisq.BRR(model, lavaan.fit, data, mwgtname, repwgtnames,
  fayfactor = 0.5, estimator = c("ML"), test = c("satorra.bentler"))
```

Arguments

model	The model being fitted. It is written in lavaan model syntax (Rosseel, 2012).
lavaan.fit	The model fit results using 'ML' estimator with sample main weights, but without adjusting the fit statistics or standard errors for complex surveys. Note that it is a lavaan object.
data	The raw data including the variables of interest and the survey weights. It should be a dataset or dataframe.
mwgtname	The variable name indicating the sample main weight in the dataset. See balanced repeated replications method (<i>Wolter</i> , 2007) for more information about the main weight.
repwgtnames	The variable names indicating the set of replicate weights in the dataset. See balanced repeated replications method (<i>Wolter</i> , 2007) for more information about the replicate weight.
fayfactor	The fayfactor used in the standard error calculation by fay's method (Fay & Train, 1995; Judkins, 1990) for balanced repeated replications. Fayfactor is a value between 0 and 1. The default is 0.5.
estimator	The method used to estimate the model. 'ML' is the default option and the only available option for current version. It is not required.
test	The method used to generate adjusted standard errors. 'satorra.bentler' is the default option and the only available option for current version. It is not required.

Value

The model fit results as a lavaan object (Rosseel, 2012) with the adjusted model fit statistics.

References

Fay, R. E., & Train, G. F. (1995). Aspects of survey and model-based postcensal estimation of income and poverty characteristics for states and counties. In Proceedings of the Section on Government Statistics, American Statistical Association, Alexandria, VA (pp. 154-159).

Judkins, D. R. (1990). Fay's method for variance estimation. Journal of Official Statistics, 6(3), 223-239.

Oberski, D. (2014). lavaan. survey: An R package for complex survey analysis of structural equation models. Journal of Statistical Software, 57(1), 1-27. DOI:10.18637/jss.v057.i01

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Rosseel, Y. (2012). Lavaan: An R package for structural equation modeling and more. Version 0.5–12 (BETA). Journal of statistical software, 48(2), 1-36. DOI:10.18637/jss.v048.i02

Satorra, A., & Muthen, B. (1995). Complex sample data in structural equation modeling. Sociological methodology, 25(1), 267-316.

Wolter, K. (2007). Introduction to variance estimation. New York, NY: Springer.

Examples

```
R <- 160
wgtnames <- paste("repwgt", seq(0,R,by=1), sep="")</pre>
mwgtname=wgtnames[1]
repwgtnames=wgtnames[2:(R+1)]
fayfactor=0.5
model3 <- ' # outcome</pre>
            numcg ~ u0*1 + c*workban + b1*sp_adltban + b2*sp_kidsban
           # mediator
              sp_adltban ~ u1*1 + a1*workban
              sp_kidsban ~ u2*1 + a2*workban
           # indirect effect (a*b)
              a1b1 := a1*b1
             a2b2 := a2*b2
           # total effect
             total := c + (a1*b1) + (a2*b2)
fit <- lavaan::sem(model=model3, data=MedData, estimator='ML', test='standard')</pre>
chisq.BRR(model3,fit,MedData,mwgtname, repwgtnames)
#
# MedSurvey 1.1.0 Adjusted Model Fit Statistics using BRR
#
# chisa
          df
               pvalue
                          CFI
                                   RMSEA
                                               SRMR
                                                            AIC
                                                                       BIC
# 305.25
          1 0.00000
                         0.40561 0.27852
                                            0.07416
                                                       88699.43
                                                                  88768.45
```

med.fit.BRR

Estimate the mediation effects and standard errors adjusting for complex surveys with BRR

Description

This function is used to estimate the mediation effects adjusted for complex surveys with balanced repeated replications (BRR) (Mai, Ha, Soulakova, 2019).

Usage

```
med.fit.BRR(model = NULL, data = NULL, mwgtname = NULL,
  repwgtnames = NULL, fayfactor = 0.5, estimator = c("ML"),
  test = c("satorra.bentler"), parallel = c("no", "parallel", "snow"),
  ncore = Sys.getenv("NUMBER_OF_PROCESSORS"), cl = NULL, ...)
```

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Arguments

model The model being fitted. It is written in lavaan model syntax (Rosseel, 2012). data The raw data including the variables of interest and the survey weights. It should be a dataset or dataframe. The variable name indicating the sample main weight in the dataset. See balmwgtname anced repeated replications method (Wolter, 2007) for more information about the main weight. The variable names indicating the set of replicate weights in the dataset. See balrepwgtnames anced repeated replications method (Wolter, 2007) for more information about the replicate weight. fayfactor The fayfactor used in the standard error calculation by fay's method (Fay & Train, 1995; Judkins, 1990) for balanced repeated replications. Fayfactor is a value between 0 and 1. The default is 0.5. estimator The method used to estimate the model. 'ML' is the default option and the only available option for current version. It is not required. The method used to generate adjusted standard errors. 'satorra.bentler' is the test default option and the only available option for current version. It is not required. Parallel computing ("no" or "parallel" or "snow"). It is "no" by default, parallel which means it will not use parallel computing. The option "parallel" is to use multiple cores in a computer for parallel computing. It is used with the number of cores (ncore). The option "snow" is to use clusters for parallel computing. It is used with the number of clusters (cl). Number of processors used for parallel computing. By default, ncore = Sys.getenv ncore ('NUMBER_OF_PROCESSORS'). cl Number of clusters. It is NULL by default. When it is NULL, the program will detect the number of clusters automatically. Extra arguments. For example, ordered=c('z1','z2') is an argument to tell 'z1' and 'z2' are ordinal variables. It is not required.

Value

The model fit results as a lavaan object with the adjusted estimates, standard errors, and model fit statistics. It is a lavaan object (Rosseel, 2012).

References

Fay, R. E., & Train, G. F. (1995). Aspects of survey and model-based postcensal estimation of income and poverty characteristics for states and counties. In Proceedings of the Section on Government Statistics, American Statistical Association, Alexandria, VA (pp. 154-159).

Judkins, D. R. (1990). Fay's method for variance estimation. Journal of Official Statistics, 6(3), 223-239.

Mai, Y., Ha, T., & Soulakova, J. N. (2019). Multimediation Method With Balanced Repeated Replications For Analysis Of Complex Surveys. Structural Equation Modeling: A Multidisciplinary Journal. DOI:10.1080/10705511.2018.1559065

Rosseel, Y. (2012). Lavaan: An R package for structural equation modeling and more. Version 0.5–12 (BETA). Journal of statistical software, 48(2), 1-36. DOI:10.18637/jss.v048.i02

Wolter, K. (2007). Introduction to variance estimation. New York, NY: Springer.

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Examples

```
R <- 160
wgtnames <- paste("repwgt", seq(0,R,by=1), sep="")</pre>
mwgtname=wgtnames[1]
repwgtnames=wgtnames[2:(R+1)]
model2 <- ' # outcome</pre>
             numcg ~ u0*1 + c*workban + b1*sp_adltban + b2*sp_kidsban
            # mediator
              sp_adltban ~ u1*1 + a1*workban
              sp_kidsban ~ u2*1 + a2*workban
            #covariance of residuals
              sp_adltban \sim sp_kidsban
            # indirect effect (a*b)
              a1b1 := a1*b1
              a2b2 := a2*b2
            # total effect
              total := c + (a1*b1) + (a2*b2)
fit.BRR <- med.fit.BRR(model=model2, data=MedData, mwgtname=mwgtname,</pre>
         repwgtnames=repwgtnames, fayfactor=0.5, parallel='parallel', ncore=2)
lavaan::summary(fit.BRR)
#
\# lavaan 0.6-3 ended normally after 41 iterations
# Optimization method
                                                NLMINB
# Number of free parameters
                                                    12
# Number of observations
                                                  3922
# Estimator
                                                    ML
                                                            Robust
# Model Fit Test Statistic
                                                 0.000
                                                             0.000
# Degrees of freedom
                                                                 0
                                       0.0000000000000
# Minimum Function Value
# Scaling correction factor
                                                                NA
# for the Satorra-Bentler correction
# Parameter Estimates:
#
   Information
                                                Expected
# Information saturated (h1) model
                                           Structured
# Standard Errors
                                                   BRR
# Regressions:
                     Estimate Std.Err z-value P(>|z|)
#
# numcg ~
#
    workban (c)
                     -0.101
                                0.039
                                        -2.572
                                                  0.010
    sp_adltbn (b1)
                      -0.253
                                0.048
                                        -5.270
                                                  0.000
#
    sp_kidsbn (b2)
                                0.051
                                        -7.006
                                                  0.000
#
                     -0.361
# sp_adltban ~
   workban (a1)
                       0.069
                                0.018
                                         3.753
                                                  0.000
# sp_kidsban ~
#
     workban (a2)
                       0.020
                                0.016
                                         1.250
                                                  0.211
```

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```
# Covariances:
#
                    Estimate Std.Err z-value P(>|z|)
  .sp_adltban \sim\sim
#
#
    .sp_kidsban
                       2.784
                                0.195
                                        14.300
                                                  0.000
#
# Intercepts:
                    Estimate Std.Err z-value P(>|z|)
#
   .numcg
              (u0)
                    18.485
                               0.566 32.668
                                                 0.000
   .sp_adltbn (u1)
#
                      4.221
                               0.167
                                       25.281
                                                 0.000
                               0.143
                                       55.272
                                                 0.000
   .sp_kidsbn (u2)
                      7.926
# Variances:
                    Estimate Std.Err z-value P(>|z|)
#
#
                     54.283
                             1.716 31.628
   .numcg
                                                 0.000
#
                     11.011
                               0.239
                                       46.140
                                                 0.000
   .sp_adltban
#
                      9.402
                              0.209
   .sp_kidsban
                                       44.998
                                                 0.000
#
# Defined Parameters:
#
                    Estimate Std.Err z-value P(>|z|)
#
    a1b1
                     -0.017
                               0.006
                                       -2.905
                                                 0.004
#
    a2b2
                     -0.007
                               0.006
                                       -1.234
                                                 0.217
#
    total
                     -0.125
                               0.040
                                       -3.169
                                                 0.002
```

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To adjust the p values for multimediation tests

Description

This function is used to adjust the p values when there are multiple mediators (Mai et al., 2019).

Usage

```
med.p.adjust(fit = NULL, med.eff = NULL, p.adj.method = c("holm",
   "hochberg", "hommel", "bonferroni", "BH", "BY", "fdr"))
```

Arguments

fit	The model fit results of a model with multiple mediators. Note that it is a lavaan object.
med.eff	A vector of labels. The labels should be of the mediation effects in the estimated model.
p.adj.method	The method used to adjust for multiplicity ('holm' or 'hochberg' or 'hommel' or 'bonferroni' or 'BH' or 'BY' or 'fdr'). Conservative method includes the Bonferroni correction ('bonferroni') in which the p-values are multiplied by the number of comparisons. Less conservative corrections are also included by Holm (1979) ('holm'), Hochberg (1988) ('hochberg'), Hommel (1988) ('hommel'), Benjamini & Hochberg (1995) ('BH' or its alias 'fdr'), and Benjamini & Yekutieli (2001) ('BY'), respectively. It is 'holm' by default. It is not required.

Value

The adjusted p values along with the effect labels and original p values. It is a list.

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References

Benjamini, Y., & Hochberg, Y. (1995). Controlling the false discovery rate: a practical and powerful approach to multiple testing. Journal of the Royal Statistical Society Series B, 57, 289–300. DOI:10.2307/2346101

Benjamini, Y., & Yekutieli, D. (2001). The control of the false discovery rate in multiple testing under dependency. Annals of Statistics, 29, 1165–1188. DOI:10.1214/aos/1013699998

Holm, S. (1979). A simple sequentially rejective multiple test procedure. Scandinavian Journal of Statistics, 6, 65–70.

Hommel, G. (1988). A stagewise rejective multiple test procedure based on a modified Bonferroni test. Biometrika, 75, 383–386. DOI:10.1093/biomet/75.2.383

Hochberg, Y. (1988). A sharper Bonferroni procedure for multiple tests of significance. Biometrika, 75, 800–803. DOI:10.1093/biomet/75.4.800

Rosseel, Y. (2012). Lavaan: An R package for structural equation modeling and more. Version 0.5–12 (BETA). Journal of statistical software, 48(2), 1-36. DOI:10.18637/jss.v048.i02

Shaffer, J. P. (1995). Multiple hypothesis testing. Annual Review of Psychology, 46, 561–576.

Sarkar, S. (1998). Some probability inequalities for ordered MTP2 random variables: a proof of Simes conjecture. Annals of Statistics, 26, 494–504. DOI:10.1214/aos/1028144846

Examples

```
R <- 160
wgtnames <- paste("repwgt", seq(0,R,by=1), sep="")</pre>
mwgtname=wgtnames[1]
repwgtnames=wgtnames[2:(R+1)]
fayfactor=0.5
model2 <- ' # outcome</pre>
              numcg ~ u0*1 + c*workban + b1*sp_adltban + b2*sp_kidsban
              sp_adltban ~ u1*1 + a1*workban
              sp_kidsban ~ u2*1 + a2*workban
            #covariance of residuals
              sp_adltban ~~ sp_kidsban
            # indirect effect (a*b)
              a1b1 := a1*b1
              a2b2 := a2*b2
            # total effect
              total := c + (a1*b1) + (a2*b2)
fit.BRR2 <- med.fit.BRR(model=model2, data=MedData, mwgtname=mwgtname,</pre>
             repwgtnames=repwgtnames, fayfactor, parallel='parallel', ncore=4)
temp <- med.p.adjust(fit=fit.BRR2, med.eff=c('a1b1' , 'a2b2'))</pre>
# Adjustment for multi mediation tests:
#
#
       Effect
                        p Value
                                     adj.p Value
#
        a1b1
                  0.003667674
                                    0.007335347
#
        a2b2
                  0.217228711
                                    0.217228711
# NOTE:
        p Value adjustment method is holm
```

8 med.summary

med.summary

To print the summary results of the mediation analysis

Description

This function is used to print the summary results of the mediation analysis with adjustment for multiplicity.

Usage

```
med.summary(fit = NULL, med.eff = NULL, p.adj.method = c("holm",
   "hochberg", "hommel", "bonferroni", "BH", "BY", "fdr"))
```

Arguments

fit The model fit results of a mediation model. Note that it is a lavaan object.

med.eff A vector of labels. The labels should be of the mediation effects in the estimated

model.

p.adj.method The method used to adjust for multiplicity ('holm' or 'hochberg' or 'hommel'

or 'bonferroni' or 'BH' or 'BY' or 'fdr'). Conservative method includes the Bonferroni correction ('bonferroni') in which the p-values are multiplied by the number of comparisons. Less conservative corrections are also included by Holm (1979) ('holm'), Hochberg (1988) ('hochberg'), Hommel (1988) ('hommel'), Benjamini & Hochberg (1995) ('BH' or its alias 'fdr'), and Benjamini & Yekutieli (2001) ('BY'), respectively. It is 'holm' by default. It is not required.

Value

A list including the effect labels, estimates, standard errors, p values, and adjusted p values if there are more than one mediation effects.

References

Benjamini, Y., & Hochberg, Y. (1995). Controlling the false discovery rate: a practical and powerful approach to multiple testing. Journal of the Royal Statistical Society Series B, 57, 289–300. DOI:10.2307/2346101

Benjamini, Y., & Yekutieli, D. (2001). The control of the false discovery rate in multiple testing under dependency. Annals of Statistics, 29, 1165–1188. DOI:10.1214/aos/1013699998

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Holm, S. (1979). A simple sequentially rejective multiple test procedure. Scandinavian Journal of Statistics, 6, 65–70.

Hommel, G. (1988). A stagewise rejective multiple test procedure based on a modified Bonferroni test. Biometrika, 75, 383–386. DOI:10.1093/biomet/75.2.383

Hochberg, Y. (1988). A sharper Bonferroni procedure for multiple tests of significance. Biometrika, 75, 800–803. DOI:10.1093/biomet/75.4.800

Mai, Y., Ha, T., & Soulakova, J. N. (2019). Multimediation Method With Balanced Repeated Replications For Analysis Of Complex Surveys. Structural Equation Modeling: A Multidisciplinary Journal. DOI:10.1080/10705511.2018.1559065

Rosseel, Y. (2012). Lavaan: An R package for structural equation modeling and more. Version 0.5–12 (BETA). Journal of statistical software, 48(2), 1-36. DOI:10.18637/jss.v048.i02

Shaffer, J. P. (1995). Multiple hypothesis testing. Annual Review of Psychology, 46, 561–576.

Sarkar, S. (1998). Some probability inequalities for ordered MTP2 random variables: a proof of Simes conjecture. Annals of Statistics, 26, 494–504. DOI:10.1214/aos/1028144846

Examples

```
R <- 160
wgtnames <- paste("repwgt", seq(0,R,by=1), sep="")</pre>
mwgtname=wgtnames[1]
repwgtnames=wgtnames[2:(R+1)]
fayfactor=0.5
model2 <- ' # outcome</pre>
              numcg \sim u0*1 + c*workban + b1*sp_adltban + b2*sp_kidsban
            # mediator
              sp_adltban ~ u1*1 + a1*workban
              sp_kidsban ~ u2*1 + a2*workban
            #covariance of residuals
              sp_adltban ~~ sp_kidsban
            # indirect effect (a*b)
              a1b1 := a1*b1
              a2b2 := a2*b2
            # total effect
              total := c + (a1*b1) + (a2*b2)
fit.BRR2 <- med.fit.BRR(model=model2, data=MedData, mwgtname=mwgtname,</pre>
             repwgtnames=repwgtnames, fayfactor, parallel='parallel')
temp <- med.summary(fit=fit.BRR2, med.eff=c('a1b1' , 'a2b2'))</pre>
#
# MedSurvey 1.1.0
#
# Multimediation with Complex Survey Data:
#
   Effect
#
                        Est.
                                      BRR SE.
                                                        p Value
                                                                      adj.p Value
#
             -0.017475544
                                0.006014820
                                                  0.003667674
                                                                    0.007335347
#
   a1b1
#
   a2b2
             -0.007244189
                                0.005870823
                                                  0.217228711
                                                                    0.217228711
# NOTE:
    p Value adjustment method is holm
   Standard errors type is BRR SE.
```

```
# To catch the unformatted results:
temp
# $med.label
# [1] "a1b1" "a2b2"
# $med.est
# [1] -0.017475544 -0.007244189
# $med.se
# [1] 0.006014820 0.005870823
# $org.p.value
# [1] 0.003667674 0.217228711
# $adj.p.value
# [1] 0.007335347 0.217228711
# $se.type
# [1] "BRR SE."
# $p.adj.method
# [1] "holm"
```

MedData

CPS-TUS data 2014-2015

Description

Data from 2014-15 CPS Tobacco Use Supplement (TUS; U.S. Department of Commerce and U.S. Census Bureau 2016), employed adult daily smokers (Non-Hispanic White males only). Missing data are removed from the dataset. Due to the CRAN limitation of the size (5MB) of an R package, only half of the observations remained in this internal dataset for the purpose of illustration.

Usage

MedData

Format

A data frame with 3922 rows and 167 variables:

```
PRTAGE Age
PESEX Gender, 0=Male, 1=Female
repwgt0 Sample main weights
repwgt1 BRR replicate weights
repwgt2 BRR replicate weights
```

repwgt3 BRR replicate weights
repwgt4 BRR replicate weights
repwgt5 BRR replicate weights
repwgt6 BRR replicate weights
repwgt7 BRR replicate weights
repwgt8 BRR replicate weights
repwgt9 BRR replicate weights
repwgt10 BRR replicate weights
repwgt11 BRR replicate weights
repwgt12 BRR replicate weights
repwgt13 BRR replicate weights
repwgt14 BRR replicate weights
repwgt15 BRR replicate weights
repwgt16 BRR replicate weights
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repwgt33 BRR replicate weights
repwgt34 BRR replicate weights
repwgt35 BRR replicate weights
repwgt36 BRR replicate weights
repwgt37 BRR replicate weights
repwgt38 BRR replicate weights
repwgt39 BRR replicate weights
repwgt40 BRR replicate weights
repwgt41 BRR replicate weights
repwgt42 BRR replicate weights

MedData MedData

repwgt43	BRR replicate weights
repwgt44	BRR replicate weights
repwgt45	BRR replicate weights
repwgt46	BRR replicate weights
repwgt47	BRR replicate weights
repwgt48	BRR replicate weights
repwgt49	BRR replicate weights
repwgt50	BRR replicate weights
repwgt51	BRR replicate weights
repwgt52	BRR replicate weights
repwgt53	BRR replicate weights
repwgt54	BRR replicate weights
repwgt55	BRR replicate weights
repwgt56	BRR replicate weights
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repwgt58	BRR replicate weights
repwgt59	BRR replicate weights
repwgt60	BRR replicate weights
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repwgt159 BRR replicate weights
repwgt160 BRR replicate weights
sp_adltban Supporting toward smoking ban in adults-exclusive areas (e.g., casinos, bars)
sp_kidsban Supporting toward smoking ban in kids-related areas (e.g., children playground)
workban Score of smoking restriction score at workplace
numcg Number of cigarretes smoked per day
```

Source

http://thedataweb.rm.census.gov/ftp/cps_ftp.html#cpssupps

References

U.S. Department of Commerce, & U.S. Census Bureau. (2016). National Cancer Institute and Food and Drug Administration co-sponsored Tobacco Use Supplement to the Current Population Survey. 2014-15.

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