

# Example R and Stata code for a Mendelian randomization analysis

Sanderson et al.

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## R code

- Read in the data

```
library(tidyverse)
library(haven)
library(remotes)

dat <- read_dta("https://raw.githubusercontent.com/remlapmot/mrrobust/master/dodata.dta")
dat <- dat %>% filter(ldlcp2 < 1e-8)
```

## Example code using the MendelianRandomization package

- Install the package and load it into memory

```
library(MendelianRandomization)
```

- Convert our data frame to the required class

```
datfmt <- mr_mvinput(
  bx = as.matrix(cbind(dat$ldlcbeta, dat$hdlcbeta, dat$tgbeta)),
  bxse = as.matrix(cbind(dat$ldlcse, dat$hdlcse, dat$tgse)),
  by = dat$chdbeta,
  byse = dat$chdse,
  exposure = "exposure",
  outcome = "outcome",
  snps = "snps",
  effect_allele = dat$a1,
  other_allele = dat$a2,
  eaf = NA
)
```

- Fit an MVMR/MVIVW model

```
mvivwfit <- mr_mvivw(datfmt)
mvivwfit
```

```
##
## Multivariable inverse-variance weighted method
## (variants uncorrelated, random-effect model)
```

```
##
## Number of Variants : 73
##
## -----
##      Exposure Estimate Std Error  95% CI      p-value
## exposure_1      0.429      0.061  0.309, 0.548  0.000
## exposure_2     -0.194      0.131 -0.451, 0.062  0.138
## exposure_3      0.226      0.123 -0.016, 0.468  0.067
## -----
## Residual standard error = 1.490
## Heterogeneity test statistic = 155.3766 on 70 degrees of freedom, (p-value = 0.0000)
```

- Fit an MVMR-Egger model

```
mvmreggerfit <- mr_mvegger(datfmt)
mvmreggerfit
```

```
##
## Multivariable MR-Egger method
## (variants uncorrelated, random-effect model)
##
## Orientated to exposure : 1
## Number of Variants : 73
## -----
##      Exposure Estimate Std Error  95% CI      p-value
## exposure_1      0.567      0.100  0.371, 0.764  0.000
## exposure_2     -0.136      0.133 -0.398, 0.125  0.306
## exposure_3      0.274      0.125  0.030, 0.518  0.028
## (intercept)    -0.009      0.005 -0.020, 0.001  0.084
## -----
## Residual standard error = 1.469
## Heterogeneity test statistic = 148.9290 on 69 degrees of freedom, (p-value = 0.0000)
```

## Example code using the MVMR and RMVMR packages

- Install the package and load it into memory

```
if (!requireNamespace("MVMR", quietly = TRUE)) {
  remotes::install_github("WSpiller/MVMR")
}
library(MVMR)
```

- Create a data object of the required structure

```
r_input <- format_mvrm(
  BXGs = dat[,c("ldlcbeta", "hdlcbeta", "tgbeta")],
  BYG = dat$chdbeta,
  seBXGs = dat[,c("ldlcse", "hdlcse", "tgse")],
  seBYG = dat$chdse,
  RSID = dat$rsid
)
```

- Fit an MVMR model

```
mvmrfit <- ivw_mvvr(r_input)
```

```
##
## Multivariable MR
##
##           Estimate Std. Error   t value    Pr(>|t|)
## exposure1  0.4286200  0.0609661   7.030464 1.099077e-09
## exposure2 -0.1941989  0.1308289  -1.484372 1.421994e-01
## exposure3  0.2260456  0.1232828   1.833554 7.097168e-02
##
## Residual standard error: 1.49 on 70 degrees of freedom
```

- Conditional F-statistics for instrument strength (Sanderson, Spiller, and Bowden 2021)

```
strength_mvvr(r_input)
```

```
## Warning in strength_mvvr(r_input): Covariance between effect of genetic variants on each exposure
## not specified. Fixing covariance at 0.
```

```
##
## Conditional F-statistics for instrument strength
##
##           exposure1 exposure2 exposure3
## F-statistic  126.7447  35.29937  39.32731
```

```
##           exposure1 exposure2 exposure3
## F-statistic  126.7447  35.29937  39.32731
```

- Fit a radial MVMR model

```
if (!requireNamespace("RMVMR", quietly = TRUE)) {
  remotes::install_github("WSpiller/RMVMR")
}
library(RMVMR)

rmvmr_input <- mrmvininput_to_rmvmr_format(datfmt)

rmvmr_fit <- ivw_rmvmr(rmvmr_input, summary = TRUE)
```

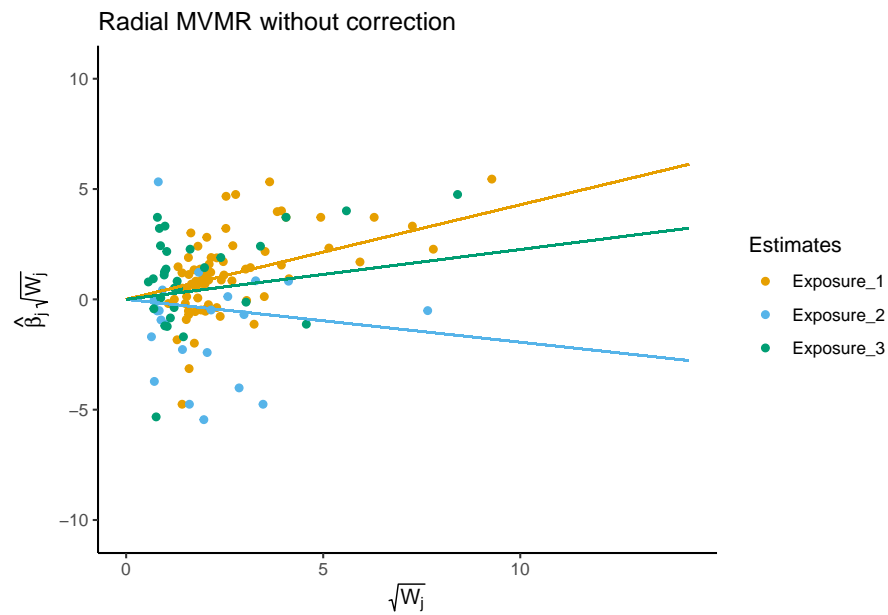
```
##
## Radial Multivariable MR
##
##           Estimate Std. Error   t value    Pr(>|t|)
## exposure1  0.4286200  0.0609661   7.030464 1.099077e-09
## exposure2 -0.1941989  0.1308289  -1.484372 1.421994e-01
## exposure3  0.2260456  0.1232828   1.833554 7.097168e-02
##
## Residual standard error: 1.49 on 70 degrees of freedom
```

```
rmvmr_fit$coef
```

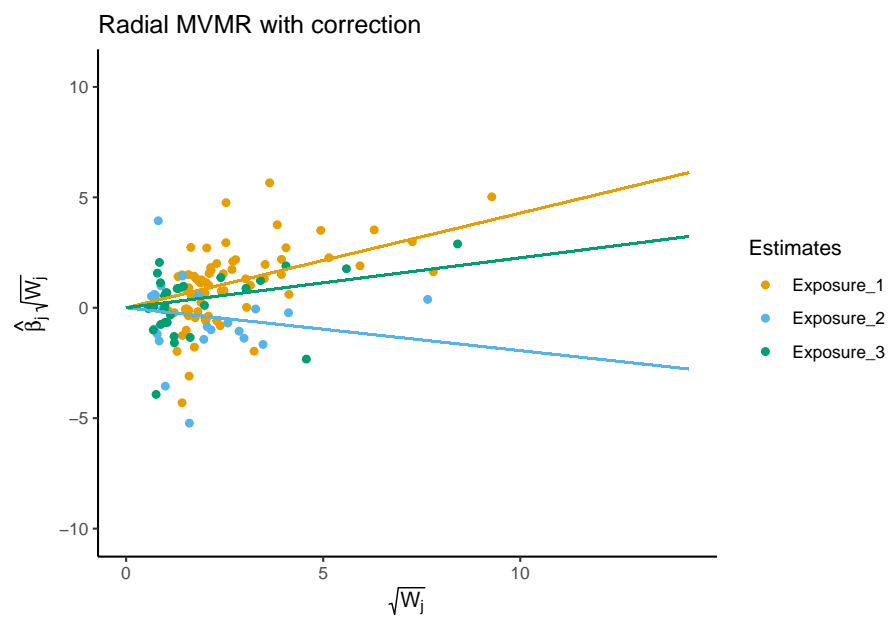
```
##           Estimate Std. Error   t value    Pr(>|t|)
## exposure1  0.4286200  0.0609661   7.030464 1.099077e-09
## exposure2 -0.1941989  0.1308289  -1.484372 1.421994e-01
## exposure3  0.2260456  0.1232828   1.833554 7.097168e-02
```

- Plot the radial MVMR models

```
plt_rmvmr <- plot_rmvmr(rmvmr_input, rmvmr = rmvmr_fit)
plt_rmvmr$p1
```



```
plt_rmvmr$p2
```



- Heterogeneity statistics

```
pleio_rmvmr <- pleiotropy_rmvmr(rmvmr_input, rmvmr = rmvmr_fit)
pleio_rmvmr$gq
```

##	q_statistic	p_value
## Exposure_1	76.37356	2.812609e-01
## Exposure_2	59.58894	8.243650e-06
## Exposure_3	45.88633	1.308596e-02

```
head(pleio_rmvmr$qdat)
```

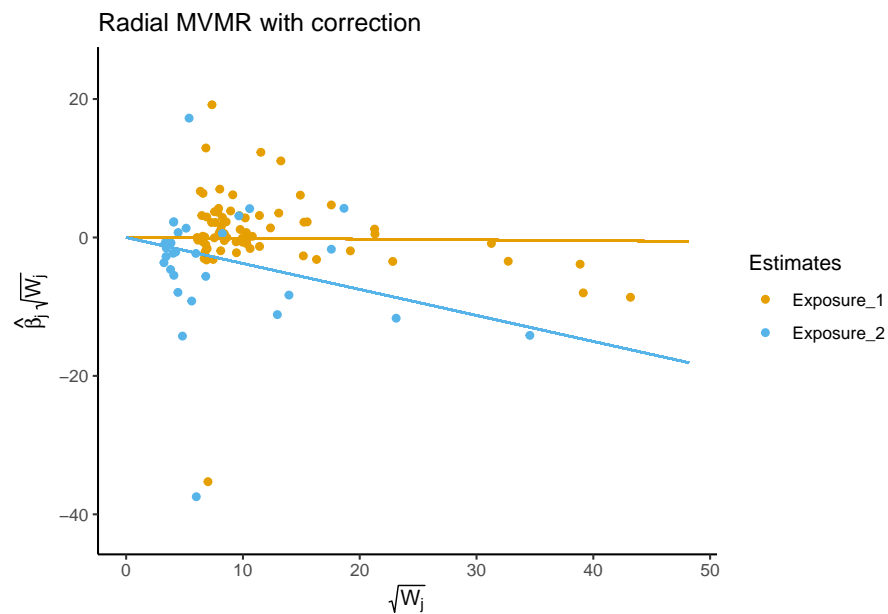
```
##      snp      wj corrected_beta      qj      qj_p ref_exposure
## 1 snp_1 2.414215      0.3141338 0.031643343 0.85881269 Exposure_1
## 2 snp_2 3.938760      0.3821443 0.008507667 0.92650973 Exposure_1
## 3 snp_3 3.246657     -0.6057993 3.473998077 0.06234046 Exposure_1
## 4 snp_4 2.001191      0.3273484 0.020524059 0.88608310 Exposure_1
## 5 snp_5 9.278826      0.5414293 0.118081748 0.73112437 Exposure_1
## 6 snp_6 1.317671      1.0713541 0.544339092 0.46064001 Exposure_1
```

- Conditional F-statistics for instrument strength (Sanderson, Spiller, and Bowden 2021)

```
str_rmvmr <- strength_rmvmr(rmvmr_input)
```

```
## Warning in MVMR::strength_mvmm(r_input, gencov): Covariance between effect of genetic variants on
## each exposure not specified. Fixing covariance at 0.
```

```
str_rmvmr$plot[[2]]
```



```
str_rmvmr$qstat[[2]]
```

```
##      q_statistic      p_value
## Exposure_1      335.0602 6.799449e-36
## Exposure_2      371.0830 1.403374e-61
```

## Stata code

- Load the Statamarkdown package to enable Stata code chunks in an R Markdown file

```
library(Statamarkdown)
```

- Read in the data and create an indicator variable to select observations with  $p$ -value between the genotype and LDL-C  $< 10^{-8}$

```
use https://raw.githubusercontent.com/remlapmot/mrrobust/master/dodata, clear
gen byte sel1 = (ldlcp2 < 1e-8)
```

## Example code using the mrrobust package

- Install the mrrobust package using the github package

```
// Note: output suppressed
net install mrrobust, from("https://raw.githubusercontent.com/remlapmot/mrrobust/master/")
mrdeps
```

- Fit IVW (with fixed effect standard errors)

```
mregger chdbeta ldlcbeta [aw=1/(chdse^2)] if sel1==1, ivw fe
```

```

                                     Number of genotypes = 73
                               Residual standard error constrained at 1
-----+-----
          | Coefficient  Std. err.      z    P>|z|      [95% conf. interval]
-----+-----
chdbeta   |
  ldlcbeta |   .4815055   .038221    12.60   0.000    .4065938   .5564173
-----+-----
```

- Fit MR-Egger reporting I2GX statistic and heterogeneity Q-test

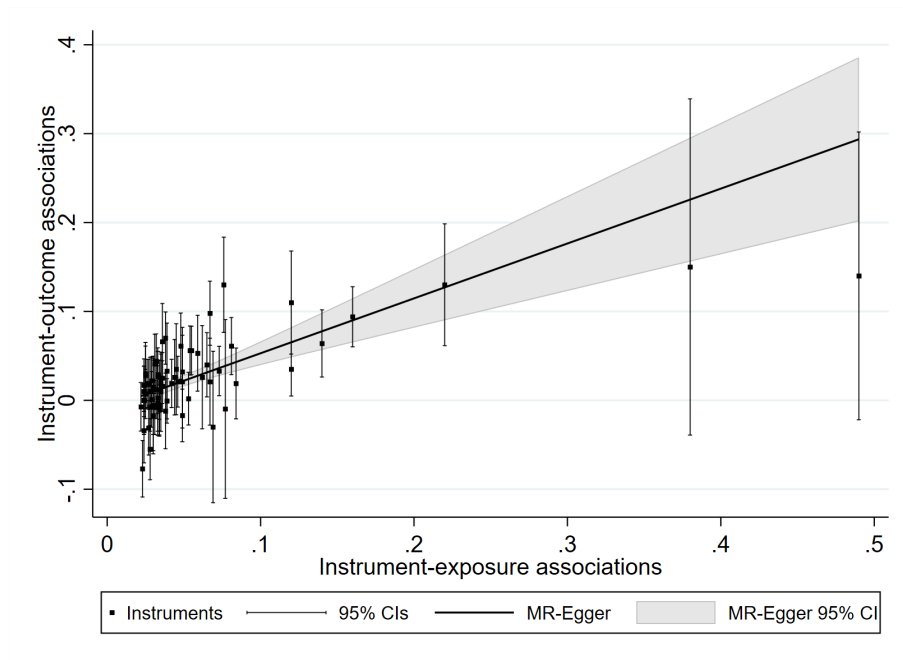
```
mregger chdbeta ldlcbeta [aw=1/(chdse^2)] if sel1==1, gxse(ldlcse) heterogi
```

```

                                     Q_GX statistic (weighted) = 3454.26
                               I2_GX statistic (weighted) = 97.92%
                                     Number of genotypes = 73
                               Residual standard error = 1.548
Ruecker's Q for heterogeneity; chi2(71) = 170.11 (p = 0.0000)
                               I-squared statistic = 58.3% (95% CI 45.8%, 67.8%)
-----+-----
          | Coefficient  Std. err.      z    P>|z|      [95% conf. interval]
-----+-----
chdbeta   |
  slope   |   .6173131   .1034573    5.97   0.000    .4145405   .8200858
  _cons   |  -.0087706   .0054812   -1.60   0.110   -.0195136   .0019723
-----+-----
```

- Simple plot of IVW and MR-Egger univariate estimates

```
mreggerplot chdbeta chdse ldlcbeta ldlcse if sel1==1
qui gr export mreggerplot-1.png, replace
```



- Adding the modal and median estimates onto the plot

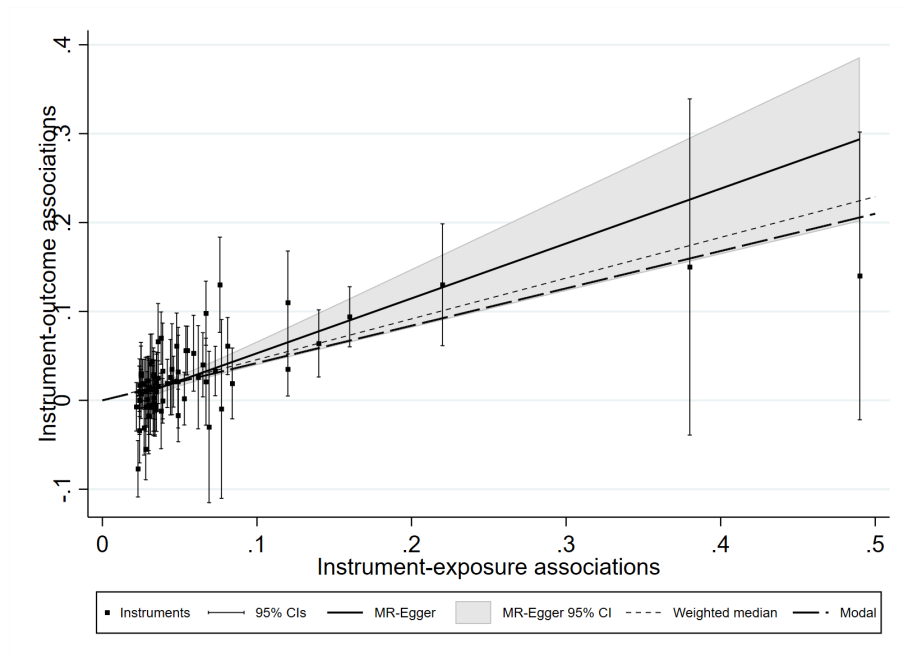
```
mreggerplot chdbeta chdse ldlcbeta ldlcse if sel1==1
mrmedian chdbeta chdse ldlcbeta ldlcse if sel1==1, weighted
addplot : function _b[beta]*x if sel1==1, ///
    range(0 0.5) lc(gs0) lp(shortdash) lw(vthin)
mrmodal chdbeta chdse ldlcbeta ldlcse if sel1==1, phi(.25)
addplot : function _b[beta]*x if sel1==1, ///
    range(0 0.5) lc(gs0) lp(longdash) ///
    legend(order(5 "Instruments" ///
        4 "95% CIs" 3 "MR-Egger" 2 "MR-Egger 95% CI" ///
        6 "Weighted median" 7 "Modal") ///
    rows(1) si(vsmall) symx(*.5))
qui gr export mreggerplot-2.png, replace
```

Number of genotypes = 73  
Replications = 1000

	Coefficient	Std. err.	z	P> z	[95% conf. interval]	
beta	.4582573	.0633137	7.24	0.000	.3341648	.5823499

Number of genotypes = 73  
Replications = 1000  
Phi = .25

	Coefficient	Std. err.	z	P> z	[95% conf. interval]	
beta	.4198713	.2260632	1.86	0.063	-.0232045	.8629471



- Fit and MVMR model with phenotypes LDL-c and HDL-c (Burgess, Dudbridge, and Thompson 2015).

```
mvmr chdbeta ldlcbeta hdlcbeta [aw=1/(chdse^2)] if sel1==1
```

Number of genotypes = 73  
 Number of phenotypes = 2  
 Standard errors: Random effect  
 Residual standard error = 1.514

	Coefficient	Std. err.	z	P> z	[95% conf. interval]	
chdbeta						
ldlcbeta	.4670719	.0581901	8.03	0.000	.3530214	.5811224
hdlcbeta	-.2930048	.1211822	-2.42	0.016	-.5305175	-.0554921

- Additionally include a third phenotype – triglycerides.

```
mvmr chdbeta ldlcbeta hdlcbeta tgbeta [aw=1/(chdse^2)] if sel1==1
```

Number of genotypes = 73  
 Number of phenotypes = 3  
 Standard errors: Random effect  
 Residual standard error = 1.490

	Coefficient	Std. err.	z	P> z	[95% conf. interval]	
chdbeta						
ldlcbeta	.42862	.0609661	7.03	0.000	.3091286	.5481113
hdlcbeta	-.1941989	.1308289	-1.48	0.138	-.4506189	.0622211
tgbeta	.2260456	.1232828	1.83	0.067	-.0155842	.4676755

- Report the QA statistic for instrument validity and the conditional F-statistics for instrument strength for each phenotype (Sanderson et al. 2019; Sanderson, Spiller, and Bowden 2021).



```
mvmr chdbeta ldlcbeta hdlcbeta tgbeta [aw=1/(chdse^2)] if sel1==1, gxse(ldlcse hdlcse tgse)
```

```
> e hdlcse tgse)
```

```
Number of genotypes = 73
Number of phenotypes = 3
Standard errors: Random effect
Residual standard error = 1.490
```

	Coefficient	Std. err.	z	P> z	[95% conf. interval]	
chdbeta						
ldlcbeta	.42862	.0609661	7.03	0.000	.3091286	.5481113
hdlcbeta	-.1941989	.1308289	-1.48	0.138	-.4506189	.0622211
tgbeta	.2260456	.1232828	1.83	0.067	-.0155842	.4676755

Q\_A statistic for instrument validity; chi2(70) = 152.88 (p = 0.0000)

Conditional F-statistics for instrument strength:

F\_x1 = 130.31 (ldlcbeta)

F\_x2 = 36.29 (hdlcbeta)

F\_x3 = 40.44 (tgbeta)

- Fit an MVMR-Egger regression (Rees, Wood, and Burgess 2017), orienting the model with respect to the first phenotype in the main *varlist*.

```
mrmvegger chdbeta ldlcbeta hdlcbeta tgbeta [aw=1/(chdse^2)] if sel1==1
```

MVMR-Egger model oriented wrt: ldlcbeta

Number of genotypes = 73

Number of phenotypes = 3

Residual standard error = 1.469

	Coefficient	Std. err.	z	P> z	[95% conf. interval]	
chdbeta						
ldlcbeta	.5672993	.1002611	5.66	0.000	.370791	.7638075
hdlcbeta	-.1364113	.1332727	-1.02	0.306	-.3976209	.1247983
tgbeta	.2739803	.1246927	2.20	0.028	.0295871	.5183735
_cons	-.0093655	.0054187	-1.73	0.084	-.019986	.001255

We can also orient the model with respect to HDL-C.

```
mrmvegger chdbeta ldlcbeta hdlcbeta tgbeta [aw=1/(chdse^2)] if sel1==1, orient(2)
```

```
> t(2)
```

MVMR-Egger model oriented wrt: hdlcbeta

Number of genotypes = 73

Number of phenotypes = 3

Residual standard error = 1.501

	Coefficient	Std. err.	z	P> z	[95% conf. interval]	
chdbeta						

ldlcbeta		.4286398	.0614056	6.98	0.000	.308287	.5489926
hdlcbeta		-.1989637	.1541909	-1.29	0.197	-.5011723	.1032449
tgbeta		.2256794	.1243221	1.82	0.069	-.0179875	.4693463
_cons		.0002155	.0036218	0.06	0.953	-.006883	.0073141

Or we can orient the model with respect to triglycerides.

```
mrmvegger chdbeta ldlcbeta hdlcbeta tgbeta [aw=1/(chdse^2)] if sel1==1, orient(3)
```

```
> t(3)
```

MVMR-Egger model oriented wrt: tgbeta							
Number of genotypes = 73							
Number of phenotypes = 3							
Residual standard error = 1.499							
		Coefficient	Std. err.	z	P> z	[95% conf. interval]	
chdbeta							
ldlcbeta		.4203073	.0660026	6.37	0.000	.2909447	.54967
hdlcbeta		-.1903089	.1321536	-1.44	0.150	-.4493252	.0687075
tgbeta		.2065651	.1365427	1.51	0.130	-.0610537	.474184
_cons		.0013499	.003951	0.34	0.733	-.0063939	.0090936

## Session information for reproducibility

### R session information

```
library(sessioninfo)
session_info()
```

```
- Session info -----
setting value
version R version 4.2.3 (2023-03-15 ucrt)
os Windows 10 x64 (build 22621)
system x86_64, mingw32
ui RTerm
language (EN)
collate English_United Kingdom.utf8
ctype English_United Kingdom.utf8
tz Europe/London
date 2023-03-20
pandoc 2.19.2 @ C:/Program Files/RStudio/resources/app/bin/quarto/bin/tools/ (via rmarkdown)

- Packages -----
package * version date (UTC) lib source
arrangements 1.1.9 2020-09-13 [1] CRAN (R 4.2.2)
backports 1.4.1 2021-12-13 [1] CRAN (R 4.2.0)
base64enc 0.1-3 2015-07-28 [1] CRAN (R 4.2.0)
checkmate 2.1.0 2022-04-21 [1] CRAN (R 4.2.1)
cli 3.6.0 2023-01-09 [1] CRAN (R 4.2.2)
cluster 2.1.4 2022-08-22 [2] CRAN (R 4.2.3)
```

codetools	0.2-19	2023-02-01	[2]	CRAN	(R 4.2.3)
colorspace	2.1-0	2023-01-23	[1]	CRAN	(R 4.2.2)
curl	5.0.0	2023-01-12	[1]	CRAN	(R 4.2.2)
data.table	1.14.8	2023-02-17	[1]	CRAN	(R 4.2.2)
DEoptimR	1.0-11	2022-04-03	[1]	CRAN	(R 4.2.0)
digest	0.6.31	2022-12-11	[1]	CRAN	(R 4.2.2)
dplyr	* 1.1.0	2023-01-29	[1]	CRAN	(R 4.2.2)
ellipsis	0.3.2	2021-04-29	[1]	CRAN	(R 4.2.0)
evaluate	0.20	2023-01-17	[1]	CRAN	(R 4.2.2)
fansi	1.0.4	2023-01-22	[1]	CRAN	(R 4.2.2)
farver	2.1.1	2022-07-06	[1]	CRAN	(R 4.2.1)
fastmap	1.1.1	2023-02-24	[1]	CRAN	(R 4.2.2)
forcats	* 1.0.0	2023-01-29	[1]	CRAN	(R 4.2.2)
foreach	1.5.2	2022-02-02	[1]	CRAN	(R 4.2.2)
foreign	0.8-84	2022-12-06	[2]	CRAN	(R 4.2.3)
Formula	1.2-5	2023-02-24	[1]	CRAN	(R 4.2.2)
generics	0.1.3	2022-07-05	[1]	CRAN	(R 4.2.1)
ggplot2	* 3.4.1	2023-02-10	[1]	CRAN	(R 4.2.2)
glmnet	4.1-6	2022-11-27	[1]	CRAN	(R 4.2.2)
glue	1.6.2	2022-02-24	[1]	CRAN	(R 4.2.0)
gmp	0.7-1	2023-02-07	[1]	CRAN	(R 4.2.2)
gridExtra	2.3	2017-09-09	[1]	CRAN	(R 4.2.1)
gtable	0.3.2	2023-03-17	[1]	CRAN	(R 4.2.3)
haven	* 2.5.2	2023-02-28	[1]	CRAN	(R 4.2.2)
Hmisc	* 5.0-1	2023-03-08	[1]	CRAN	(R 4.2.2)
hms	1.1.2	2022-08-19	[1]	CRAN	(R 4.2.1)
htmlTable	2.4.1	2022-07-07	[1]	CRAN	(R 4.2.1)
htmltools	0.5.4	2022-12-07	[1]	CRAN	(R 4.2.2)
htmlwidgets	1.6.2	2023-03-17	[1]	CRAN	(R 4.2.3)
httr	1.4.5	2023-02-24	[1]	CRAN	(R 4.2.2)
iterators	1.0.14	2022-02-05	[1]	CRAN	(R 4.2.2)
iterpc	0.4.2	2020-01-10	[1]	CRAN	(R 4.2.2)
jsonlite	1.8.4	2022-12-06	[1]	CRAN	(R 4.2.2)
knitr	1.42.5	2023-03-14	[1]	Github	(yihui/knitr@c50d307)
labeling	0.4.2	2020-10-20	[1]	CRAN	(R 4.2.0)
lattice	0.20-45	2021-09-22	[2]	CRAN	(R 4.2.3)
lazyeval	0.2.2	2019-03-15	[1]	CRAN	(R 4.2.1)
lifecycle	1.0.3	2022-10-07	[1]	CRAN	(R 4.2.1)
lubridate	* 1.9.2	2023-02-10	[1]	CRAN	(R 4.2.2)
magrittr	2.0.3	2022-03-30	[1]	CRAN	(R 4.2.0)
MASS	7.3-58.3	2023-03-07	[1]	CRAN	(R 4.2.2)
Matrix	1.5-3	2022-11-11	[1]	CRAN	(R 4.2.2)
MatrixModels	0.5-1	2022-09-11	[1]	CRAN	(R 4.2.1)
MendelianRandomization	* 0.7.0	2023-01-09	[1]	CRAN	(R 4.2.2)
munsell	0.5.0	2018-06-12	[1]	CRAN	(R 4.2.0)
MVMR	* 0.3	2023-03-20	[1]	Github	(WSpiller/MVMR@3a0b54b)
nnet	7.3-18	2022-09-28	[2]	CRAN	(R 4.2.3)
pillar	1.8.1	2022-08-19	[1]	CRAN	(R 4.2.1)
pkgconfig	2.0.3	2019-09-22	[1]	CRAN	(R 4.2.0)
plotly	4.10.1	2022-11-07	[1]	CRAN	(R 4.2.2)
purrr	* 1.0.1	2023-01-10	[1]	CRAN	(R 4.2.2)
quantreg	5.94	2022-07-20	[1]	CRAN	(R 4.2.1)
R6	2.5.1	2021-08-19	[1]	CRAN	(R 4.2.0)
RadialMR	1.0	2023-02-17	[1]	Github	(WSpiller/RadialMR@d63d3fc)
Rcpp	1.0.10	2023-01-22	[1]	CRAN	(R 4.2.2)
readr	* 2.1.4	2023-02-10	[1]	CRAN	(R 4.2.2)
remotes	* 2.4.2	2021-11-30	[1]	CRAN	(R 4.2.0)
rjson	0.2.21	2022-01-09	[1]	CRAN	(R 4.2.0)

rlang	1.1.0	2023-03-14	[1]	CRAN (R 4.2.2)
rmarkdown	2.20	2023-01-19	[1]	CRAN (R 4.2.2)
RMVMR	* 0.2	2023-03-20	[1]	Github (WSpiller/RMVMR@a162f9e)
robustbase	0.95-0	2022-04-02	[1]	CRAN (R 4.2.2)
rpart	4.1.19	2022-10-21	[2]	CRAN (R 4.2.3)
rstudioapi	0.14	2022-08-22	[1]	CRAN (R 4.2.1)
scales	1.2.1	2022-08-20	[1]	CRAN (R 4.2.1)
sessioninfo	* 1.2.2	2021-12-06	[1]	CRAN (R 4.2.1)
shape	1.4.6	2021-05-19	[1]	CRAN (R 4.2.0)
SparseM	1.81	2021-02-18	[1]	CRAN (R 4.2.0)
Statamarkdown	* 0.7.2	2023-02-15	[1]	CRAN (R 4.2.2)
stringi	1.7.12	2023-01-11	[1]	CRAN (R 4.2.2)
stringr	* 1.5.0	2022-12-02	[1]	CRAN (R 4.2.2)
survival	3.5-5	2023-03-12	[2]	CRAN (R 4.2.2)
tibble	* 3.2.1	2023-03-20	[1]	CRAN (R 4.2.3)
tidyr	* 1.3.0	2023-01-24	[1]	CRAN (R 4.2.2)
tidyselect	1.2.0	2022-10-10	[1]	CRAN (R 4.2.1)
tidyverse	* 2.0.0	2023-02-22	[1]	CRAN (R 4.2.2)
timechange	0.2.0	2023-01-11	[1]	CRAN (R 4.2.2)
tzdb	0.3.0	2022-03-28	[1]	CRAN (R 4.2.0)
utf8	1.2.3	2023-01-31	[1]	CRAN (R 4.2.2)
vctrs	0.6.0	2023-03-16	[1]	CRAN (R 4.2.2)
viridisLite	0.4.1	2022-08-22	[1]	CRAN (R 4.2.1)
withr	2.5.0	2022-03-03	[1]	CRAN (R 4.2.0)
xfun	0.37	2023-01-31	[1]	CRAN (R 4.2.2)
yaml	2.3.7	2023-01-23	[1]	CRAN (R 4.2.2)

[1] C:/Users/tom/AppData/Local/R/win-library/4.2

[2] C:/Program Files/R/R-4.2.3/library

## Stata session information

```
about
ado describe mrrobust
```

Stata/MP 17.0 for Windows (64-bit x86-64)

Revision 08 Mar 2023

Copyright 1985-2021 StataCorp LLC

Total physical memory: 32.00 GB

Available physical memory: 15.27 GB

Stata license: Unlimited-user 2-core network, expiring 21 Jan 2024

Serial number: 501709378202

Licensed to: Tom Palmer

University of Bristol

```
[47] package mrrobust from https://raw.githubusercontent.com/remlapmot/mrrobust
> /master
```

TITLE

'mrrobust': Stata package for two-sample Mendelian randomization analyses

DESCRIPTION/AUTHOR(S)

Author: Tom Palmer

Distribution-Date: 20210917

INSTALLATION FILES

m\mrmedian.ado  
m\mrmedian.sthlp  
m\mrmedianobs.ado  
m\mrmedianobs\_work.ado  
m\mrmedianobs.sthlp  
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m\mrleaveoneout.sthlp  
m\mrrobust-author.ihlp

INSTALLED ON

17 Feb 2023

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