

Package ‘ImpactIV’

February 19, 2015

Type Package

Title Identifying Causal Effect for Multi-Component Intervention Using Instrumental Variable Method

Version 1.0

Date 2010-12-12

Author Peng Ding <dingyunyiqu@163.com>

Maintainer Peng Ding <dingyunyiqu@163.com>

Description In this package, you can find two functions proposed in Ding, Geng and Zhou (2011) to estimate direct and indirect causal effects with randomization and multiple-component intervention using instrumental variable method.

Depends nnet

License GPL (>= 2)

LazyLoad yes

Repository CRAN

Date/Publication 2012-10-29 08:57:11

NeedsCompilation no

R topics documented:

ImpactIV-package	2
heter_IV2	3
homo_IV1	5
impact	6

Index	8
--------------	----------

ImpactIV-package

Identifying Causal Effect for Multi-Component Intervention Using IV

Description

In this package, you can find two functions proposed in Ding, Geng and Zhou (2011) to estimate direct and indirect causal effects with randomization and multiple-component intervention using instrumental variable method.

Details

Package:	ImpactIV
Type:	Package
Version:	1.0
Date:	2010-12-12
License:	GPL (>=2)
LazyLoad:	yes

Author(s)

Maintainer: Peng Ding <dingyunyiqu@163.com>

References

Ding, P., Geng, Z. and Zhou, X. H. (2011). Identifying Causal Effect for Multi-Component Intervention Using Instrumental Variable Method: with A Case Study of IMPACT Data. Technical Report.

See Also

homo_IV1, heter_IV2

Examples

```
data(impact)
Z=impact$Z
A=impact$A
M=impact$M
Y=scale(impact$Y)
X=as.matrix(impact[,5:12])
##continuos variables of X
Xcon = X[, c(1,4,6,8)]
##discrete variables of X
```

```

Xdis = X[, c(2,3,5,7)]
##X^2
X2 = cbind(X, poly(Xcon, degree = 2, raw = TRUE),
           Xcon*Xdis[,1], Xcon*Xdis[,2], Xcon*Xdis[,3], Xcon*Xdis[,4])

method1 = homo_IV1(Z = Z,A = A,M = M,Y = Y,X = X)
method2 = heter_IV2(Z = Z,A = A,M = M,Y = Y,X = X2,
                    polydegree = 1, step1 = method1,
                    truncate = 0.25, select = "AIC")

```

heter_IV2

Estimation causal effect under Assumption 7 in Ding et al. (2011)

Description

Estimation causal effect under Assumption 6 in Ding et al. (2011) when the second order moment of the error term is not constant.

Usage

```

heter_IV2(Z, A, M, Y, X, polydegree = 2, step1 = NULL,
          truncate = 0.25, select = NULL)

```

Arguments

Z	A vector of the randomization variable.
A	A vector of the first mediator: whether a patient receives antidepressant medication.
M	A vector of the second mediator: whether a patient receives mental health therapy.
Y	A vector of the outcome of interest.
X	A matrix of all the covariates.
polydegree	The order of the polynomial function.
step1	The result of the first step estimation from homo_IV1.
truncate	Truncate the estimated Omega using a positive constant.
select	Using AIC or BIC for variable selection in the polynomial regression, the default is null.

Details

For background of the problem, see Ding et al. (2011).

Value

beta	beta coefficients of Z, A, M and AM.
phat	proportion of randomization to the treatment group.
residual	residuals of the first step regression.
se	standard errors of beta coefficients.
zvalue	z-values of the beta coefficients.
pvalue	p-values of the beta coefficients.
CI	confidence intervals of the beta coefficients.
COV	covariance matrix of the beta coefficients.
ser	robust version of standard errors of beta coefficients.
zvaluer	robust version of z-values of the beta coefficients.
pvaluer	robust version of p-values of the beta coefficients.
CIr	robust version of confidence intervals of the beta coefficients.
COVr	robust version of covariance matrix of the beta coefficients.
N	sample size
G	G is defined in Ding et al. (2010).
W	W is defined in Ding et al. (2010).
Omegahat	Omegahat is is defined in Ding et al. (2010).

Author(s)

Peng Ding <dingyunyiqu@163.com>

References

Ding, P., Geng, Z. and Zhou, X. H. (2011). Identifying Causal Effect for Multi-Component Intervention Using Instrumental Variable Method: with A Case Study of IMPACT Data. Technical Report.

Examples

```
##See help for "ImpactIV"
```

homo_IV1

*Estimation causal effect under Assumption 6 in Ding et al. (2011)***Description**

Estimation causal effect under Assumption 6 in Ding et al. (2011) when the second order moment of the error term is constant.

Usage

```
homo_IV1(Z, A, M, Y, X)
```

Arguments

Z	A vector of the randomization variable.
A	A vector of the first mediator: whether a patient receives antidepressant medication.
M	A vector of the second mediator: whether a patient receives mental health therapy.
Y	A vector of the outcome of interest.
X	A matrix of all the covariates.

Details

For background of the problem, see Ding et al. (2011).

Value

beta	beta coefficients of Z, A, M and AM.
phat	proportion of randomization to the treatment group.
residual	residuals of the regression.
se	standard errors of beta coefficients.
zvalue	z-values of the beta coefficients.
pvalue	p-values of the beta coefficients.
CI	confidence intervals of the beta coefficients.
COV	covariance matrix of the beta coefficients.
ser	robust version of standard errors of beta coefficients.
zvaluer	robust version of z-values of the beta coefficients.
pvaluer	robust version of p-values of the beta coefficients.
CIr	robust version of confidence intervals of the beta coefficients.
COVr	robust version of covariance matrix of the beta coefficients.
N	sample size
G	G is defined in Ding et al. (2010).
W	W is defined in Ding et al. (2010).
Omega	Omega is defined in Ding et al. (2010).

Author(s)

Peng Ding <dingyunyiqu@163.com>

References

Ding, P., Geng, Z. and Zhou, X. H. (2011). Identifying Causal Effect for Multi-Component Intervention Using Instrumental Variable Method: with A Case Study of IMPACT Data. Technical Report.

Examples

```
##See help for "ImpactIV"
```

impact

The data for IMPACT program.

Description

The data for IMPACT program used in Ding et al. (2011).

Usage

```
data(impact)
```

Format

A data frame with 1783 observations on the following 12 variables.

Z randomization

A whether taking antidepressant medications

M whether taking mental health therapy

Y outcome of interest

age age

gender gender

race race

edu education

marry married or not

sc10 baseline SCL score

work0 work or not

inc income/10000

Details

See Ding et al. (2011) for more details of IMPACT program.

References

Ding, P., Geng, Z. and Zhou, X. H. (2011). Identifying Causal Effect for Multi-Component Intervention Using Instrumental Variable Method: with A Case Study of IMPACT Data. Technical Report.

Examples

```
data(impact)
```

Index

*Topic **causal effect**

heter_IV2, [3](#)

homo_IV1, [5](#)

ImpactIV-package, [2](#)

*Topic **datasets**

impact, [6](#)

*Topic **instrumental variable**

heter_IV2, [3](#)

homo_IV1, [5](#)

ImpactIV-package, [2](#)

heter_IV2, [3](#)

homo_IV1, [5](#)

impact, [6](#)

ImpactIV (ImpactIV-package), [2](#)

ImpactIV-package, [2](#)