OmicsNet Tutorial: Network building



Computer Requirement

- Modern browser supporting **WebGL**
- Chrome 50+, Firefox 47+, Safari 10.1+ and Edge 12+
- Please make sure WebGL is enabled in your browser
 - Please consult this web page to verify: <u>https://get.webgl.org/</u>
- If not enabled, please consult our FAQ page for instructions
- For best performance and visualization, use:
- Latest version of Google Chrome
- A modern computer with at least 4GB of physical RAM
- A 15-inch screen or bigger (larger is better)
- Retina Display is supported

Motivation

- Explore the different interacting partners of lists of molecules.
- Approach:
 - Single input list: search for interacting partners of seed molecules and expand to other omics types
 - Multiple input lists: build interaction networks from each individual input list and merge them to form composite network
- This tutorial will showcase examples of different network building processes.

Network building for single list

- If gene/protein list is uploaded, the seed gene/proteins can be used to query one or multiple interaction networks based on interaction type and database selected.
- The interaction networks will be merged into connected multi-omics subnetworks



Network building for single list (2)

- If features from omics type other than gene/protein are uploaded, network building is an iterative process where primary interaction network is used to expand the composite network
- Primary interaction network is composed of seeds and its immediate interacting partners.
- Secondary interactions will query for interactions against molecules contained in the primary network, not only seeds.
 - Add edges : PPI as secondary or tertiary interaction will add edges to existing gene/proteins in the network by default.
 - Add nodes and edges: All other interaction types.



Network building for multiple lists

- Each list of molecules are used to build individual omics interaction network
- The individual omics networks are merged to form composite network through shared nodes.



Data upload



☆ Home ☆ Overview ひ Tutorials ⑦ FAQs ☑ Gallery ≡ Updates ☑ Contact

Multi-omics Integration via Biological Networks



- Added support to allow users to browse, search and manually remove edges or nodes from individual omics networks (04/01/2022);
- Added support for adding only edges when using PPI databases (03/25/2022); NEW
- Added R command history for more reproducible analysis (03/17/2022); NEW
- Added support for tissue-based filtering to improve network accuracy (03/15/2022); NEW.
- Fixed a bug for network file upload based on user feedback (03/01/2022); NEW
- Improved interface for network building (02/15/2022); NEW.
- · Updated tutorials for MS peak and taxon list inputs (01/21/2022);
- Improved SNP-gene mapping functions for better performance (01/18/2022);

Data Upload(2)



Case 1: Build PPI network from list of gene/proteins



This procedure identifies which proteins directly interact with seed proteins from the molecular interaction database



« Previous

Proceed >>

Browse View

				Use this function to delete edges at once using differen options on selected colu	multiple t filtering umns	✓ Navigate to:	
Individual Network Interaction Table							
You can browse, search or manual delete an interaction (edge), or use the Advanced Filter to exclude a node (and its all associated edges).							
iat f#	ld2 ↑↓	Name1 ↑↓	Name2	tt.	Astion		
					Action		
100289462	4316	DEFB4A	MMP7		Delete		
10068	1051	IL188P	CEBPB		Delete	Delete individual row by clicking	
10068	27178	IL18BP	IL37		Delete	on those buttons	
10068	3606	IL18BP	IL18		Delete	on these buttons	
101060478	1471	RNF115	CST3		Delete		
117156	8685	SCGB3A2	MARCO		Delete		
1230	6346	CCR1	CCL1		Delete		
1233	6367	CCR4	CCL22		Delete		
1236	6363	CCR7	CCL19		Delete		
1236	6367	CCR7	CCL22		Delete		
1237	6346	CCR8	CCL1		Delete		
1462	1404	VCAN	HAPLN1		Delete		
1462	2199	VCAN	FBLN2		Delete		
1462	2200	VCAN	FBN1		Delete		
1462	2833	VCAN	CXCR3		Delete		
	(1 of 33) << < 1 2 3 4 5 6 7 8 9 10 > >> 15 <						
	« Previous			Proceed »			

Click on proceed so the changes take effect

Node coloring and size

Network Visualization

By default, seed nodes are highlighted by blue "halo" effect around the nodes.



Module-based force-directed layout

Regulation explorer identifies gene regulators



Case 2: Build regulatory network



This procedure identifies which enzymes are interacting the metabolites and the PPI of these enzymes.

Network Building



Network Visualization



Case 3: Build metabolite-enzyme network



*PPI as secondary interaction **only adds connections** between existing protein/gene nodes.

This procedure identifies which enzymes are interacting the metabolites and the PPI of these enzymes.

PPI as secondary interaction

- By default, this process does not add new nodes targeting existing proteins in the network.
- Adding new nodes will dilute the focus on the molecules of interest.
- It will query for interactions between the existing protein nodes in the network, enriching the network with PPI information.

Network Building

							V Navigate to:	
Input list(s) ?	Database Selection Databases are organized	Set PPI as second molecular interact	ary tion creation based on your analysis o	bjectives. Multiple types of net	works will be merged	l (based on shared node	es) 1. dataset<-Init.Data() 2. dataset<-PrepareInputList(dataSet,"Yo	
	Protein-protein Metabolite-protein InnateDB Manu STRING Comp IntAct Manu HuRI Refer	Protein-protein Metabolite-protein InnateDB Manually curated comprehensive PPI (numary mouse) (updated on 01/04/2022) STRING Comprehensive PPI containing both known and predicted PPI (updated on 01/04/2022) (parameters) IntAct Manually curated experimentally validated PPI (updated on 01/04/2022) HuRI Reference interactome map of human binary protein interactions (updated on 01/04/2022)				Add edges only Do not introduce new nodes. Only identify connections within current nodes. 7. CreateGraph() 8. SetPpiZero(FALSE) 9. dataSet<-QueryMet(dataSet, "gene", mate", "met") 10. CreateGraph() 11. SetPpiZero(TRUE) 12. dataSet<-QueryMet(dataSet, "gene", mate", "met") 13. CreateGraph() 14. SetPpiZero(TRUE) 15. dataSet<-QueryMet(dataSet, "gene", mate", "met") 16. CreateGraph() 17. CreateGraph() 18. SetPpiZero(TRUE) 19. dataSet 19. dataSet 10. CreateGraph() 10. CreateGraph() 11. SetPpiZero(TRUE) 12. dataSet 13. CreateGraph() 14. SetPpiZero(TRUE) 15. dataSet 16. CreateGraph() 17. CreateGraph() 18. SetPpiZero(TRUE) 19. CreateGraph() 19. CreateGraph() 19. CreateGraph() 10. CreateGraph() 10. CreateGraph() 11. SetPpiZero(TRUE) 12. dataSet 12. dataSet 13. CreateGraph() 14. CreateGraph() 15. CreateGraph() 16. CreateGraph() 17. CreateGraph() 18. SetPpiZero(TRUE) 19. CreateGraph() 19. CreateGraph() 19. CreateGraph() 10. CreateGraph() 10. CreateGraph() 10. CreateGraph() 11. SetPpiZero(TRUE) 12. CreateGraph() 13. CreateGraph() 14. CreateGraph() 15. CreateGraph() 16. CreateGraph() 17. CreateGraph() 18. CreateGraph() 19. CreateGraph() <		
	Individual Omics Networks Each network is created independe	 DeleteIndWet("gene_gene") SetPpiZero(TRUE) dataset<-QueryWet(dataset, "gene", "innate", "met") CreateGraph() 						
	Input Type	Network Type	Sizes (node# - edge# - seed#)	Browse	Download	Delete		
	Metabolite	Metabolite-protein	186 - 189 - 38	Œ	ٹ	⊡		
	Gene	PPI	9 - 7 - 0	⊞	<u>ب</u>	0		
		« (<u>1</u> > »						

Proceed »

« Previous

Network Visualization

Path explorer enables quick identification of the shortest path connecting two nodes within the subnetwork



Spherical layout

Case 4: Build transcriptional regulatory network



*PPI as secondary interaction only adds connections between existing protein/gene nodes.

This procedure identifies which genes are targeted by the TF and the PPI of these genes.

Change background color and select layout options here Styling options to customize nodes and edges appearance

Network Visualization

Edge occlusion is an issue that particularly affects negatively network visualization, decrease edge opacity or/and apply edge bundling to alleviate the issue.



Case 5: Integrate SNPs in PPI network



*PPI as secondary interaction only adds connections between existing protein/gene nodes.

This procedure identifies which genes associated with uploaded SNPs and link their protein products through PPI.

Network building



Network building (2)

Navigate to: Network Tools 🕐 Multi-omics Network Building R Command History لي Save If more than one network was generated in the previous page, they will be merged together to form multi-omics network through shared nodes. The network is then decomposed into connected 1. dataSet<-Init.Data() Degree Filter subnetworks available for visual analysis in the next page. If the resulting subnetwork1 is too large, you can trim the network to be suitable for visual analysis (< 2000 nodes) using the Network Tools on dataSet<-PrepareInputList(dataSet,"You the left. r input list", "hsa", "snp", "rsid"); Betweenness Filter dataSet<-QueryNet(dataSet, "snp", "Phe noScanner", "snp") CreateGraph() Subnetworks Sizes (node# - edge# - seed#) Download Topology Minimum Network SetPpiZero(FALSE) subnetwork1 602 - 715 - 5 dataSet<-QueryNet(dataSet, "gene", "in View 🗄 Download nate", "snp") CreateGraph() 1 > >> << < CreateGraph() **Tissue Filter** CreateGraph() P-value Filter Zero-order Network Reset to First-order



« Previous

Network Visualization



Case 6: Build transcriptional regulatory network from list of TFs and genes



The composite network is build by merging the individual omics network through shared nodes (gene/protein node in this case)

Network Building

							✓ Navigate to:
Input list(s) ? Gene (51) Transcription factor (6)	Database Selection Databases are organized under d and customized in the next page	Database Selection Databases are organized under different tabs. Please choose proper database(s) for network creation based on your analysis objectives. Multiple types of networks will be merged (based on shared nodes) and customized in the next page					
	TRRUST <u>ENCODE</u> JASPAR	● TRRUST TF-gene interactions constructed using text mining, followed by manual curation (updated on 01/04/2022) ● ENCODE TF-gene interactions derived from ENCODE CHIP-seq data (updated on 01/04/2022) ● JASPAR TF-gene interactions derived from transcription factor binding profiles (updated on 01/04/2022) ● Submit					
	Individual Omics Networks Each network is created independ	ently by searching input list against a selected	d database. The network usually contains several disconne	cted subnetworks.			
	Input Type	Network Type	Sizes (node# - edge# - seed#)	Browse	Download	Delete	
	Gene	PPI	449 - 489 - 46	⊞	⊻	۵	
	Transcription factor	TF-gene	487 - 501 - 8	B	<u>ب</u>	D	



Proceed »

« < 1 > »

Network Visualization

By default, nodes are colored by their omics types and seed nodes are highlighted in blue halos



Color nodes by expression values

Coloring by expression values facilitate the visualization of expression pattern of seed nodes.



Trimming network

Use Network Tools to trim current network to reduce network size and focus on more relevant nodes.

<u> </u>								
Network Tools ? Degree Filter Betweenness Filter) Multi-omics Network Building If more than one network was generated in the previous page, they will be merged together to form multi-omics network through shared nodes. The network is then decomposed into connected subnetworks available for visual analysis in the next page. If the resulting subnetwork1 is too large, you can trim the network to be suitable for visual analytics (< 2000 nodes) using the Network Tools on the left.							
Minimum Network	Subnetworks	Sizes (node# - edge# - seed#)	Тороlоду	Download	<pre>ur input list", "hsa", "tf", "symbo l");</pre>			
Steiner Forest (PCSF)	subnetwork1	75 - 110 - 45	View	🕹 Download	 dataSet<-QueryNetMulti(dataSet, "gen e", "innate", "gene") 			
Tissue Filter	 CreateGraph() dataset<-QueryNetMulti(dataset, "tf", "trrust", "tf") 							
P-value Filter	Minimum network tr	to identify a			 CreateGraph() CreateGraph() 			
Zero-order Network Reset to First-order	minimal network con nodes by computing	 dataset<-PrepareNetwork(dataset, "sub network1", "omicsnet_0.json") dataset<-BuildMinConnectedGraphs(data Set) 						
	between no	odes.						





Network visualization

Notice that the network is composed mostly of seed nodes and has a minimum amount of interacting nodes that are necessary to form subnetwork



Enrichment analysis uncovers involved pathways

Selecting enriched pathways will highlight corresponding genes in the network.



Concentric circle layout reveals the 2. Select concentric connection patterns of a focal node to Concentric circle layout circle layout option here the rest of the network. The further a node is to the focal node, the greater number of steps it takes to reach it. Network: subnetwork1 V Background: Purple-gradient V Layout: Concentric circle ✓ Styling: Label ▼ Download: -- Specify -- ▼ More Options ✓ Scope: -- Specify --Global Node Styles Fuction Explorer Query: All nodes v Color Туре Size Э Database: KEGG (gene) ✓ Submit Save Gene CCR1 4+ BCL6 Name Col CCL24 Tf Ð Phosphatidylinositol 12 8.68e-12 2.92e-9 Sphingolipid signali 14 1e-10 1.69e-8 Q Node Explorer FoxO signaling path 14 3.21e-8 0.00000276 ò. Fc epsilon RI signal 9 3.29e-8 🔍 Search 🔳 Delete 📊 0.00000276 CCL1 IL3RA φ "hermogenesis 0.0000217 0.00146 Expi RNASE1 1. Select a node of interest to MMP7 Insulin resistance 0.000054 0.00303 61.1 investigate by clicking it on the Iuntington disease 0.0000766 0.00338 51.0 CXCR3 CCL7 Colorectal cancer 8 0.0000805 0.00338 23.7 node table or in the network SMAD2 FOXP3 3D TNF signaling pathy 6 0.00016 0.00597 23.7 CCR4 0.000585 0.0197 Necroptosis 188.9769 21.0 6354 CD44_ CCRL1 FPR1 IL3 MMP2 4087 SMAD2 21.0 Module Explorer ELAVL1 AHR FAM46A \checkmark 1358.68 19.7 CXCL10 SP1 17 6663 Algorithm: InfoMap ✓ Submit BRCA1 \square 6369 CCL24 19.7 RELA Color Size P-value Module Ouerv 50943 FOXP3 2 11.45238 16.2 HTRA1 0 18 183 1.25e-05 RETN 4494 16.2 MT1F 0 NFKB1 NOS3 Ă VCAN 4501 MT1X 3 49.67017 14.7 IRF1 IL18BP \square 4316 MMP7 12.0 0 CTSB CCL19 771 CA12 2 41.21334 8 19 6346 CCL1 2 21.40772 7.07 MMP9 CST3 LIPA CCL22 5 97.91905 6367 6.97 FOS <u>_____</u>МТ1Х RELB \square 5473 PPBP 3 53,67333 6.76 PLD3 \square 5.71 2357 FPR1 HIST2H2AC HNF4A PTGES \square 5265 SERPINA 3 79.86675 5.65 🔻 GPNMB |4 4 Page 1 of 3 🕨 🕨 PPBP HSD11B1 🦰 МТ1М Current Selection ALB FoxO signaling pathway SERPIGATE GERSIDHTAT CCR4 CCL22 6.97 PPRP 6.76 CCL7 21.04 IL3RA 4.21 CXCL10 -9.89 CCR1 Path Explorer CCL19 4.08 CCL1 7.07 Batch Selection CCI 24 19.76 [→ Logout

The End