Package 'MedSurvey'

October 12, 2022

Title Linear Mediation Analysis for Complex Surveys Using Balanced Repeated Replication

Version 1.1.1.3.0

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Description It is a computer tool to conduct linear mediation analysis for complex surveys using multi-stage sampling and Balanced Repeated Replication (BRR). Specifically, the mediation analysis method using balanced repeated replications was proposed by Mai, Ha, and Soulakova (2019) <DOI:10.1080/10705511.2018.1559065>. The current version can only handle continuous mediators and outcomes. 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.

Depends R (>= 2.50), stats, Matrix, survey, lavaan, parallel

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Encoding UTF-8

LazyData true

URL https://CRAN.R-project.org/package=MedSurvey

RoxygenNote 7.1.1

NeedsCompilation no

Repository CRAN

Date/Publication 2020-10-05 07:00:06 UTC

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chisq.BRR

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 with- out 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 bal- anced 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 bal- anced 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 (<i>Fay & Train, 1995; Judkins, 1990</i>) 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.

chisq.BRR

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

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.

```
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
#
         df
                                   RMSEA
                                              SRMR
                                                            AIC
                                                                      BIC
# chisq
               pvalue
                         CFI
#
# 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,
    ...
)
```

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.
mwgtname	The variable name indicating the sample main weight in the dataset. See bal- anced 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 bal- anced 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 (<i>Fay & Train, 1995; Judkins, 1990</i>) 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.

parallel	Parallel computing ("no" or "parallel" or "snow"). It is "no" by default, 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 (<i>ncore</i>). The option "snow" is to use clusters for parallel computing. It is used with the number of clusters (<i>cl</i>).
ncore	Number of processors used for parallel computing. By default, ncore = Sys.getenv ('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 to be incorporated with in the future. 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.

```
R <- 160
wgtnames <- paste("repwgt", seq(0,R,by=1), sep="")
mwgtname=wgtnames[1]
repwgtnames=wgtnames[2:(R+1)]
model2 <- ' # outcome
        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 ~~ sp_kidsban
        # indirect effect (a*b)
        a1b1 := a1*b1</pre>
```

```
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
#
                                                 NLMINB
# Optimization method
# Number of free parameters
                                                     12
#
# Number of observations
                                                   3922
#
# Estimator
                                                     ML
                                                             Robust
                                                              0.000
# Model Fit Test Statistic
                                                  0.000
# Degrees of freedom
                                                                  0
                                                      0
# Minimum Function Value
                                        0.000000000000
# 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)
                      -0.361
                                0.051
                                         -7.006
                                                   0.000
#
# sp_adltban ~
     workban
                       0.069
                                0.018
                                          3.753
                                                   0.000
#
               (a1)
# sp_kidsban ~
    workban
#
               (a2)
                       0.020
                                0.016
                                          1.250
                                                   0.211
#
 Covariances:
#
                     Estimate Std.Err z-value P(>|z|)
#
#
  .sp_adltban ~~
#
     .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
#
#
    .sp_kidsbn (u2)
                       7.926
                                0.143
                                         55.272
                                                   0.000
#
# Variances:
#
                     Estimate Std.Err z-value P(>|z|)
#
    .numcg
                      54.283
                                1.716
                                         31.628
                                                   0.000
```

med.p.adjust

#	.sp_adltban	11.011	0.239	46.140	0.000
#	.sp_kidsban	9.402	0.209	44.998	0.000
#					
#	Defined Paramete	ers:			
#		Estimate	Std.Err	z-value	P(> z)
# #	a1b1	Estimate -0.017	Std.Err 0.006	z-value -2.905	P(> z) 0.004
	a1b1 a2b2				

med.p.adjust

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.

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

```
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
            # 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', 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
```

med.summary

```
#
       a1b1
                0.003667674
                               0.007335347
#
       a2b2
                0.217228711
                               0.217228711
#
# NOTE:
       p Value adjustment method is holm
#
*****
# To catch the unformatted results:
temp
#
# $med.eff
# [1] "a1b1" "a2b2"
#
# $org.p.value
# [1] 0.003667674 0.217228711
#
# $adj.p.value
# [1] 0.007335347 0.217228711
```

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

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

```
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
            # 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
                                    BRR SE.
                                                                   adj.p Value
#
                      Est.
                                                      p Value
#
            -0.017475544
                                                                 0.007335347
#
   a1b1
                              0.006014820
                                                0.003667674
#
   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 repwgt17 BRR replicate weights
- repwgt18 BRR replicate weights
- repwgt19 BRR replicate weights
- repwgt20 BRR replicate weights
- repwgt21 BRR replicate weights
- repwgt22 BRR replicate weights
- repwgt23 BRR replicate weights
- repwgt24 BRR replicate weights

repwgt25 BRR replicate weights repwgt26 BRR replicate weights repwgt27 BRR replicate weights repwgt28 BRR replicate weights repwgt29 BRR replicate weights repwgt30 BRR replicate weights repwgt31 BRR replicate weights repwgt32 BRR replicate weights 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 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 repwgt57 BRR replicate weights repwgt58 BRR replicate weights repwgt59 BRR replicate weights repwgt60 BRR replicate weights repwgt61 BRR replicate weights

MedData

repwgt62 BRR replicate weights repwgt63 BRR replicate weights repwgt64 BRR replicate weights repwgt65 BRR replicate weights repwgt66 BRR replicate weights repwgt67 BRR replicate weights repwgt68 BRR replicate weights repwgt69 BRR replicate weights repwgt70 BRR replicate weights repwgt71 BRR replicate weights repwgt72 BRR replicate weights repwgt73 BRR replicate weights repwgt74 BRR replicate weights repwgt75 BRR replicate weights repwgt76 BRR replicate weights repwgt77 BRR replicate weights repwgt78 BRR replicate weights repwgt79 BRR replicate weights repwgt80 BRR replicate weights repwgt81 BRR replicate weights repwgt82 BRR replicate weights repwgt83 BRR replicate weights repwgt84 BRR replicate weights repwgt85 BRR replicate weights repwgt86 BRR replicate weights repwgt87 BRR replicate weights repwgt88 BRR replicate weights repwgt89 BRR replicate weights repwgt90 BRR replicate weights repwgt91 BRR replicate weights repwgt92 BRR replicate weights repwgt93 BRR replicate weights repwgt94 BRR replicate weights repwgt95 BRR replicate weights repwgt96 BRR replicate weights repwgt97 BRR replicate weights repwgt98 BRR replicate weights

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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

https://cancercontrol.cancer.gov/brp/tcrb/tus-cps/questionnaires-data

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