'BWMR' Package for causal inference based on summary statistics

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1 Overview

This vignette provides an introduction to the 'BWMR' package. R package 'BWMR' implements BWMR (Bayesian Weighted Mendelian Randomization), an efficient statistical approach for causal inference based on summary statistics. It provides model parameter estimation as well as statistical inference.

The package can be loaded with the command:

> library("BWMR")

This vignette is organized as follows. Section 2.1 discusses the data processing precedure for the BWMR. Section 2.2 explains command lines for performing MR analysis and visualizing data and MR result using the 'BWMR' package.

Please feel free to contact Jia Zhao at jiazh0103@163.com for any questions or suggestions regarding the 'BWMR' package.

2 Workflow

2.1 Data preprocessing for the BWMR

In this vignette, we use an ExampleData (LDL cholesterol - dyslipidemia) from our real data analysis in the package. Users can find the SNP-exposure effects in the 'ExampleData\$beta.exposure', the SNP-outcome effects in the 'ExampleData\$beta.outcome', standard errors of SNP-exposure effects and SNP-outcome effects in 'ExampleData\$se.exposure' and 'ExampleData\$se.outcome respectively.

- > data(ExampleData)
- > head(ExampleData)

```
SNP beta.exposure beta.outcome se.exposure se.outcome 1 rs10102164 0.03147495 0.03342949 0.004482193 0.007697326 2 rs10195252 0.02206123 0.01835167 0.003615076 0.006201982 3 rs10401969 0.10908016 0.04495502 0.006633253 0.011908614 4 rs10490626 0.04634563 0.04448907 0.006294977 0.011214789 5 rs10832962 0.03025899 0.01336614 0.003782374 0.006907564 6 rs10893499 0.04761681 0.03467897 0.004843937 0.009269974
```

When we analyze real data, we need to ensure that the summary statistics for 'BWMR' are obtained through the following data preprocessing process:

- 1. Select SNPs that are significant for the exposure as instrument variables (IVs) by a p-value threshold (e.g. GWAS significant threshold 5×10^{-8}).
 - 2. Perform LD clumping to ensure that the IVs are independent.
 - 3. Extract the IVs from summary-level database for the outcomes of interest.
- 4. Perform harmonisation of effects to ensure that SNP-exposure effect and SNP-outcome effect correspond to the same allele.
 - 5. Scale the summary-level data if necessary.

2.2 Fitting the BWMR

We are now ready to fit BWMR using the data described above:

Besides, the 'BWMR' function also provides the visualization of data and MR result.

Plot of data with standard error bar:

Plot of evidence lower bound (elbo):

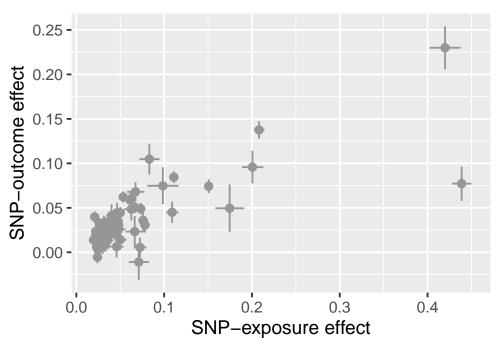
Posterior mean of weight of each observation:

Plot of weighted data and its regression result:

If we are only interested in the MR results, we can run the 'BWMR_quick' function without visualization, which is more time-efficient.

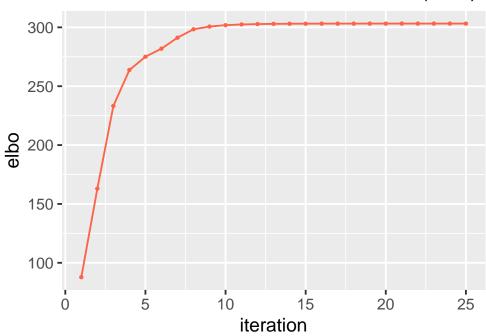
```
> fit.BWMR_quick <- BWMR_quick(gammahat = ExampleData$beta.exposure,
+ Gammahat = ExampleData$beta.outcome,
+ sigmaX = ExampleData$se.exposure,
+ sigmaY = ExampleData$se.outcome)</pre>
```

Plot1: Plot of data with standard error bar



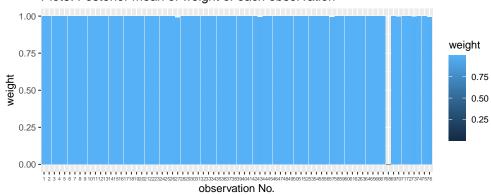
> fit.BWMR\$plot2

Plot2: Plot of evidence lower bound (elbo)



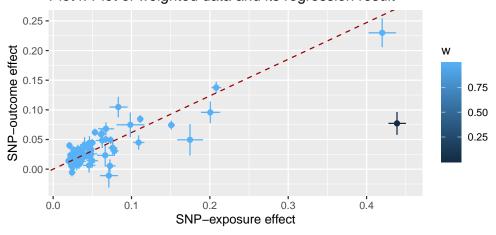
> fit.BWMR\$plot3

Plot3: Posterior mean of weight of each observation



> fit.BWMR\$plot4

Plot4: Plot of weighted data and its regression result



- > fit.BWMR_quick\$beta
- [1] 0.6177372
- > fit.BWMR_quick\$se_beta
- [1] 0.0256347
- > fit.BWMR_quick\$P_value
- [1] 2.642837e-128

References

Jia Zhao, Jingsi Ming, Xianghong Hu, Gang Chen, Jin Liu, Can Yang; Bayesian weighted Mendelian randomization for causal inference based on summary statistics.