

Practical MR session using R

Daniele Bottigliengo

Institute for Biomedicine, Eurac Research, Bolzano, Italy

July 22th, 2021



Outline

1. Introduction

- *Two-sample MR framework*
- *R packages for Two-sample MR*

2. Data preparation

- *Summary GWAS data*
- *Data harmonization*

3. Performing MR analysis

- `MendelianRandomization` and `TwoSampleMR` functions
- *Displaying results*

4. Sensitivity analyses

- *Evaluating MR assumptions*
- *Using MR robust methods*

5. Case study

Introduction

- Two sample MR analysis
- Summary instrument-exposure and instrument-outcome association results
- Summary-level data from GWAS
- Non-overlapping sets of individuals

R packages

TwoSampleMR

- GWAS database
- Data harmonization
- Most commonly used MR methods
- Visualization and sensitivity analyses
- Installation from Github:

```
install.packages("remotes")  
remotes::install_github("MRCIEU/TwoSampleMR")
```

MendelianRandomization

- GWAS database
- Several MR methods are implemented
- Visualization and sensitivity analyses
- Installation from CRAN:

```
install.packages("MendelianRandomization", dependencies
```

Data sources

- Retrieving summary-level GWAS data:
 - NHGRI-EBI GWAS Catalog
 - dbGaP CHARGE Consortium GWAS results (*study accession phs000930*)
 - Neale lab
 - GeneATLAS
 - FinnGen cohort
 - Phenoscanner → MendelianRandomization
 - IEU GWAS database → TwoSampleMR
 - GWAS consortium websites, e.g. **CKDGen**, **GIANT**, **MAGIC**, **ENIGMA**, **ThyroidOmics**, **COVID-19 HGI**

IVs selection

- Several strategies for selecting genetic instruments
- Accounting for:
 - weak instrument bias
 - linkage disequilibrium (LD), i.e. correlations between instruments
 - horizontal pleiotropy

IVs selection

1. Selection of genome-wide significant instruments from GWAS exposure data ($p\text{-value} < 5 \times 10^{-8}$)
 - One can select functional genetic variants, i.e. genome-wide significant SNPs in the encoding gene
 - Check $F\text{-statistic}$ to evaluate the strength of the instrument
2. Perform LD clumping to select independent genetic instruments
 - Choice of r^2 threshold might depend on the situation
 - $r^2 < 0.001$ is very conservative but might not work well for rare variants in small samples
 - $r^2 < 0.01$ is less conservative but acceptable for many situations
 - Some studies used more liberal thresholds, e.g. < 0.2
3. Proxies can be used to replace exposure variants that are missing in the outcome dataset
 - The proxy should be in high LD with the missing genetic instrument
4. Applying Steiger filter to remove IVs for which there might evidence of reverse causation

Data harmonization

Harmonise exposure and outcome effects

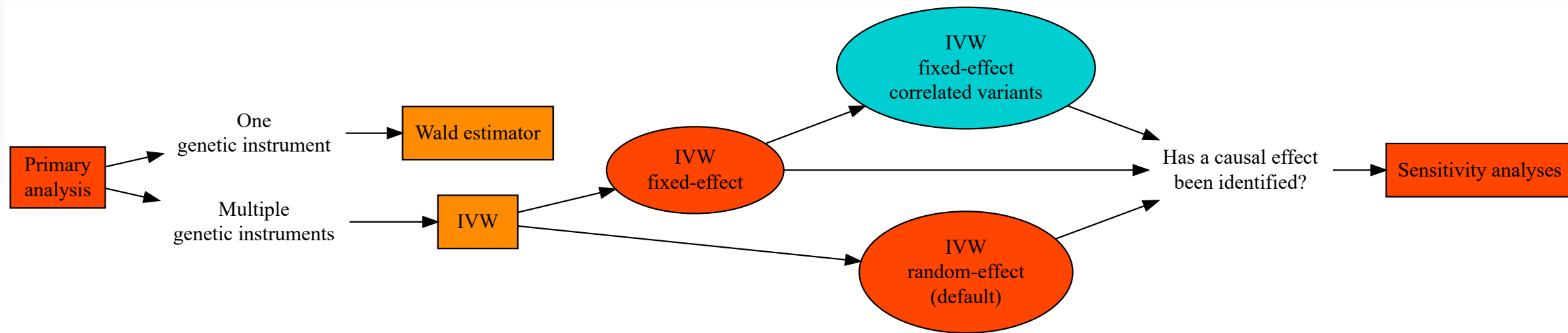
SNP	Exposure GWAS				Outcome GWAS			
	Effect	Effect allele	Other allele	Effect allele frequency	Effect	Effect allele	Other allele	Effect allele frequency
rs12345	0.132	A	G	0.28	0.022	A	G	0.26
rs23456	-0.485	G	T	0.41	0.056	T	G	0.61
rs34567	0.203	G	C	0.11	-0.046	G	C	0.88



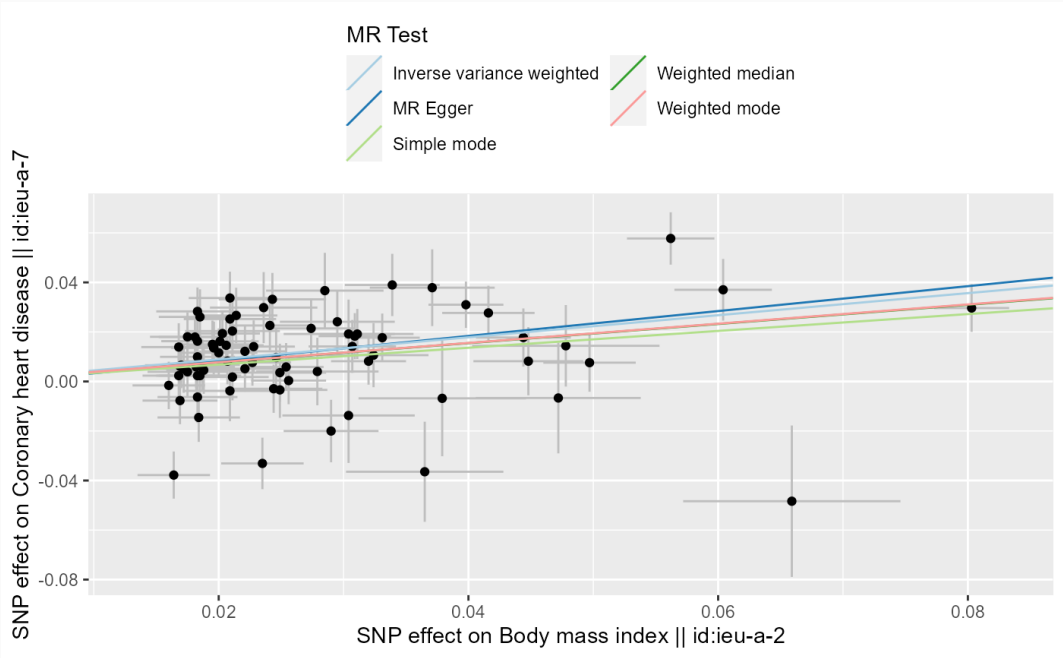
SNP	Exposure GWAS				Outcome GWAS			
	Effect	Effect allele	Other allele	Effect allele frequency	Effect	Effect allele	Other allele	Effect allele frequency
rs12345	0.132	A	G	0.28	0.022	A	G	0.26
rs23456	-0.485	G	T	0.41	-0.056	G	T	0.39
rs34567	0.203	G	C	0.11	0.046	G	C	0.12

<https://elifesciences.org/articles/34408>

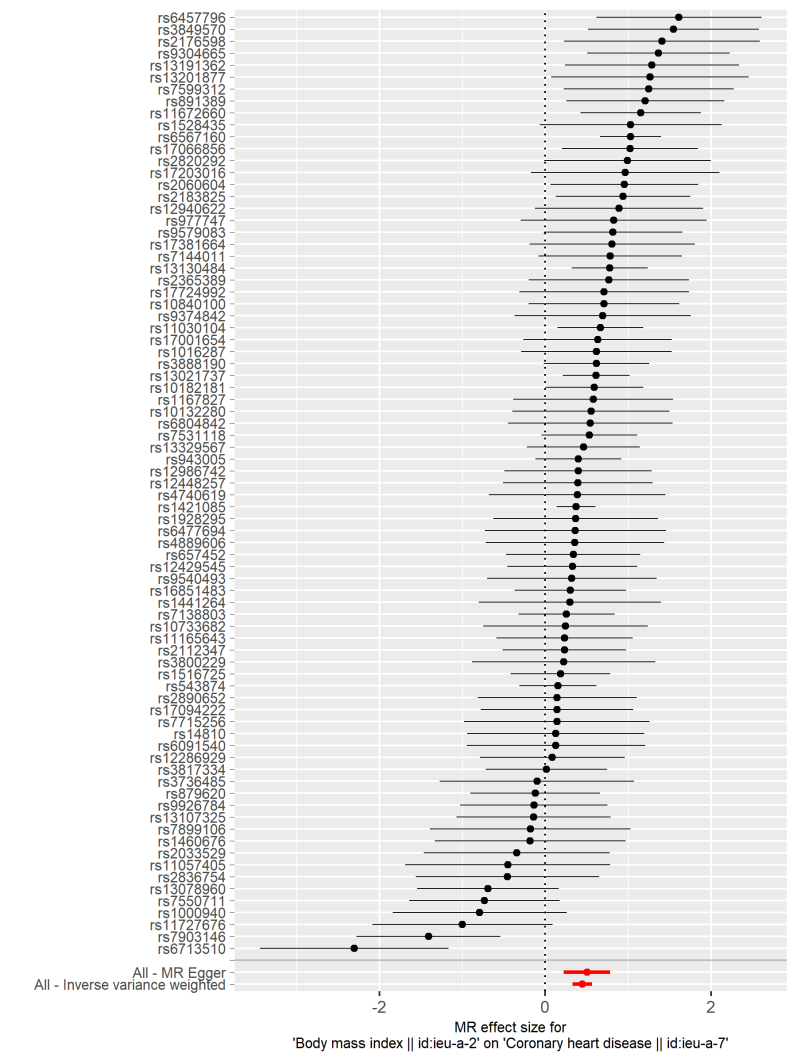
See **here** for more details.



MR analysis



https://mrcieu.github.io/TwoSampleMR/articles/perform_mr.html



Sensitivity analyses

- Evaluation of MR assumptions
- Check for the presence of pleiotropic effects
 - Heterogeneity of Wald estimates using Q-statistic or I^2
 - Intercept of MR-Egger
- Leave-one-out analysis to assess the reliance of MR analysis on a particular variant

Robust MR methods

- Valid causal inferences under weaker assumptions than the standard IVW methods
- Using multiple methods that make different assumptions about the nature of the underlying pleiotropy
- Most commonly used robust methods:
 - *MR-Egger* → TwoSampleMR and MendelianRandomization
 - *Median based method* → TwoSampleMR and MendelianRandomization
 - *Mode based method* → TwoSampleMR and MendelianRandomization
 - *MR-PRESSO* → MR-PRESSO R package
- A simulation study¹ has shown that two recent methods have good statistical properties in a wide range of scenarios:
 - *Contamination mixture* → implemented in MendelianRandomization with mr_conmix function
 - *MR-Mix* → MR-Mix R package

[1] Slob EAW, Burgess S. **A comparison of robust Mendelian randomization methods using summary data.** *Genetic Epidemiology*. 2020;44:313–29.

Other approaches to sensitivity analysis

- Using different datasets
- Positive and negative control outcomes
- Subgroup analyses
- Evaluate association with potentially pleiotropic variables

Case study

- **Aim:** to evaluate the causal relationship between kidney function and blood pressure (BP)
- **Exposure:** estimated glomerular filtration rate from serum creatinine (eGFR_{cr})
- **Outcome:** systolic blood pressure (SBP)

Case study

Two sets of IVs:

- 256 SNPs associated with eGFRcr at genome-wide threshold ($p - value < 5 \times 10^{-8}$)
- 40 SNPs associated with both eGFRcr and blood urea nitrogen (BUN) at genome-wide threshold ($p - value < 5 \times 10^{-8}$) → prioritizing IVs most likely relevant for kidney function

The case study was inspired the study from Yu et al. (2020)²

[2] Yu Z, Coresh J, Qi G, Grams M, Boerwinkle E, Snieder H, Teumer A, Pattaro C, Köttgen A, Chatterjee N, Tin A. **A bidirectional Mendelian randomization study supports causal effects of kidney function on blood pressure.** *Kidney Int.* 2020 Sep;98(3):708-716. doi: 10.1016/j.kint.2020.04.044. Epub 2020 May 23. PMID: 32454124; PMCID: PMC7784392.