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Title Analysis of Matched Case Control Data with a Mismeasured Exposure that is Accompanied by Instrumental Variables

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Description Applying the methodology from Manuel et al. to estimate parameters using a matched case control data with a mismeasured exposure variable that is accompanied by instrumental variables (Submitted).

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mccmeiv-package	<i>Analysis of Matched Case Control Data with a Mismeasured Exposure that is Accompanied by Instrumental Variables</i>
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Description

This package implements the methodology found in Manuel et al. to estimate parameters of a logistic model using data from a matched case control design with a mismeasured exposure variable with the help of instrumental variables.

Author(s)

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References

Manuel, CM, Sinha, S and Wang, S. Matched case-control data with a misclassified exposure: What can be done with instrumental variables? (Submitted).

matcdata	<i>A sample dataset representing matched case control data.</i>
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Description

An example data set with 1 binary response (y), 1 stratification variable (sv), 1 mismeasured binary exposure (w), 1 prognostic factor (z), and 1 instrumental variable (xs) for use with the `meivm3` or `meivm4` functions.

Usage

```
data(matcdata)
```

See Also

[meivm3](#) [meivm4](#) [matcdatamult](#)

Examples

```
data(matcdata)  
out1=with(matcdata,meivm4(y=y,sv=sv,xs=xs,w=w,z=z,alpha=0.1))
```

matcdatamult	<i>A sample dataset representing matched case control data. Similar to matcdata except with multiple stratification/instrumental/miscellaneous variables.</i>
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Description

An example data set with 1 binary response (y), 2 stratification variable (sv1 and sv2), 1 mismeasured binary exposure (w), 2 prognostic factors (z1 and z2), and 2 instruments (xs1 and xs2) for use with the meivm3 or meivm4 functions.

Usage

```
data(matcdatamult)
```

See Also

[meivm3](#) [meivm4](#) [matcdata](#)

Examples

```
data(matcdatamult)
out=with(matcdatamult,meivm3(y=y, sv=sv1,xs=xs1, w=w,z=cbind(z1,z2)))
```

meivm3	<i>Two-step methodology for estimating parameters for a matched case control design with a mismeasured exposure using instrumental variables</i>
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Description

Applies the two-step methodology from Manuel et al. to estimate parameters of a logistic model using matched case control data with a mismeasured exposure variable that is accompanied by a set of instrumental variables.

Usage

```
meivm3(y, sv, xs, w, z, sv.factor = NULL, xs.factor = NULL, z.factor = NULL,
alpha = 0.05, scale = TRUE, setalpha0.to.0 = FALSE, setalpha1.to.0 = FALSE)
```

Arguments

<code>y</code>	A vector of the response variable, representing case ($y=1$) or control ($y=0$) for each observation. For a 1:M matched case-control dataset with n matched sets, y will be a vector of length $N=nx(1+M)$, and it is $y=\text{rep}(c(1, \text{rep}(0, M)), n)$.
<code>sv</code>	A data frame or matrix of confounding variables used for matching each case with the control(s). This data frame should have N number of rows.
<code>xs</code>	A data frame or matrix of instrumental variables used as proxies for the mis-measured variable. This data frame should have N number of rows.
<code>w</code>	A vector of the mis-measured binary exposure variables for each observation. The length of this vector must be N .
<code>z</code>	A data frame or matrix of prognostic factors used to study the association between the response and the mis-measured exposure. This data frame should have N number of rows.
<code>sv.factor</code>	Specify whether any stratification variables are categorical by using the name of the column(s). For example if there is a factor variable in the user specified data frame <code>sv</code> , which is labeled as "Political Affiliation", then <code>sv.factor="Political Affiliation"</code> . Alternatively if there are two factor variables in <code>sv</code> labeled as "Political Affiliation" and "SES", then <code>sv.factor=c("Political Affiliation", "SES")</code> . Any stratification variable that is numeric binary does not need to be declared as a factor.
<code>xs.factor</code>	Specify whether any instrumental variables are categorical by using the name of the column(s). For example if there is an instrument in the user specified <code>xs</code> , which is labeled as "Race of Mother", then <code>xs.factor="Race of Mother"</code> . Alternatively if there are two factor variables in <code>xs</code> variable labeled as "Race of Mother" and "Race of Mother", then <code>xs.factor=c("Race of Mother", "Race of Father")</code> . Any instrumental variable that is numeric binary does not need to be declared as a factor.
<code>z.factor</code>	Specify whether any prognostic variables are categorical by using the name of the column(s). For example, if there is a factor variable in the user specified <code>z</code> labeled "Season", then <code>z.factor="Season"</code> . Alternatively if there are two factor variables in <code>z</code> labeled "Season" and a factor variable labeled Ethnicity, then <code>z.factor=c("Season", "Ethnicity")</code> . Any variable that is numeric binary does not need to be declared as a factor.
<code>alpha</code>	Specify the level of significance for calculating the $(1-\text{alpha})100\%$ Wald confidence intervals of the odds ratio parameter. For example, the default <code>alpha=0.05</code> generates a 95% confidence interval of the odds ratio.
<code>scale</code>	By default, all non factors/numeric variables (those that are not specified as either response, mis-measured covariate, or factors specified in <code>xs.factor</code> / <code>sv.factor</code> / <code>z.factor</code>) are automatically scaled and centered unless the user sets <code>scale=FALSE</code> . Moreover, no numeric binary variables are scaled.
<code>setalpha0.to.0</code>	Sets the misclassification probability $\Pr(W=1 X=0, Y=0) = \Pr(W=1 X=0, Y=1)$ to 0 so that it is not estimated. By default this set to FALSE. Note that this option and <code>setalpha1.to.0</code> cannot be both set to TRUE simultaneously.
<code>setalpha1.to.0</code>	Sets the misclassification probability $\Pr(W=0 X=1, Y=0) = \Pr(W=0 X=1, Y=1)$ to 0 so that it is not estimated. By default this is set to FALSE. Note that this option and <code>setalpha0.to.0</code> cannot be both set to TRUE simultaneously.

Details

Estimation of the parameters is done in two steps. In the first step the set of parameters γ used to model the probability of true exposure status X , $\Pr(X=1|SV, XS, Y=0)$, and the parameters η_0 and η_1 used to model the misclassification probabilities $\alpha_0 = \Pr(W=1|X=0, Y=0) = \Pr(W=1|X=0, Y=1)$ and $\alpha_1 = \Pr(W=0|X=1, Y=0) = \Pr(W=0|X=1, Y=1)$ are estimated. This information is then used in the second step to calculate $\Pr(Y=1|SV, XS, Z)$. The solution is found via the optim function, using the "L-BFGS-B" method. For the first step, the starting values come from a logistic regression of W on the instruments, confounders, and prognostic factors. The starting values for η_0 and η_1 are set to 0. Finally, the starting values for the second step use the naive beta estimates.

Value

`two.step.results`

Provides estimates for the beta parameters of the logistic model for the response y using the two step instrumental variable analysis. Standard errors, p-values, and the (1-alpha)100% Wald confidence intervals for $\exp(\beta)$ are also included in the output.

`naive.results`

Provides estimates for the beta parameters of the logistic model for the response y using the naive approach. Standard errors, p-values, and the (1-alpha)100% Wald confidence intervals for $\exp(\beta)$ are also included in the output.

Author(s)

Chris M. Manuel, Samiran Sinha, and Suojin Wang

References

Manuel, CM, Sinha, S, and Wang, S. Matched case-control data with a misclassified exposure: What can be done with instrumental variables? (Submitted)

See Also

[meivm4](#) [matcdata](#) [matcdatamult](#)

Examples

```
data(matcdata)
out=with(matcdata,meivm3(y=y,sv=sv,xs=xs,w=w,z=z,alpha=0.05))
#For running data with multiple confounders/instruments/prognostic factors see 'matcdatamult'.
```

meivm4

Efficient procedure for estimating parameters for a matched case control design with a mismeasured exposure using instrumental variables

Description

Apply the efficient procedure from Manuel et al. to estimate parameters of a logistic model for matched case control data with a mismeasured exposure that is accompanied by a set of instrumental variables.

Usage

```
meivm4(y, sv, xs, w, z, sv.factor = NULL, xs.factor = NULL, z.factor = NULL,
alpha = 0.05, scale = TRUE, setalpha0.to.0 = FALSE, setalpha1.to.0 = FALSE)
```

Arguments

- | | |
|-----------|--|
| y | A vector of the response variable, representing case (y=1) or control (y=0) for each observation. For a 1:M matched case-control dataset with n matched sets, y will be a vector of length $N=n \times (1+M)$, and it is $y = \text{rep}(c(1, \text{rep}(0, M)), n)$. |
| sv | A data frame or matrix of confounding variables used for matching each case with the control(s). This data frame should have N number of rows. |
| xs | A data frame or matrix of instrumental variables used as proxies for the mismeasured variable. This data frame should have N number of rows. |
| w | A vector of the mismeasured binary exposure variables for each observation. The length of this vector must be N. |
| z | A data frame or matrix of prognostic factors used to study the association between the response and the mismeasured exposure. This data frame should have N number of rows. |
| sv.factor | Specify whether any stratification variables are categorical by using the name of the column(s). For example if there is a factor variable in the user specified data frame sv, which is labeled as "Political Affiliation", then <code>sv.factor="Political Affiliation"</code> . Alternatively if there are two factor variables in sv labeled as "Political Affiliation" and "SES", then <code>sv.factor=c("Political Affiliation","SES")</code> . Any stratification variable that is numeric binary does not need to be declared as a factor. |
| xs.factor | Specify whether any instrumental variables are categorical by using the name of the column(s). For example if there is an instrument in the user specified xs, which is labeled as "Race of Mother", then <code>xs.factor="Race of Mother"</code> . Alternatively if there are two factor variables in xs variable labeled as "Race of Mother" and "Race of Mother", then <code>xs.factor=c("Race of Mother","Race of Father")</code> . Any instrumental variable that is numeric binary does not need to be declared as a factor. |

<code>z.factor</code>	Specify whether any prognostic variables are categorical by using the name of the column(s). For example, if there is a factor variable in the user specified <code>z</code> labeled "Season", then <code>z.factor="Season"</code> . Alternatively if there are two factor variables in <code>z</code> labeled "Season" and a factor variable labeled Ethnicity , then <code>z.factor=c("Season","Ethnicity")</code> . Any variable that is numeric binary does not need to be declared as a factor.
<code>alpha</code>	Specify the level of significance for calculating the (1-alpha)100% Wald confidence intervals of the odds ratio parameter. For example, the default <code>alpha=0.05</code> generates a 95% confidence interval of the odds ratio.
<code>scale</code>	By default, all non factors/numeric variables (those that are not specified as either response, mismeasured covariate, or factors specified in <code>xs.factor / sv.factor / z.factor</code>) are automatically scaled and centered unless the user sets <code>scale=FALSE</code> . Moreover, no numeric binary variables are scaled.
<code>setalpha0.to.0</code>	Sets the misclassification probability $\Pr(W=1 X=0,Y=0) = \Pr(W=1 X=0,Y=1)$ to 0 so that it is not estimated. By default this set to FALSE. Note that this option and <code>setalpha1.to.0</code> cannot be both set to TRUE simultaneously.
<code>setalpha1.to.0</code>	Sets the misclassification probability $\Pr(W=0 X=1,Y=0) = \Pr(W=0 X=1,Y=1)$ to 0 so that it is not estimated. By default this is set to FALSE. Note that this option and <code>setalpha0.to.0</code> cannot be both set to TRUE simultaneously.

Details

In comparison to the methodology used in the function `meivm3`, the efficient estimation approach estimates the parameters in one step. These parameters include γ used to model the probability of true exposure status X , $\Pr(X=1|SV,XS,Y=0)$, the parameters η_0 and η_1 used to model the misclassification probabilities $\alpha_0 = \Pr(W=1|X=0,Y=0) = \Pr(W=1|X=0,Y=1)$ and $\alpha_1 = \Pr(W=0|X=1,Y=0) = \Pr(W=0|X=1,Y=1)$, and the beta parameters used in modeling $\Pr(Y=1|SV,XS,Z)$. The solution is found via the `optim` function, using the "L-BFGS-B" method. The starting values for γ come from a logistic regression of W on the instruments, confounders, and prognostic factors. The starting values for η_0 and η_1 are set to 0, and the starting values for beta are the naive estimates.

Value

<code>efficient.results</code>	Provides estimates for the beta parameters of the logistic model for the response y using the efficient estimator approach. Standard errors, p-values, and the (1-alpha)100% Wald confidence interval for $\exp(\beta)$ are also included in the output.
<code>naive.results</code>	Provides estimates for the beta parameters of the logistic model for the response y using the naive approach. Standard errors, p-values, and the (1-alpha)100% Wald confidence intervals for $\exp(\beta)$ are also included in the output.

Author(s)

Chris M. Manuel, Samiran Sinha, and Suojin Wang

References

Manuel, CM, Sinha, S, and Wang, S. Matched case-control data with a misclassified exposure: What can be done with instrumental variables? (Submitted)

See Also

[meivm3](#) [matcdata](#) [matcdatamult](#)

Examples

```
data(matcdata)
out2=with(matcdata,meivm4(y=y,sv=sv,xs=xs,w=w,z=z,alpha=0.01))
#For running data with multiple confounders/instruments/prognostic factors see 'matcdatamult'.
```


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