

MetaboAnalyst 6.0

-- a unified platform for metabolomics data processing,

analysis and interpretation

Causal Analysis using Mendelian randomization

2024-03-10

Module Overview

This module offers functions to estimate the causal relationship between metabolites and phenotypes through Mendelian randomization (MR) analysis.

- ✓ There are many metabolomics-based genome-wide association studies (mGWAS) conducted to understanding the genetic regulations of metabolites in complex phenotype.
- ✓ By leveraging those SNP-tagged metabolites and summary statistics from public GWAS repositories, we can now test potential causal relationships between those genetically influenced metabolites and a disease outcome of interest using the well-established two-sample Mendelian randomization method.
- ✓ MR can estimate whether a relationship between a metabolite and a phenotype is causal, while reduce the impact of confounding factors and reverse causality that often plague observational studies.

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1. Introduction

Background

- It is now possible to estimate causal relationship between metabolites and a phenotype of interest.
- If a metabolite is causal for a given disease, genetic variants which influence the levels of that metabolite, either directly or indirectly, should result in a higher risk of the disease.

Data Formats

- No data upload required
- Users select an exposure (i.e., metabolites) and an outcome (i.e., diseases) of interest from the available options from our built-in databases

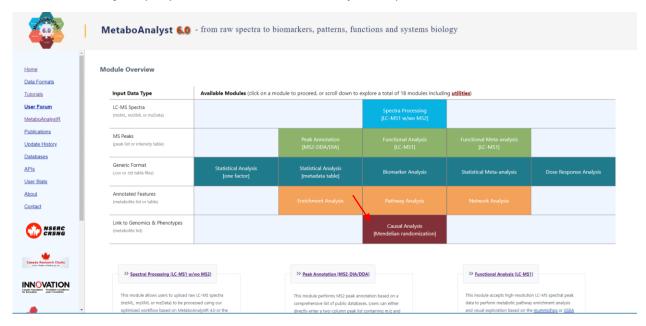
Expected Results

This module provides user comprehensive results on potential causal relationships between exposure-outcome based on two-sample MR

- i. Intermediate results from harmonization steps
- ii. The results from various statistical routines to estimate the causal effects and associated diagnostic and visualization plots

2. Choose the Module

Go to MetaboAnalyst (https://www.metaboanalyst.ca), and select the module



3. Causal Analysis via two-sample Mendelian randomisation (2SMR)

3.1 Specify metabolite and phenotype of interest

	Please specify metabolites (exposure) and	d outcome of interest	
	number of mGWAS studies and two-sample MR method permit c 1. Identify SNPs that are significantly associate with a metabo 2. Obtain the estimates of associations between these same 3. Perform SNP filtering and harmonize the effect sizes for SN 4. Conduct MR analysis, sensitivity analyses, and explore the ; Please note you may not be able to perform causal analysis in son 1. Select a metabolite of interest (exposure):	ausal analysis between metabolite and outcome of interes lite of interest from our large collections of the recent <u>mG</u> INPs with an outcome of interest from public repository. V IPs on the exposures and the outcomes to be for the same graphical outputs ne cases when no suitable SNPs are found in the two repo	WAS studies (covering > 4000 metabolites including their ratios); Ve use <u>Open GWAS Project</u> . reference allele.
Users should first select an exposure e., metabolites) and an outcome (i.e., iseases) of interest.	Q cys		 Search to choose the metabolite of interest (e.g. <i>Cystathionine</i>) from the left box. Once you select it, it will be automatically added into the right box.
J	2. Specify an outcome of interest: Enter a key word to see available options from the public reposito Type 2 diabetes finn-b-E4_DM2	Proceed Click "Proceed" to continue	For instance, we are interested in Type 2 diabetes. Type the name to see a list of match studies. Here we choose finn-b-E4_DM2

3.2 SNP filtering and harmonization

- Multiple SNPs could be identified as potential instrumental variables (IV) from the mGWAS and GWAS studies.
- > To perform proper 2SMR, the IVs should be
- Independent (i.e. not correlated with each other)
- Showing strong effect (i.e. significant p-values)
- No horizontal pleiotropy (i.e. affect the outcome only through the metabolite).
- Users need to carefully examine SNPs and apply different filtering and harmonization methods for each criterion

3.2 SNP Filtering and harmonization

To properly conduct two-sample MR analysis, the instrumental variables (IV) should be independent (i.e. not correlated with each other), showing strong effect (i.e. significant p-values), and no horizontal pleiotropy (i.e. affect the outcome only through the metabolite). The step provides following procedures to facilitate proper MR analysis: · Acquisition of independent IVs by performing linkage disequilibrium (LD) clumping. In cases where the SNP query is absent in the outcome GWAS, a proxy SNP in LD with the input SNP, utilizing the 1000 Genomes Project (phase 3). Harmonizing exposure and outcome data to make sure that the effects of the SNPs on exposure and outcome are associated with the same allele. You should also review the table below to perform further harmonization based on other metadata (such as population, study info, etc) · To control horizontal pleiotropy, you should manually exclude SNPs that are associated with multiple metabolites. O Do not check for LD between SNPs Harmonization steps require 1. LD Clumping Use clumping to prune SNPs for LD intensive computing and also O Do not use proxies access via remote server. 2. LD Proxies Use proxies and allow palindrome SNPs (advanced settings) Assume all alleles are presented on the forward strand It could take a long time or time 3. Allele Harmonization O Try to infer the forward strand alleles using allele frequency information Correct the strand for non-palindromic SNPs, but drop all palindromic SNPs out. Please be patient or try more time SNP ID ↑↓ Associated Metabolites ↑↓ Nearest Gene ↑↓ P-value Î↓ Biofluid 1↓ Population ↑↓ Study 1↓ Include L-Cystathionine (7) rs117782586 L-Cystathionine JRKL 4.637e-08 Blood European 28263315 rs146276253 L-Cystathionine ANKRD13C 3.436e-08 Blood European 28263315 rs150320192 L-Cystathionine SRSF11 3.566e-08 Blood European 28263315 1

3.3 Select statistical methods

 Wald ratio ? Maximum likelihood ? MR Egger ? Simple median ? Weighted median ? Inverse variance weighted radial ? MR Egger ? A total of 14 MR methods a offered currently. Some of them are more robust and o better tolerate violations of accumptions to cortain dog 	
 Inverse variance weighted (MRE) [®] Inverse variance weighted (FE) [®] Simple mode [®] Simple mode [®] Offered currently. Some of them are more robust and better tolerate violations of 	
Inverse variance weighted (MRE) ⁽¹⁾ Inverse variance weighted (FE) ⁽²⁾ simple mode ⁽²⁾ them are more robust and better tolerate violations of	are
Weighted mode (VOME) (V	
Sign concordance test ⑦ Unweighted regression ⑦	

- Mouse over the question marks for each method to see their main features.
- You can also find more detailed introduction on the forum: https://omicsforum.ca/t/what-are-the-differences-between-the-mr-analysis-methods/1045

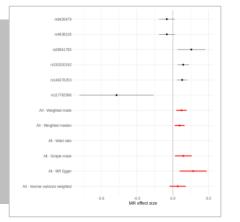
3.4 Mendelian randomization results

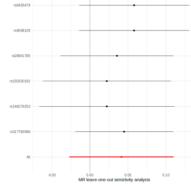
∨ L-Cystathionine										
Methods SNP Count	SND Count	Causal Effect Estimates		Heterogeneity Tests		Horizontal Pleiotropy				
	Beta	SE	P value	Q	Q_df	Q_pval	Egger Intercept	SE	P value	
Inverse variance weighted	6	0.041396	0.034939	0.23609	35.367	5	1.2709e-06	-	-	-
MR Egger	6	0.17071	0.057839	0.041906	14.006	4	0.0072767	-0.1179	0.047732	0.068948
Simple mode	6	0.088203	0.032068	0.040286	-	-	-	-	-	-
Weighted median	6	0.057145	0.021665	0.0083484	-	-	-	-	-	-
Weighted mode	6	0.072379	0.0219	0.021358	-	-	-	-	-	-

- The MR results are organized per metabolite (exposure).
- For metabolite, its shows the SNPs instrumental variables, along with their corresponding causal effect estimates, standard errors and p-values. Key values such as the MR-Egger regression intercept and its corresponding p-value are presented.
- Not all methods selected from the previous page would yield results depending of the data used.

3.4 Graphical outputs from MR analysis

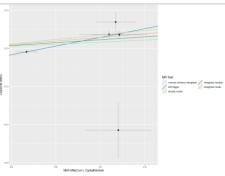
Forest plot compares the causal effect calculated using the methods that include all the SNPs to using each SNP separately.

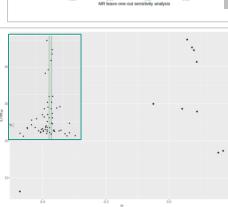




Leave one out sensitivity analysis: assesses whether a single SNP is having a disproportionately larger impact on an association. Each dot represents the MR analysis excluding that specific SNP using IVW method.

Scatter plot shows the relationships between SNP effects on exposure vs on the outcome. The slopes indicating the causal association





Funnel plot: Funnel shape will become more obvious with many SNPs (i.e. green box inset). Its asymmetry and wider spread may suggest horizontal pleiotropy.

4. Download Results

Download Results & Start New Journey

Please download the results (tables and images) from the **Results Download** tab below. The **Download.zip** contains all the files in your home directory. You can also generate a **PDF analysis report** using the button. Finally, you can continue to explore other compatible modules using the **Start New Journey** tab.

<u>Download.zip</u>	<u>mr_pleiotropy_results.csv</u>	,
<u>Rhistory.R</u>	L-Cystathionine mr forest plot 0 dpi72.png	All results can be
L-Cystathionine mr leaveoneout plot 0 dpi72.png	L-Cystathionine mr funnel plot 0 dpi72.png	downloaded here.
mr results.csv	mr heterogeneity results.csv	
L-Cystathionine mr scatter plot 0 dpi72.png	dis snp_restable.csv	
mr exposure data.csv		

In summary

If you have any questions, please read/post into <u>OmicsForum</u> (<u>www.omicsforum.ca</u>)

Or contact us:

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• Two-sample MR analysis allows researchers to estimate potential causal relationships between a metabolite and a phenotype of interest based on public data (mGWAS and GWAS summary statistics)

• Performing 2MSR requires identification of suitable SNPs (i.e. instrument variables) and performing filtering and harmonization. The process is computing intensive and better start with one metabolite at a time

• The module offers various MR methods with different strengths and limitations. They may give different estimates. Carefully examine the graphical outputs are necessary to reach robust conclusions