Metabolomic Data Analysis with MetaboAnalyst 6.0

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May 20, 2024

1 Background

MSEA or Metabolite Set Enrichment Analysis is a way to identify biologically meaningful patterns that are significantly enriched in quantitative metabolomic data. In conventional approaches, metabolites are evaluated individually for their significance under conditions of study. Those compounds that have passed certain significance level are then combined to see if any meaningful patterns can be discerned. In contrast, MSEA directly investigates if a set of functionally related metabolites without the need to preselect compounds based on some arbitrary cut-off threshold. It has the potential to identify subtle but consistent changes among a group of related compounds, which may go undetected with the conventional approaches.

Essentially, MSEA is a metabolomic version of the popular GSEA (Gene Set Enrichment Analysis) software with its own collection of metabolite set libraries as well as an implementation of user-friendly web-interfaces. GSEA is widely used in genomics data analysis and has proven to be a powerful alternative to conventional approaches. For more information, please refer to the original paper by Subramanian A, and a nice review paper by Nam D, Kim SY. ¹. ²

2 MSEA Overview

Metabolite set enrichment analysis consists of four steps - data input, data processing, data analysis, and results download. Different analysis procedures are performed based on different input types. In addition, users can also browse and search the metabolite set libraries as well as upload their self-defined metabolite sets for enrichment analysis. Users can also perform metabolite name mapping between a variety of compound names, synonyms, and major database identifiers.

3 Data Input

There are three enrichment analysis algorithms offered by MSEA. Accordingly, three different types of data inputs are required by these three approaches:

- A list of important compound names entered as a one column data (Over Representation Analysis (ORA));
- A single measured biofluid (urine, blood, CSF) sample- entered as tab separated two-column data with the first column for compound name, and the second for concentration values (Single Sample Profiling (SSP));

¹Subramanian Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles., Proc Natl Acad Sci USA. 2005 102(43): 15545-50

²Nam D, Kim SY. Gene-set approach for expression pattern analysis, Briefings in Bioinformatics. 2008 9(3): 189-197.

• A compound concentration table - entered as a comma separated (.csv) file with the each sample per row and each metabolite concentration per column. The first column is sample names and the second column for sample phenotype labels (Quantitative Enrichment Analysis (QEA))

You selected Over Representation Analysis (ORA) which requires a list of compound names as input.

4 Data Process

The first step is to standardize the compound labels. It is an essential step since the compound labels will be subsequently compared with compounds contained in the metabolite set library. MSEA has a built-in tool to convert between compound common names, synonyms, identifiers used in HMDB ID, PubChem, ChEBI, BiGG, METLIN, KEGG, or Reactome. 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 Com

	Query	Match	HMDB	PubChem	KEGG	SMILES
1	O-Butanoylcarnitine	Butyrylcarnitine	HMDB0002013	213144	C02862	CCCC(=O)O[C@H](CC([O-])
2	Glut ary l-carnitine	Glutarylcarnitine	HMDB0013130	53481699		C[N+](C)(C)C[C@@H](CC(O
3	Arginine-Glutamine	Arginylglutamine	${ m HMDB0028707}$	7019985		N[C@@H](CCCNC(N)=N)C(
4	Deoxycarnitine	4-Trimethy lammoniobutanoic acid	HMDB0001161	725	C01181	C[N+](C)(C)CCCC([O-])=O
5	N-carbomoyl-L-aspartate	Ureidosuccinic acid	${\rm HMDB0000828}$	93072	C00438	NC(=O)N[C@@H](CC(O)=O
6	Deoxyadenosine	Deoxyadenosine	HMDB0000101	13730	C00559	NC1=C2N=CN([C@H]3C[C@H]
7	5-Methylcytosine	5-Methylcytosine	${\rm HMDB0002894}$	65040	C02376	CC1=C(N)NC(=O)N=C1
8	Raffinose	Raffinose	HMDB0003213	439242	C00492	OC[C@H]10[C@@](CO)(O[C
9	Glucose	D-Glucose	${ m HMDB0000122}$	5793	C00031	OC[C@H]1O[C@@H](O)[C@H]
10	L-Octanoylcarnitine	Octanoylcarnitine	${\rm HMDB0000791}$	11953814	C02838	CCCCCCC(=0)O[C@H](CCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCC
11	Malate	Malic acid	${ m HMDB0000156}$	222656	C00149	O[C@@H](CC(O)=O)C(O)=O
12	Creatinine	Creatinine	${ m HMDB0000562}$	588	C00791	CN1CC(=O)NC1=N
13	Lysine-Glut amine	Lysylglutamine	${ m HMDB0028949}$	196305		NCCCC[C@H](N)C(=O)N[C@H]
14	Serine	Serine	HMDB0000187	5951	C00065	N[C@@H](CO)C(O)=O
15	Arginine-Valine	Arginylvaline	${\rm HMDB0028722}$	6992654		CC(C)[C@H](NC(=O)[C@@H
16	Guanidinoacetate	Guanidoacetic acid	${ m HMDB0000128}$	763	C00581	NC(=N)NCC(O)=O
17	Uric acid	Uric acid	${\rm HMDB0000289}$	1175	C00366	O=C1NC2=C(N1)C(=O)NC
18	AMP	Adenosine monophosphate	${ m HMDB0000045}$	6083	C00020	NC1=C2N=CN([C@@H]3O[C]
19	Hypotaurine	Hypotaurine	${ m HMDB0000965}$	107812	C00519	NCCS(O) = O
20	Guanine	Guanine	HMDB0000132	764	C00242	NC1=NC(=O)C2=C(N1)N=C
21	Asparagine	L-Asparagine	${\rm HMDB0000168}$	6267	C00152	N[C@@H](CC(N)=O)C(O)=0
22	Phenethylamine	Pheny lethy lamine	${ m HMDB0012275}$	1001	C05332	NCCC1=CC=CC=C1
23	Thiamine	Thiamine	${ m HMDB0000235}$	1130	C00378	CC1=C(CCO)SC=[N+]1CC1
24	Histidine	Histidine	HMDB0000177	6274	C00135	N[C@@H](CC1=CN=CN1)C(
25	Cyclic AMP	Cyclic AMP	${ m HMDB0000058}$	6076	C00575	[H][C@@]12COP(O)(=O)O[C]
26	Hypoxanthine	Hypoxanthine	${ m HMDB0000157}$	790	C00262	OC1=NC=NC2=C1NC=N2
27	2-Hydroxyglutarate	2-Hydroxy glutarate	${ m HMDB0059655}$	43	C02630	OC(CCC(O)=O)C(O)=O
28	Nicotinamide riboside	Nicotinamide riboside	${ m HMDB0000855}$	439924	C03150	NC(=O)C1=C[N+](=CC=C1
29	Glutamine	Glutamine	HMDB0000641	5961	C00064	N[C@@H](CCC(N)=O)C(O)=
30	N-acetyl-L-ornithine	N2-Acetylornithine	${ m HMDB0003357}$	439232	C00437	CC(=O)N[C@@H](CCCN)C(
31	NADH	NADH	${ m HMDB0001487}$	439153	C00004	NC(=O)C1=CN(C=CC1)[CQ]
32	Cellobiose	Cellobiose	${ m HMDB0000055}$	10712	C00185	OC[C@H]1O[C@@H](O[C@H]
33	Norvaline	Norvaline	${ m HMDB0013716}$	439575	C01799	CCC[C@@H](N)C(O)=O
34	Indole	Indole	${ m HMDB0000738}$	798	C00463	N1C=CC2=C1C=CC=C2
35	Orotate	Orotic acid	${ m HMDB0000226}$	967	C00295	OC(=O)C1=CC(=O)NC(=O)
36	Fructose	D-Fructose	${ m HMDB0000660}$	439709	C00095	OC[C@H]1O[C@](O)(CO)[C@
37	Nicotinate	Nicotinic acid	$\mathrm{HMDB0001488}$	938	C00253	OC(=O)C1=CN=CC=C1
38	5- Methylthioadenosine	5'-Methylthioadenosine	${ m HMDB0001173}$	439176	C00170	CSC[C@H]1O[C@H]([C@H](C
39	Ornithine	Ornithine	${ m HMDB0000214}$	6262	C00077	NCCC[C@H](N)C(O)=O
40	Uridine	Uridine	${ m HMDB0000296}$	6029	C00299	OC[C@H]1O[C@H]([C@H](O)
41	Phosphocholine	Phosphorylcholine	${ m HMDB0001565}$	1014	C00588	C[N+](C)(C)CCOP(O)(O)=0
42	Cytosine	Cytosine	${ m HMDB0000630}$	597	C00380	NC1=CC=NC(=O)N1
43	L-Palmitoy lcarnitine	L-Palmitoylcarnitine	${ m HMDB0240774}$	16902		[H][C@](CC([O-])=O)(C[N+])
44	UMP	Uridine 5'-monophosphate	HMDB0000288	6030	C00105	O[C@H]1[C@@H](O)[C@@H](
45	Glucuronic acid	D-Glucuronic acid	${ m HMDB0000127}$	444791	C00191	O[C@H]1O[C@@H]([C@@H](
46	Carnitine	L-Carnitine	${ m HMDB0000062}$	10917	C00487	C[N+](C)(C)C[C@H](O)CC(C
47	ADP-D-Glucose	ADP-glucose	${ m HMDB0006557}$	16500	C00498	NC1 = C2N = CN([C@@H]3O[C])
48	Montiporic Acid C	NA	NA	NA	NA	NA
49	Montiporic Acid D	NA	NA	NA	NA	NA
50	Dihydroxyacetone phosphate	Dihydroxyacetone phosphate	HMDB0001473	668	C00111	OCC(=O)COP(O)(O)=O
51	Montiporic Acid A	NA	NA	NA	NΑ	NA
52	Nicotinamide ribotide	Nicotinamide ribotide	${ m HMDB0000229}$	14181	C00455	NC(=O)C1=C[N+](=CC=C1
53	Montiporic Acid B	NA	NA	NA	NA	NA
54	Porphobilinogen	Porphobilinogen	${ m HMDB0000245}$	1021	C00931	NCC1=C(CC(O)=O)C(CCC(O)=O)
55	Tryptophan	L-Tryptophan	${ m HMDB0000929}$	6305	C00078	N[C@@H](CC1=CNC2=C1C)
56	a-ketoglutarate	Oxoglutaric acid	${ m HMDB0000208}$	51	C00026	OC(=O)CCC(=O)C(O)=O
57	Citrate	Citric acid	${ m HMDB0000094}$	311	C00158	OC(=O)CC(O)(CC(O)=O)C
58	ADP	ADP	HMDB0001341	6022	C00008	NC1=NC=NC2=C1N=CN2[0

59	Thymidine	Thymidine	${ m HMDB0000273}$	5789	C00214	CC1=CN([C@H]2C[C@H](O)
60	N-octanoylglycine	Capryloylglycine	${ m HMDB0000832}$	84290		CCCCCCC(=0)NCC(0)=0
61	ADP-ribose	Adenosine diphosphate ribose	${ m HMDB0001178}$	192	C00301	NC1=C2N=ČN(Ć3OC(ČÓP(
62	1-Methylimidazole acetic acid	Methylimidazoleacetic acid	${ m HMDB0002820}$	75810	C05828	CN1C=NC(CC(O)=O)=C1
63	GMP	Guanosine monophosphate	${ m HMDB0001397}$	6804	C00144	NC1=NC2=C(N=CN2[C@@H
64	Cresol	p-Cresol	${ m HMDB0001858}$	2879	C01468	CC1=CC=C(O)C=C1
65	Itaconic acid	Itaconic acid	${ m HMDB0002092}$	811	C00490	OC(=O)CC(=C)C(O)=O
66	Carbamoyl phosphate	Carbamoyl phosphate	$_{ m HMDB0001096}$	278	C00169	NC(=O)OP(O)(O)=O
67	Phenylalanine	Phenylalanine	${ m HMDB0000159}$	6140	C00079	N[C@@H](CC1=CC=CC=C1
68	5-methylthioadenosine	5'-Methylthioadenosine	HMDB0001173	439176	C00170	CSC[C@Ĥ]10[С@Н]([С@Н](С
69	Isoleucine	Isoleucine	${ m HMDB0000172}$	6306	C00407	CC[C@H](C)[C@H](N)C(O)=
70	4-Aminobutyrate	gamma-Aminobutyric acid	${ m HMDB0000112}$	119	C00334	NCCCC(O) = O
71	GTP	Guanosine triphosphate	${\rm HMDB0001273}$	6830	C00044	NC1=NC2=C(N=CN2[C@@H
72	Glutamate	Glutamic acid	HMDB0000148	33032	C00025	N[C@@H](CCC(O)=O)C(O)=
73	ATP	Adenosine triphosphate	${ m HMDB0000538}$	5957	C00002	NC1=NC=NC2=C1N=CN2[C
74	Hippuric acid	Hippuric acid	${ m HMDB0000714}$	464	C01586	OC(=O)CNC(=O)C1=CC=O
75	Methionine	Methionine	$_{ m HMDB0000696}$	6137	C00073	CSCC[C@H](N)C(O)=O
76	Pterin	Pterin	${ m HMDB0000802}$	73000	C00715	NC1=NC2=NC=CN=C2C(=
77	Glucose-6-phosphate	Glucose 6-phosphate	${ m HMDB0001401}$	5958	C00092	OC1O[C@H](COP(O)(O)=O)
78	Lumichrome	NA	NA	NA	NA	NA
79	Succinate	Succinic acid	${\rm HMDB0000254}$	1110	C00042	OC(=O)CCC(O)=O
80	Coenzyme A	Coenzyme A	${\rm HMDB0001423}$	87642	C00010	CC(C)(COP(O)(=O)OP(O)(O)(O)OP(O)(O)(O)(O)(O)(O)(O)(O)(O)(O)(O)(O)(O)(
			-	·		

The second step is to check concentration values. For SSP analysis, the concentration must be measured in umol for blood and CSF samples. The urinary concentrations must be first converted to $umol/mmol_creatinine$ in order to compare with reported concentrations in literature. No missing or negative values are allowed in SSP analysis. The concentration data for QEA analysis is more flexible. Users can upload either the original concentration data or normalized data. Missing or negative values are allowed (coded as NA) for QEA.

5 Selection of Metabolite Set Library

Before proceeding to enrichment analysis, a metabolite set library has to be chosen. There are seven built-in libraries offered by MSEA:

- Metabolic pathway associated metabolite sets (currently contains 99 entries);
- Disease associated metabolite sets (reported in blood) (currently contains 344 entries);
- Disease associated metabolite sets (reported in urine) (currently contains 384 entries)
- Disease associated metabolite sets (reported in CSF) (currently contains 166 entries)
- Metabolite sets associated with SNPs (currently contains 4598 entries)
- Predicted metabolite sets based on computational enzyme knockout model (currently contains 912 entries)
- Metabolite sets based on locations (currently contains 73 entries)
- Drug pathway associated metabolite sets (currently contains 461 entries)

In addition, MSEA also allows user-defined metabolite sets to be uploaded to perform enrichment analysis on arbitrary groups of compounds which researchers want to test. The metabolite set library is simply a two-column comma separated text file with the first column for metabolite set names and the second column for its compound names (**must use HMDB compound name**) separated by "; ". Please note, the built-in libraries are mainly from human studies. The functional grouping of metabolites may not be valid. Therefore, for data from subjects other than human being, users are suggested to upload their self-defined metabolite set libraries for enrichment analysis.

6 Enrichment Analysis

Over Representation Analysis (ORA) is performed when a list of compound names is provided. The list of compound list can be obtained through conventional feature selection methods, or from a clustering algorithm, or from the compounds with abnormal concentrations detected in SSP, to investigate if some biologically meaningful patterns can be identified.

ORA was implemented using the *hypergeometric test* to evaluate whether a particular metabolite set is represented more than expected by chance within the given compound list. One-tailed p values are provided after adjusting for multiple testing. **Figure 2** below summarizes the result.

Metabolite Sets Enrichment Overview

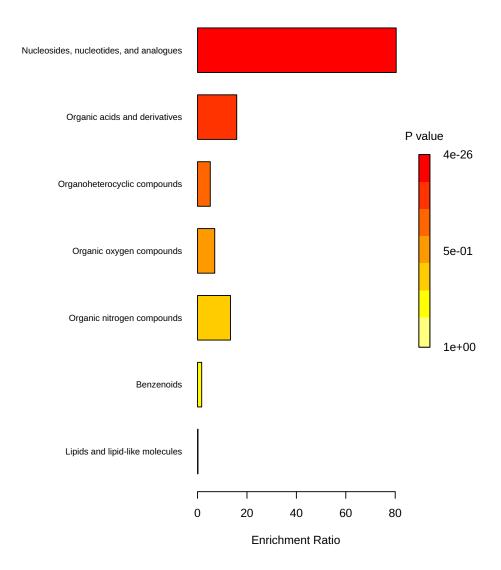


Figure 1: Summary Plot for Over Representation Analysis (ORA) $\,$

Table 2: Result from Over Representation Analysis

	total	expected	hits	Raw p	Holm p	FDR
Nucleosides, nucleotides, and analogues	560	0.20	16	3.54E-26	1.24E-24	1.24E-24
Organic acids and derivatives	4600	1.64	26	2.05E-24	$6.95 ext{E-} 23$	3.58E-23
Organoheterocyclic compounds	6560	2.33	12	3.38E-06	1.12E-04	3.95E-05
Organic oxygen compounds	3240	1.15	8	2.09E-05	6.68E-04	1.83E-04
Organic nitrogen compounds	632	0.23	3	1.54E-03	4.77 E-02	1.08E-02
Benzenoids	5030	1.79	3	2.65E-01	1.00E + 00	1.00E+00
Lipids and lipid-like molecules	94100	33.50	6	1.00E+00	1.00E + 00	1.00E+00

7 Appendix: R Command History

```
[1] "mSet<-InitDataObjects(\"conc\", \"msetora\", FALSE)"
 [2] "cmpd.vec<-c(\"0-Butanoylcarnitine\",\"Glutaryl-carnitine\",\"Arginine-Glutamine\",\"Deoxycarni
 [3] "mSet<-Setup.MapData(mSet, cmpd.vec);"
 [4] "mSet<-CrossReferencing(mSet, \"name\");"
 [5] "mSet<-CreateMappingResultTable(mSet)"
 [6] "mSet<-PerformDetailMatch(mSet, \"Glutaryl-carnitine\");"</pre>
 [7] "mSet<-GetCandidateList(mSet);"
 [8] "mSet<-SetCandidate(mSet, \"Glutaryl-carnitine\", \"Glutarylcarnitine\");"
 [9] "mSet<-PerformDetailMatch(mSet, \"Arginine-Glutamine\");"
[10] "mSet<-GetCandidateList(mSet);"</pre>
[11] "mSet<-SetCandidate(mSet, \"Arginine-Glutamine\", \"Arginylglutamine\");"
[12] "mSet<-PerformDetailMatch(mSet, \"N-carbomoyl-L-aspartate\");"
[13] "mSet<-GetCandidateList(mSet);"
[14] "mSet<-SetCandidate(mSet, \"N-carbomoyl-L-aspartate\", \"Ureidosuccinic acid\");"
[15] "mSet<-PerformDetailMatch(mSet, \"Lysine-Glutamine\");"
[16] "mSet<-GetCandidateList(mSet);"</pre>
[17] "mSet<-SetCandidate(mSet, \"Lysine-Glutamine\", \"Lysylglutamine\");"
[18] "mSet<-PerformDetailMatch(mSet, \"Arginine-Valine\");"
[19] "mSet<-GetCandidateList(mSet);"</pre>
[20] "mSet<-SetCandidate(mSet, \"Arginine-Valine\", \"Arginylvaline\");"
[21] "mSet<-PerformDetailMatch(mSet, \"5- Methylthioadenosine\");"
[22] "mSet<-GetCandidateList(mSet);"</pre>
[23] "mSet<-SetCandidate(mSet, \"5- Methylthioadenosine\", \"5'-Methylthioadenosine\");"
[24] "mSet<-PerformDetailMatch(mSet, \"1-Methylimidazole acetic acid\");"
[25] "mSet<-GetCandidateList(mSet);"</pre>
[26] "mSet<-SetCandidate(mSet, \"1-Methylimidazole acetic acid\", \"Methylimidazoleacetic acid\");"
[27] "mSet<-PerformDetailMatch(mSet, \"Cresol\");"</pre>
[28] "mSet<-GetCandidateList(mSet);"</pre>
[29] "mSet<-SetCandidate(mSet, \"Cresol\", \"p-Cresol\");"</pre>
[30] "mSet<-PerformDetailMatch(mSet, \"Lumichrome\");"
[31] "mSet<-GetCandidateList(mSet);"</pre>
[32] "mSet<-SetMetabolomeFilter(mSet, F);"
[33] "mSet<-SetCurrentMsetLib(mSet, \"main_class\", 2);"
[34] "mSet<-CalculateHyperScore(mSet)"
[35] "mSet<-PlotORA(mSet, \"ora_0_\", \"net\", \"png\", 72, width=NA)"
[36] "mSet<-PlotEnrichDotPlot(mSet, \"ora\", \"ora_dot_0_\", \"png\", 72, width=NA)"
[37] "mSet<-PlotEnrichPieChart(mSet, \"ora\", \"ora_pie_0_\", \"png\", 72)"
[38] "mSet<-CalculateHyperScore(mSet)"
[39] "mSet<-PlotORA(mSet, \"ora_1_\", \"net\", \"png\", 72, width=NA)"
[40] "mSet<-PlotEnrichDotPlot(mSet, \"ora\", \"ora_dot_1_\", \"png\", 72, width=NA)"
[41] "mSet<-PlotEnrichPieChart(mSet, \"ora\", \"ora_pie_1_\", \"png\", 72)"
[42] "mSet<-CalculateHyperScore(mSet)"
[43] "mSet<-PlotORA(mSet, \"ora_2_\", \"net\", \"png\", 72, width=NA)"
[44] "mSet<-PlotEnrichDotPlot(mSet, \"ora\", \"ora_dot_2_\", \"png\", 72, width=NA)"
[45] "mSet<-PlotEnrichPieChart(mSet, \"ora\", \"ora_pie_2_\", \"png\", 72)"
[46] "mSet<-CalculateHyperScore(mSet)"
[47] "mSet<-PlotORA(mSet, \"ora_3\\", \"net\\", \"png\\", 72, width=NA)\"
[48] "mSet<-PlotEnrichDotPlot(mSet, \"ora\", \"ora_dot_3_\", \"png\", 72, width=NA)"
[49] "mSet<-PlotEnrichPieChart(mSet, \"ora\", \"ora_pie_3_\", \"png\", 72)"
[50] "mSet<-CalculateHyperScore(mSet)"
[51] "mSet<-PlotORA(mSet, \"ora_4_\", \"net\", \"png\", 72, width=NA)"
[52] "mSet<-PlotEnrichDotPlot(mSet, \"ora\", \"ora_dot_4_\", \"png\", 72, width=NA)"
[53] "mSet<-PlotEnrichPieChart(mSet, \"ora\", \"ora_pie_4_\", \"png\", 72)"
[54] "mSet<-CalculateHyperScore(mSet)"
[55] "mSet<-PlotORA(mSet, \"ora_5_\", \"net\", \"png\", 72, width=NA)"
[56] "mSet<-PlotEnrichDotPlot(mSet, \"ora\", \"ora_dot_5_\", \"png\", 72, width=NA)"
```

```
[57] "mSet<-PlotEnrichPieChart(mSet, \"ora\", \"ora_pie_5_\", \"png\", 72)"
[58] "mSet<-CalculateHyperScore(mSet)"
[59] "mSet<-PlotORA(mSet, \"ora_6_\", \"net\", \"png\", 72, width=NA)"
[60] "mSet<-PlotEnrichDotPlot(mSet, \"ora\", \"ora_dot_6_\", \"png\", 72, width=NA)"
[61] "mSet<-PlotEnrichPieChart(mSet, \"ora\", \"ora_pie_6_\", \"png\", 72)"
[62] "mSet<-SaveTransformedData(mSet)"
[63] "mSet<-PreparePDFReport(mSet, \"guest10196768329384931757\")\n"
[64] "mSet<-SetMetabolomeFilter(mSet, F);"
[65] "mSet<-SetCurrentMsetLib(mSet, \"kegg_pathway\", 2);"
[66] "mSet<-CalculateHyperScore(mSet)"
[67] "mSet<-PlotORA(mSet, \"ora_7_\", \"net\", \"png\", 72, width=NA)"
[68] "mSet<-PlotEnrichDotPlot(mSet, \"ora\", \"ora_dot_7_\", \"png\", 72, width=NA)"
[69] "mSet<-CalculateHyperScore(mSet)"
[70] "mSet<-PlotORA(mSet, \"ora_8_\", \"net\", \"png\", 72, width=NA)"
[71] "mSet<-PlotEnrichDotPlot(mSet, \"ora\", \"ora_dot_8_\", \"png\", 72, width=NA)"
[72] "mSet<-SaveTransformedData(mSet)"
[73] "mSet<-PreparePDFReport(mSet, \"guest10196768329384931757\")\n"
[74] "mSet<-SetMetabolomeFilter(mSet, F);"
[75] "mSet<-SetCurrentMsetLib(mSet, \"super_class\", 2);"
[76] "mSet<-CalculateHyperScore(mSet)"
[77] "mSet<-PlotORA(mSet, \"ora_9_\", \"net\", \"png\", 72, width=NA)"
[78] "mSet<-PlotEnrichDotPlot(mSet, \"ora\", \"ora_dot_9_\", \"png\", 72, width=NA)"
[79] "mSet<-PlotEnrichPieChart(mSet, \"ora\", \"ora_pie_7_\", \"png\", 72)"
[80] "mSet<-CalculateHyperScore(mSet)"
[81] "mSet<-PlotORA(mSet, \"ora_10_\", \"net\", \"png\", 72, width=NA)"
[82] "mSet<-PlotEnrichDotPlot(mSet, \"ora\", \"ora_dot_10_\", \"png\", 72, width=NA)"
[83] "mSet<-PlotEnrichPieChart(mSet, \"ora\", \"ora_pie_8_\", \"png\", 72)"
[84] "mSet<-SaveTransformedData(mSet)"
[85] "mSet<-PreparePDFReport(mSet, \"guest10196768329384931757\")\n"
```

The report was generated on Mon May 20 $16:45:56\:2024$ with R version 4.3.2 (2023-10-31), OS system: Linux, version: -Ubuntu SMP Tue Mar 5 20:16:58 UTC 2024.