

Metabolomic Data Analysis with MetaboAnalyst 5.0

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1 Background

The Pathway Analysis module combines results from powerful pathway enrichment analysis with pathway topology analysis to help researchers identify the most relevant pathways involved in the conditions under study.

There are many commercial pathway analysis software tools such as Pathway Studio, MetaCore, or Ingenuity Pathway Analysis (IPA), etc. Compared to these commercial tools, the pathway analysis module was specifically developed for metabolomics studies. It uses high-quality KEGG metabolic pathways as the backend knowledgebase. This module integrates many well-established (i.e. univariate analysis, over-representation analysis) methods, as well as novel algorithms and concepts (i.e. Global Test, GlobalAncova, network topology analysis) into pathway analysis. Another feature is a Google-Map style interactive visualization system to deliver the analysis results in an intuitive manner.

2 Data Input

The Pathway Analysis module accepts either a list of compound labels (common names, HMDB IDs or KEGG IDs) with one compound per row, or a compound concentration table with samples in rows and compounds in columns. The second column must be phenotype labels (binary, multi-group, or continuous). The table is uploaded as comma separated values (.csv).

3 Compound Name Matching

The first step is to standardize the compound labels used in user uploaded data. This is a necessary step since these compounds will be subsequently compared with compounds contained in the pathway library. There are three outcomes from the step - exact match, approximate match (for common names only), and no match. Users should click the textbfView button from the approximate matched results to manually select the correct one. Compounds without match will be excluded from the subsequently pathway analysis.

Table 1 shows the conversion results. Note: *1* indicates exact match, *2* indicates approximate match, and *0* indicates no match. A text file contain the result can be found the downloaded file *name_map.csv*

Table 1: Result from Compound

	Query	Match	HMDB	PubChem	KEGG	SMILES
1	HMDB0000763	5-Hydroxyindoleacetic acid	HMDB0000763	1826	C05635	<chem>C1=CC2=C(C(=C1O)C(=CN2)CC(=O)O</chem>
2	HMDB0000023	(S)-3-Hydroxyisobutyric acid	HMDB0000023	87	C06001	<chem>C[C@@H](CO)C(=O)O</chem>
3	HMDB0000355	3-Hydroxymethylglutaric acid	HMDB0000355	1662	C03761	<chem>CC(CC(=O)O)(CC(=O)O)O</chem>
4	HMDB0002064	N-Acetylputrescine	HMDB0002064	122356	C02714	<chem>CC(=O)NCCCCN</chem>
5	HMDB0000191	L-Aspartic acid	HMDB0000191	5960	C00049	<chem>C([C@@H](C(=O)O)N)C(=O)O</chem>
6	HMDB0000755	Hydroxyphenyllactic acid	HMDB0000755	9378	C03672	<chem>C1=CC(=CC=C1CC(C(=O)O)O)O</chem>
7	HMDB0000033	Carnosine	HMDB0000033	439224	C00386	<chem>C1=C(NC=N1)C[C@@H](C(=O)O)NC(=O)O</chem>
8	HMDB0000254	Succinic acid	HMDB0000254	1110	C00042	<chem>C(CC(=O)O)C(=O)O</chem>
9	HMDB0001401	Glucose 6-phosphate	HMDB0001401	5958	C00092	<chem>C([C@@H]1[C@H]([C@@H]([C@H](C(O1)O</chem>

97	HMDB0061652	3-hydroxyhexanoic acid	HMDB0061652	11829482		CCC[C@H](O)CC(O)=O
98	HMDB0240296	NA	NA	NA	NA	NA
99	HMDB0060014	NA	NA	NA	NA	NA
100	HMDB10737	(R)-3-Hydroxy-Octadecanoic acid	HMDB0010737	5312838		CCCCCCCCCCCCCCCC[C@H](CC(=O)O)C(=O)O
101	HMDB0007969	PC(16:0/16:1(9Z))	HMDB0007969	6443788	C00157	CCCCCCCCCCCCCCCCC(=O)OC[C@H](O)C(=O)O
102	HMDB0008046	PC(18:0/20:3(5Z,8Z,11Z))	HMDB0008046	24778855	C00157	CCCCCCCCCCCCCCCCCCC(=O)OC[C@H](O)C(=O)O
103	HMDB0240328	NA	NA	NA	NA	NA
104	HMDB0240644	NA	NA	NA	NA	NA
105	HMDB0013676	2,6-Dihydroxybenzoic acid	HMDB0013676	9338	C21298	C1=CC(=C(C(=C1)O)C(=O)O)O
106	HMDB0240388	NA	NA	NA	NA	NA
107	HMDB0011176	L-phenylalanyl-L-hydroxyproline	HMDB0011176	53480675		C1C(CN(C1C(=O)O)C(=O)C(CC2=CC=CC=C2)C(=O)O)C(=O)O
108	HMDB0028930	Leucyl-Hydroxyproline	HMDB0028930	20847829		CC(C)CC(N)C(=O)N1CC(O)CC1C(O)=O
109	HMDB0028908	Isoleucyl-Hydroxyproline	HMDB0028908	61158802		CCC(C)C(N)C(=O)N1CC(O)CC1C(O)=O
110	HMDB0037115	NA	NA	NA	NA	NA

4 Pathway Analysis

In this step, users are asked to select a pathway library, as well as specify the algorithms for pathway enrichment analysis and pathway topology analysis.

4.1 Pathway Library

There are 15 pathway libraries currently supported, with a total of 1173 pathways :

- Homo sapiens (human) [80]
- Mus musculus (mouse) [82]
- Rattus norvegicus (rat) [81]
- Bos taurus (cow) [81]
- Danio rerio (zebrafish) [81]
- Drosophila melanogaster (fruit fly) [79]
- Caenorhabditis elegans (nematode) [78]
- Saccharomyces cerevisiae (yeast) [65]
- Oryza sativa japonica (Japanese rice) [83]
- Arabidopsis thaliana (thale cress) [87]
- Escherichia coli K-12 MG1655 [87]
- Bacillus subtilis [80]
- Pseudomonas putida KT2440 [89]
- Staphylococcus aureus N315 (MRSA/VSSA)[73]
- Thermotoga maritima [57]

Your selected pathway library code is **mmu** (KEGG organisms abbreviation).

4.2 Over Representation Analysis

Over-representation analysis tests if a particular group of compounds is represented more than expected by chance within the user uploaded compound list. In the context of pathway analysis, we are testing if compounds involved in a particular pathway are enriched compared to random hits. MetPA offers two of the most commonly used methods for over-representation analysis:

- Fishers'Exact test
- Hypergeometric Test

Please note, MetPA uses one-tailed Fisher's exact test which will give essentially the same result as the result calculated by the hypergeometric test.

The selected over-representation analysis method is **Hypergeometric test**.

4.3 Pathway Topology Analysis

The structure of biological pathways represent our knowledge about the complex relationships among molecules within a cell or a living organism. However, most pathway analysis algorithms fail to take structural information into consideration when estimating which pathways are significantly changed under conditions of study. It is well-known that changes in more important positions of a network will trigger a more severe impact on the pathway than changes occurred in marginal or relatively isolated positions.

The pathway topology analysis uses two well-established node centrality measures to estimate node importance - **degree centrality** and **betweenness centrality**. Degree centrality is defined as the number of links occurred upon a node. For a directed graph there are two types of degree: in-degree for links come from other nodes, and out-degree for links initiated from the current node. Metabolic networks are directed graph. Here we only consider the out-degree for node importance measure. It is assumed that nodes upstream will have regulatory roles for the downstream nodes, not vice versa. The betweenness centrality measures the number of shortest paths going through the node. Since the metabolic network is directed, we use the relative betweenness centrality for a metabolite as the importance measure. The degree centrality measure focuses more on local connectivities, while the betweenness centrality measure focuses more on global network topology. For more detailed discussions on various graph-based methods for analyzing biological networks, please refer to the article by Tero Aittokallio, T. et al. ¹

Please note, for comparison among different pathways, the node importance values calculated from centrality measures are further normalized by the sum of the importance of the pathway. Therefore, the total/maximum importance of each pathway is 1; the importance measure of each metabolite node is actually the percentage w.r.t the total pathway importance, and the pathway impact value is the cumulative percentage from the matched metabolite nodes.

Your selected node importance measure for topological analysis is **out degree centrality**.

5 Pathway Analysis Result

The results from pathway analysis are presented graphically as well as in a detailed table.

A Google-map style interactive visualization system was implemented to facilitate data exploration. The graphical output contains three levels of view: **metabolome view**, **pathway view**, and **compound view**. Only the metabolome view is shown below. Pathway views and compound views are generated dynamically based on your interactions with the visualization system. They are available in your downloaded files.

¹Tero Aittokallio and Benno Schwikowski. *Graph-based methods for analyzing networks in cell biology*, Briefings in Bioinformatics 2006 7(3):243-255

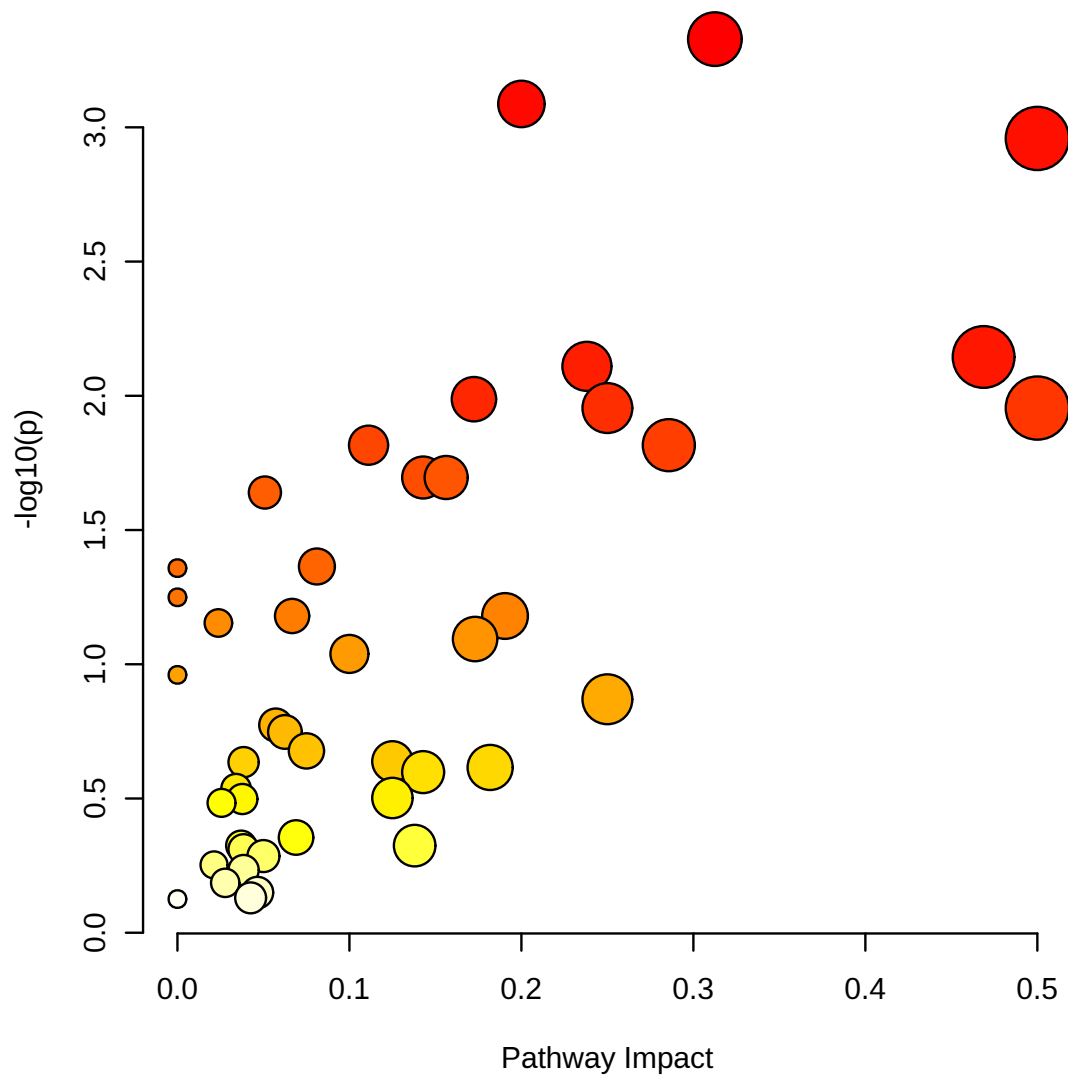


Figure 1: Summary of Pathway Analysis

The table below shows the detailed results from the pathway analysis. Since we are testing many pathways at the same time, the statistical p values from enrichment analysis are further adjusted for multiple testings. In particular, the **Total** is the total number of compounds in the pathway; the **Hits** is the actually matched number from the user uploaded data; the **Raw p** is the original p value calculated from the enrichment analysis; the **Holm p** is the p value adjusted by Holm-Bonferroni method; the **FDR p** is the p value adjusted using False Discovery Rate; the **Impact** is the pathway impact value calculated from pathway topology analysis.

Table 2: Result from Pathway Analysis

	Total	Expected	Hits	Raw p	-log10(p)	Holm adjust	FDR	Impact
Arginine biosynthesis	14	0.40	4	4.69E-04	3.33E+00	3.94E-02	3.08E-02	0.31
Histidine metabolism	16	0.46	4	8.18E-04	3.09E+00	6.79E-02	3.08E-02	0.20
Taurine and hypotaurine metabolism	8	0.23	3	1.10E-03	2.96E+00	9.02E-02	3.08E-02	0.50
Alanine, aspartate and glutamate metabolism	28	0.80	4	7.17E-03	2.14E+00	5.81E-01	1.17E-01	0.47
Nicotinate and nicotinamide metabolism	15	0.43	3	7.77E-03	2.11E+00	6.21E-01	1.17E-01	0.24
Aminoacyl-tRNA biosynthesis	48	1.37	5	1.03E-02	1.99E+00	8.13E-01	1.17E-01	0.17
Nitrogen metabolism	6	0.17	2	1.11E-02	1.95E+00	8.67E-01	1.17E-01	0.25
D-Glutamine and D-glutamate metabolism	6	0.17	2	1.11E-02	1.95E+00	8.67E-01	1.17E-01	0.50
Thiamine metabolism	7	0.20	2	1.53E-02	1.82E+00	1.00E+00	1.28E-01	0.29
Pantothenate and CoA biosynthesis	19	0.54	3	1.53E-02	1.82E+00	1.00E+00	1.28E-01	0.11
beta-Alanine metabolism	21	0.60	3	2.02E-02	1.70E+00	1.00E+00	1.41E-01	0.14
Sphingolipid metabolism	21	0.60	3	2.02E-02	1.70E+00	1.00E+00	1.41E-01	0.16
Pyrimidine metabolism	39	1.11	4	2.29E-02	1.64E+00	1.00E+00	1.48E-01	0.05
Glutathione metabolism	28	0.80	3	4.32E-02	1.36E+00	1.00E+00	2.46E-01	0.08
Phenylalanine metabolism	12	0.34	2	4.38E-02	1.36E+00	1.00E+00	2.46E-01	0.00
Neomycin, kanamycin and gentamicin biosynthesis	2	0.06	1	5.63E-02	1.25E+00	1.00E+00	2.96E-01	0.00
Butanoate metabolism	15	0.43	2	6.61E-02	1.18E+00	1.00E+00	3.08E-01	0.07
Starch and sucrose metabolism	15	0.43	2	6.61E-02	1.18E+00	1.00E+00	3.08E-01	0.19
Glycine, serine and threonine metabolism	34	0.97	3	7.02E-02	1.15E+00	1.00E+00	3.10E-01	0.02
Glycerophospholipid metabolism	36	1.03	3	8.05E-02	1.09E+00	1.00E+00	3.38E-01	0.17
Arginine and proline metabolism	38	1.08	3	9.15E-02	1.04E+00	1.00E+00	3.66E-01	0.10
Phenylalanine, tyrosine and tryptophan biosynthesis	4	0.11	1	1.10E-01	9.61E-01	1.00E+00	4.18E-01	0.00
Linoleic acid metabolism	5	0.14	1	1.35E-01	8.70E-01	1.00E+00	4.93E-01	0.25
Glycolysis / Gluconeogenesis	26	0.74	2	1.68E-01	7.74E-01	1.00E+00	5.89E-01	0.06
Galactose metabolism	27	0.77	2	1.79E-01	7.48E-01	1.00E+00	6.00E-01	0.06
Inositol phosphate metabolism	30	0.86	2	2.10E-01	6.77E-01	1.00E+00	6.79E-01	0.07
Ubiquinone and other terpenoid-quinone biosynthesis	9	0.26	1	2.30E-01	6.38E-01	1.00E+00	6.95E-01	0.12
Glyoxylate and dicarboxylate metabolism	32	0.91	2	2.32E-01	6.35E-01	1.00E+00	6.95E-01	0.04
Cysteine and methionine metabolism	33	0.94	2	2.42E-01	6.16E-01	1.00E+00	7.02E-01	0.18
Ascorbate and aldarate metabolism	10	0.29	1	2.52E-01	5.98E-01	1.00E+00	7.06E-01	0.14
Purine metabolism	66	1.88	3	2.91E-01	5.37E-01	1.00E+00	7.88E-01	0.03
alpha-Linolenic acid metabolism	13	0.37	1	3.15E-01	5.02E-01	1.00E+00	8.09E-01	0.12
Valine, leucine and isoleucine degradation	40	1.14	2	3.18E-01	4.98E-01	1.00E+00	8.09E-01	0.04
Tryptophan metabolism	41	1.17	2	3.28E-01	4.84E-01	1.00E+00	8.11E-01	0.03
Citrate cycle (TCA cycle)	20	0.57	1	4.42E-01	3.55E-01	1.00E+00	1.00E+00	0.07
Pyruvate metabolism	22	0.63	1	4.74E-01	3.25E-01	1.00E+00	1.00E+00	0.04
Pentose phosphate pathway	22	0.63	1	4.74E-01	3.25E-01	1.00E+00	1.00E+00	0.14
Propanoate metabolism	23	0.66	1	4.89E-01	3.11E-01	1.00E+00	1.00E+00	0.04
Lysine degradation	25	0.71	1	5.18E-01	2.86E-01	1.00E+00	1.00E+00	0.05
Phosphatidylinositol signaling system	28	0.80	1	5.59E-01	2.53E-01	1.00E+00	1.00E+00	0.02
Porphyrin and chlorophyll metabolism	30	0.86	1	5.84E-01	2.33E-01	1.00E+00	1.00E+00	0.04
Arachidonic acid metabolism	36	1.03	1	6.52E-01	1.86E-01	1.00E+00	1.00E+00	0.03
Tyrosine metabolism	42	1.20	1	7.09E-01	1.49E-01	1.00E+00	1.00E+00	0.05
Primary bile acid biosynthesis	46	1.31	1	7.42E-01	1.30E-01	1.00E+00	1.00E+00	0.04
Fatty acid biosynthesis	47	1.34	1	7.49E-01	1.25E-01	1.00E+00	1.00E+00	0.00

6 Appendix: R Command History

```
[1] "mSet<-InitDataObjects(\"conc\", \"pathora\", FALSE)"
[2] "cmpd.vec<-c(\"HMDB0000763\", \"HMDB0000023\", \"HMDB0000355\", \"HMDB0002064\", \"HMDB0000191\", \"")
[3] "mSet<-Setup.MapData(mSet, cmpd.vec);"
[4] "mSet<-CrossReferencing(mSet, \"hmdb\");"
[5] "mSet<-CreateMappingResultTable(mSet)"
[6] "mSet<-PerformDetailMatch(mSet, \"HMDB0240577\");"
[7] "mSet<-GetCandidateList(mSet);"
[8] "mSet<-SetKEGG.PathLib(mSet, \"mmu\", \"current\")"
[9] "mSet<-SetMetabolomeFilter(mSet, F);"
[10] "mSet<-CalculateOraScore(mSet, \"rbc\", \"hyperg\")"
[11] "mSet<-PlotPathSummary(mSet, F, \"path_view_0_\", \"png\", 72, width=NA, NA, NA )"
[12] "mSet<-SetKEGG.PathLib(mSet, \"mmu\", \"current\")"
[13] "mSet<-SetMetabolomeFilter(mSet, F);"
[14] "mSet<-CalculateOraScore(mSet, \"dgr\", \"hyperg\")"
[15] "mSet<-PlotPathSummary(mSet, F, \"path_view_1_\", \"png\", 72, width=NA, NA, NA )"
[16] "mSet<-SetKEGG.PathLib(mSet, \"mmu\", \"current\")"
[17] "mSet<-SetMetabolomeFilter(mSet, F);"
[18] "mSet<-CalculateOraScore(mSet, \"rbc\", \"hyperg\")"
[19] "mSet<-PlotPathSummary(mSet, F, \"path_view_2_\", \"png\", 72, width=NA, NA, NA )"
[20] "mSet<-SetKEGG.PathLib(mSet, \"mmu\", \"current\")"
[21] "mSet<-SetMetabolomeFilter(mSet, F);"
[22] "mSet<-CalculateOraScore(mSet, \"dgr\", \"hyperg\")"
[23] "mSet<-PlotPathSummary(mSet, F, \"path_view_3_\", \"png\", 72, width=NA, NA, NA )"
[24] "mSet<-SaveTransformedData(mSet)"
[25] "mSet<-PreparePDFReport(mSet, \"guest17886441071731322890\")\n"
```

The report was generated on Fri Jun 9 10:51:41 2023 with R version 4.2.2 (2022-10-31), OS system: Linux, version: -Ubuntu SMP Wed Feb 22 14:14:39 UTC 2023 .