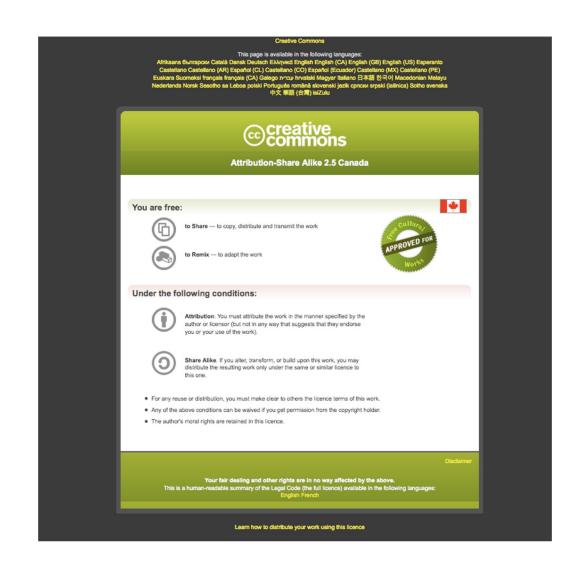


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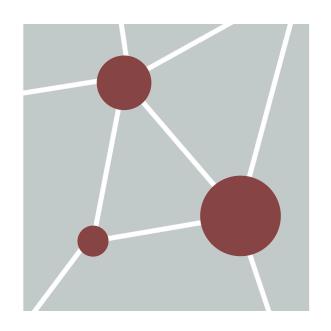
bioinformaticsdotca.github.io



Module 3 Enrichment Maps



Ruth Isserlin Pathway and Network Analysis June 5-7, 2023















Learning Objectives

- By the end of this lecture, you will:
 - Understand...
 - Be able to define...
 - Know...

Learning Objectives

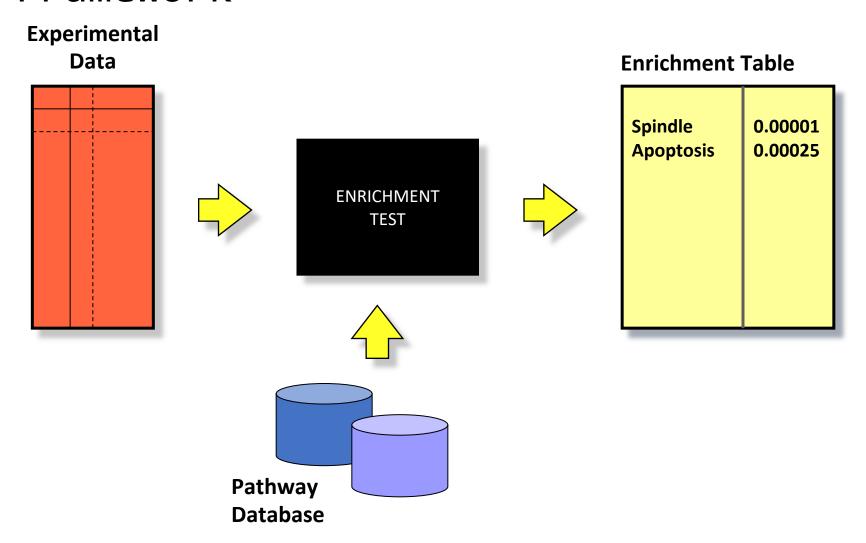
- By the end of this lecture, you will:
 - Understand how to transform your enrichment results from g:profiler, gsea or other enrichment algorithms to a Network
 - Understand the differences between a network and an Enrichment Map
 - Be able to summarize your enrichment results with annotation using the Autoannotate App.

Results of Lab 2 - Enrichment Results

GO.ID	Description	p.Val	FDR	Phenotype	Genes											
GO:0060255	regulation of macromolecule metabolic pro	oces 2.34E-30	2.34E-30		1 TP53,PTEN	,APC,VHL,KRA	S,ARID1A,EG	FR,NF1,PIK3	R1,CDKN2A,G	ATA3,RB1,NC	DTCH					
GO:0051171	regulation of nitrogen compound metaboli	c prc 6.44E-30	6.44E-30		1 TP53,PTEN	,APC,VHL,KRA	S,ARID1A,EG	FR,NF1,PIK3	R1,CDKN2A,G	ATA3,RB1,NC	DTCH					
GO:0031323	regulation of cellular metabolic process	2.80E-29	2.80E-29		1 TP53,PIK3C	A,PTEN,APC,V	HL,KRAS,ARI	D1A,EGFR,N	F1,PIK3R1,CDI	(N2A,GATA3,	RB1					
GO:0019222	regulation of metabolic process	3.50E-29	3.50E-29		1 TP53,PIK3C	A,PTEN,APC,V	HL,KRAS,ARI	D1A,EGFR,N	F1,PIK3R1,CDI	(N2A,GATA3,	RB1					
GO:0080090	regulation of primary metabolic process	4.34E-29	4.34E-29		1 TP53,PTEN	,APC,VHL,KRA	S,ARID1A,EG	FR,NF1,PIK3	R1,CDKN2A,G	ATA3,RB1,NC	OTCH					
GO:0051173	positive regulation of nitrogen compound r	meta 1.35E-28	1.35E-28		1 TP53,PTEN	,APC,VHL,KRA	S,ARID1A,EG	FR,PIK3R1,C	DKN2A,GATA3	,RB1,NOTCH	1,FB					
GO:0010604	positive regulation of macromolecule meta	abol 1.51E-28	1.51E-28		1 TP53,PTEN	,APC,VHL,KRA	S,ARID1A,EG	FR,PIK3R1,C	DKN2A,GATA3	,RB1,NOTCH	1,FB:					
GO:1903506	regulation of nucleic acid-templated transc	cript 5.40E-27	5.40E-27		1 TP53,PTEN	VHL,ARID1A,I	GFR,PIK3R1	,CDKN2A,GA	TA3,RB1,NOT	CH1,FBXW7,C	CTNN					
GO:2001141	regulation of RNA biosynthetic process	5.40E-27	5.40E-27		1 TP53,PTEN	VHL,ARID1A,I	GFR,PIK3R1	,CDKN2A,GA	TA3,RB1,NOT	CH1,FBXW7,C	CTNN					
GO:0009893	positive regulation of metabolic process	NAME							GS fol	o GS DETAILS	SIZE	ES	NES	NOM p-val	FDR q-val	
GO:0097659	nucleic acid-templated transcription	HALLMARK_INTERF	ERON_GAMMA	_RESPONSES	6MSIGDB_C2%	HALLMARK_IN	TERFERON_G/	AMMA_RESPO	NSE HALLMARK	_I Details	19	-0.8124175	-3.1241286	. 0	. (j
GO:0032774	RNA biosynthetic process	HALLMARK_INTERF	ERON_ALPHA_	RESPONSE%N	//SIGDB_C2%H	ALLMARK_INTE	RFERON_ALPI	HA_RESPONSI	HALLMARK	_I Details	9.	-0.871453	-2.969929	0	(J
GO:0019219	regulation of nucleobase-containing com	RESPONSE TO INTE	RFERON-GAMI	MA%GOBP%0	O:0034341				RESPONSE	T Details	11	-0.7410603	-2.8620038	0	(j
GO:0031325	positive regulation of Centular Infetabolic p	DEFENSE RESPONS							DEFENSE R		10		-2.8492434			
GO:0006355	regulation of transcription. DNA-template	NTERFERON SIGNA								N Details	16					
GO:0051252	regulation of RNA metabolic process	HALLMARK_ALLOGI	_				REJECTION			_/ Details	18					
	transcription DNA tomplated	CELLULAR RESPONS NTERFERON GAMI								REDetails ON Details	5					
		ANTIGEN PROCESS					EN VIA MHC CI	ASS I TAD-DI				2 -0.7555878				
	regulation of cellular macromolecule bios							LA33 I, IAF-DI	INTERFERO		6			_		
		TYPE I INTERFERON							TYPE I INTE		5			0		
	F	RESPONSE TO TYPE	INTERFERON	%GOBP%GO	:0034340				RESPONSE	T Details	5-	-0.8517665	-2.6604805	0	(j
	A	ANTIGEN PROCESS	ING AND PRESE	NTATION OF	EXOGENOUS	PEPTIDE ANTIG	EN VIA MHC C	LASS I%GOBP	%GCANTIGEN F	R Details	7	-0.7585589	-2.5932803	0	()
	E	R-PHAGOSOME PA	ATHWAY%REAC	TOME%R-HS	A-1236974.4				ER-PHAGO	S(Details	8	-0.7237594	-2.589577	0	(j
	C	COSTIMULATION B	Y THE CD28 FAN	/IILY%REACT(ME%R-HSA-3	88841.3			COSTIMUL	Al Details	6	-0.780563	-2.5822923	0	(j
	-	PROTEASOME DEG			_		SAPIENS		PROTEASO		6		-2.5736508			
	-	REGULATION OF LE							REGULATIO		14					
	-	ANTIGEN PROCESS					LASS I%GOBP	%GO:0002474		R Details	9.					
	Į.	ANTIGEN RECEPTO	K-MEDIATED SI	GNALING PA	HWAY%GOBP	%GO:0050851			ANTIGEN	E Details	16	-0.6937604	-2.5514965	0	(!

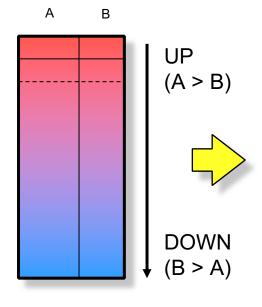
- Over the years thousands of papers have been translated into digital representations as pathways or functional annotation.
- Enrichment analysis allows us to interpret data from these papers
- But Major cognitive burden relating overlapping information

Pathway Enrichment Test: General Framework



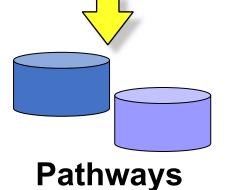
Pathway Enrichment Analysis

Ranked Gene List



Thresholded lists (selected genes)







Enrichment in Condition A vs. B

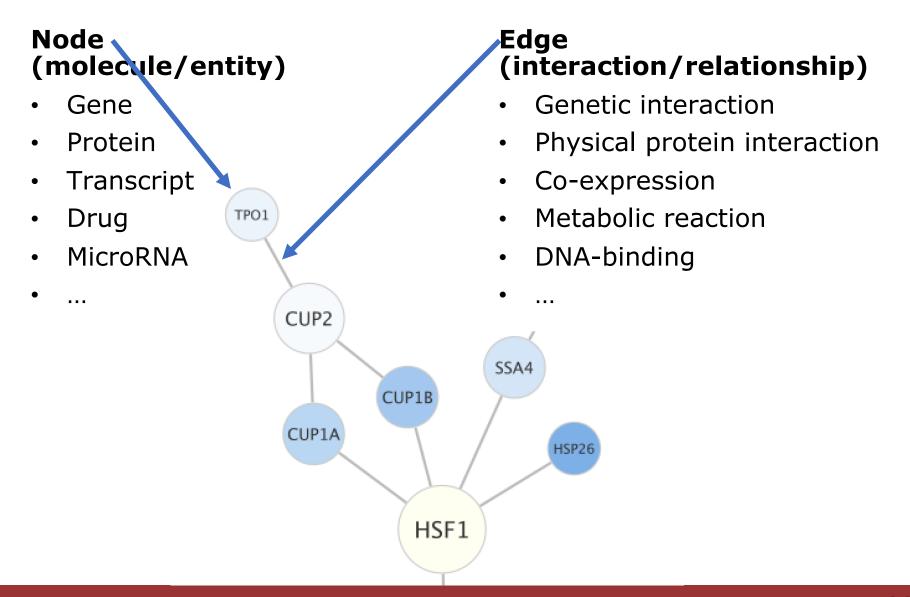
Gene-set	Significance
Cell Cycle EGF Pathway Spindle 	0.0001 0.003 0.007

Enrichment in Condition B vs. A

Gene-set	Significance
Proteasome Apoptosis Caspase 	0.0002 0.005 0.009

Merico D, Isserlin R, Stueker O, Emili A, Bader GD Enrichment map: a network-based method for gene-set enrichment visualization and interpretation PLoS One. 2010 Nov 15;5(11):e13984

Network Basics



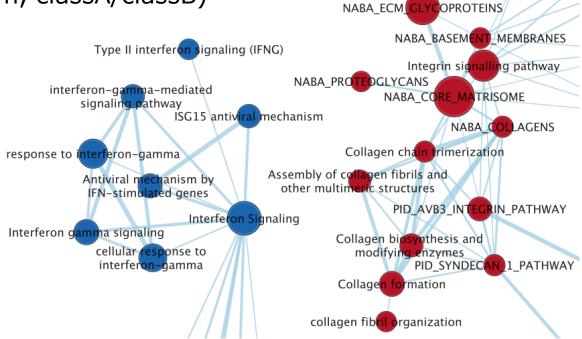
Enrichment Map Basics

Node (molecule/entity)

- Pathway or geneset
- Size is correlated to number of genes in set
- Color indicates class in below example (for example Up/Down, classA/classB)

Edge (interaction/relationship)

- Degree of overlap between two genesets
- The more genes two pathways have in common the thicker/stronger the connection



Enrichment Map

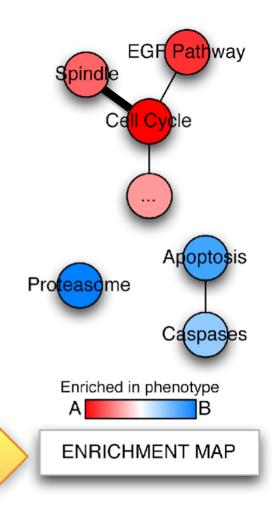
Enrichment in Condition A vs. B

Gene-set	Significance
Cell Cycle EGF Pathway Spindle 	0.0001 0.003 0.007

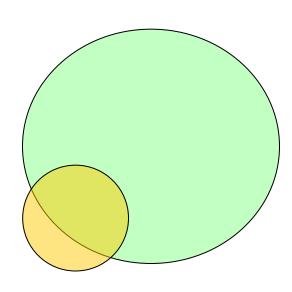
Enrichment in Condition B vs. A

Gene-set	Significance
Proteasome Apoptosis Caspase 	0.0002 0.005 0.009

GENE-SET LIST

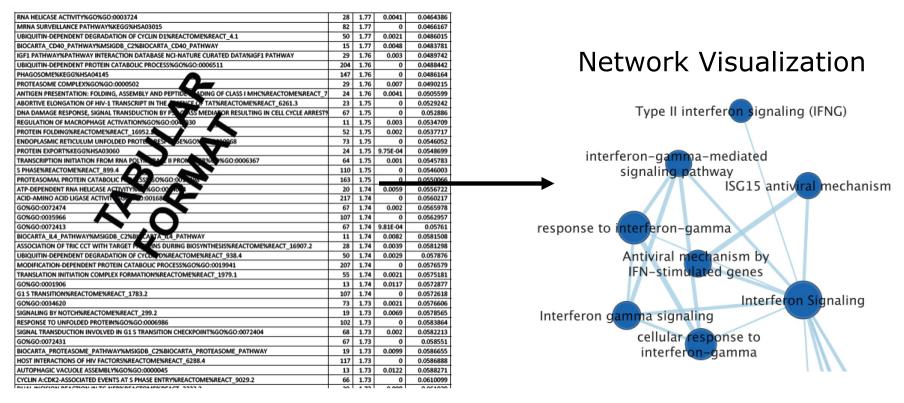


Overlap



$$\frac{|A \cap B|}{\min(|A|,|B|)}$$

Typical Output



Each row is a gene-set (pathway). It displays:

- a score associated with the magnitude of overlap between gene-set and gene list.
- a pvalue that estimates the significance of the enrichment (by chance or not).
- a corrected pvalue (FDR) that corrects for multiple hypothesis testing.

Enrichment Map: use case I Single enrichment

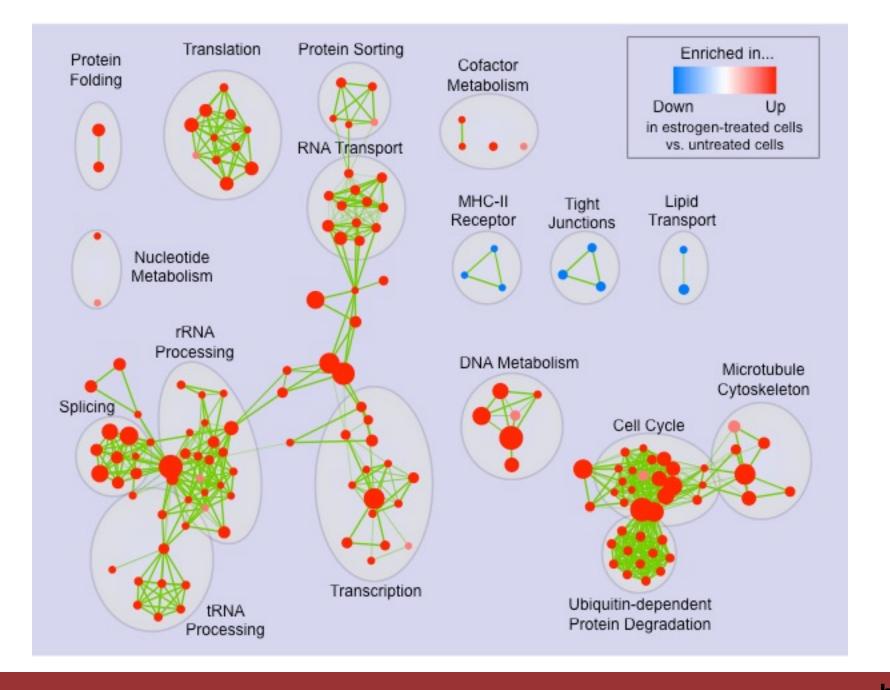
Estrogen treatment of breast cancer cells

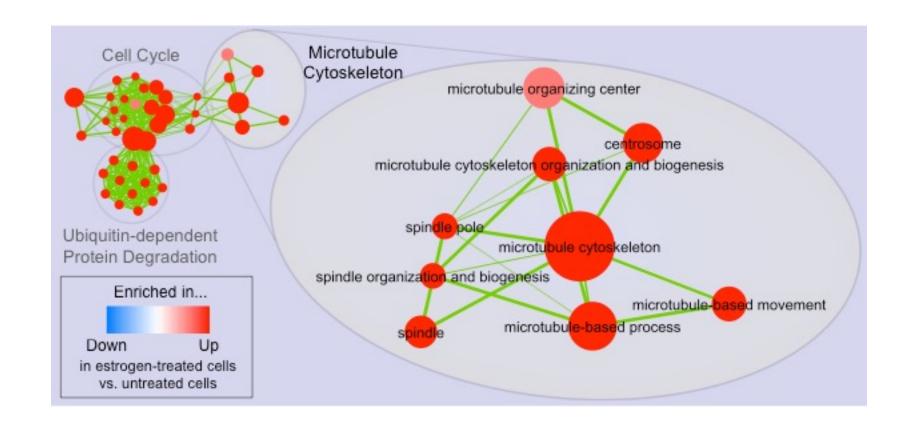
- <u>Design</u>:
 - 2-time points, two-class

	12 hrs	24 hrs
Estrogen-treated	3	3
Untreated	3	3

Gene set Database:

Gene Ontology





Enrichment Map: use case II Comparison of two enrichments

Estrogen treatment of breast cancer cells

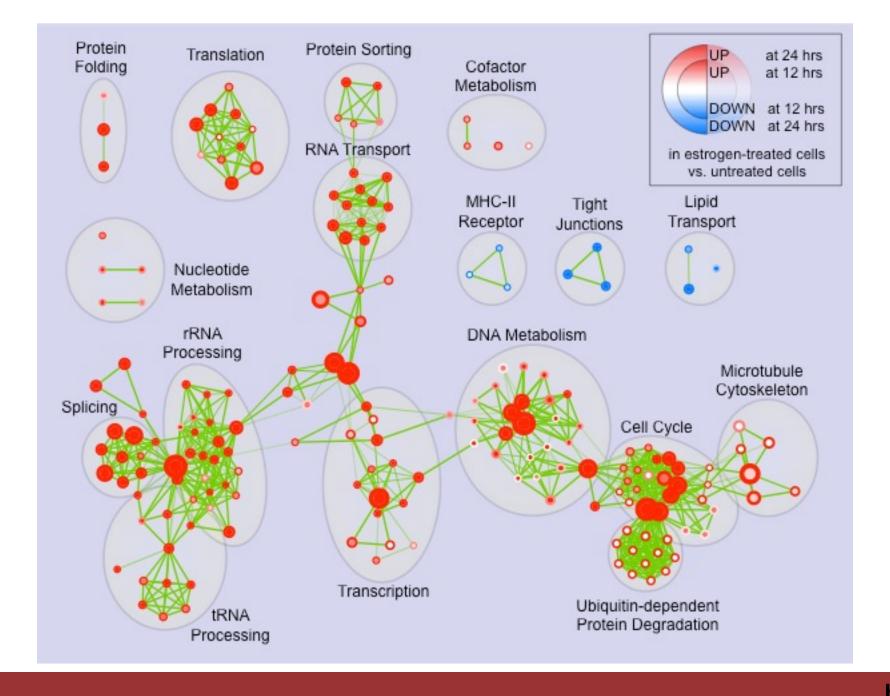
• <u>Design</u>:

2-time points, two-class

	12 hrs	24 hrs
Estrogen-treated	3	3
Untreated	3	3

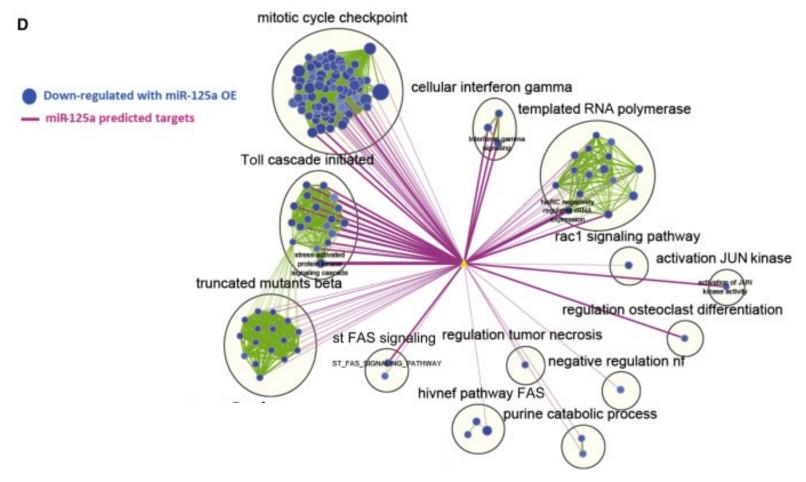
Gene set Database:

Gene Ontology



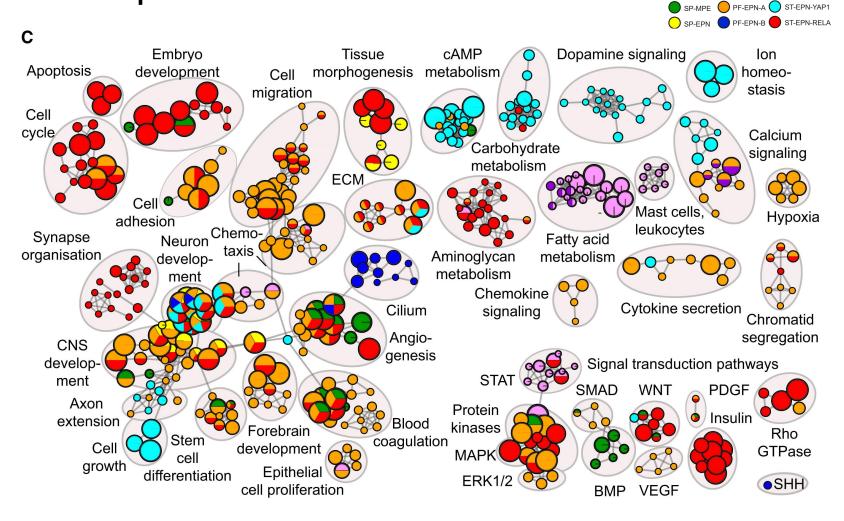


Enrichment Map: use case III Query Set Analysis



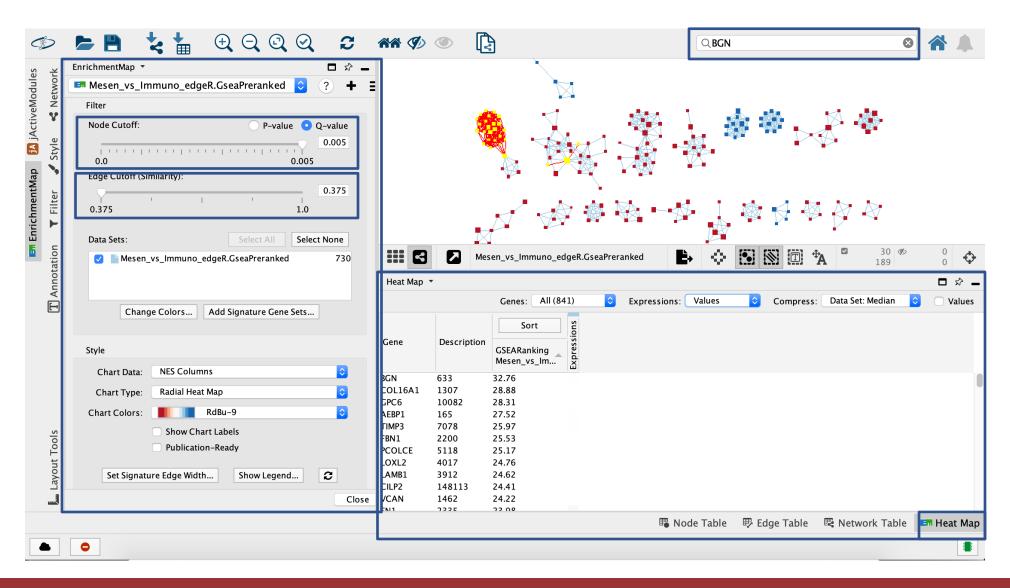
Wojtowicz EE, Lechman ER, Hermans KG, et al. Ectopic miR-125a Expression Induces Long-Term Repopulating Stem Cell Capacity in Mouse and Human Hematopoietic Progenitors. *Cell Stem Cell*. 2016;19(3):383-396

Enrichment Map: use case IV Multiple Datasets

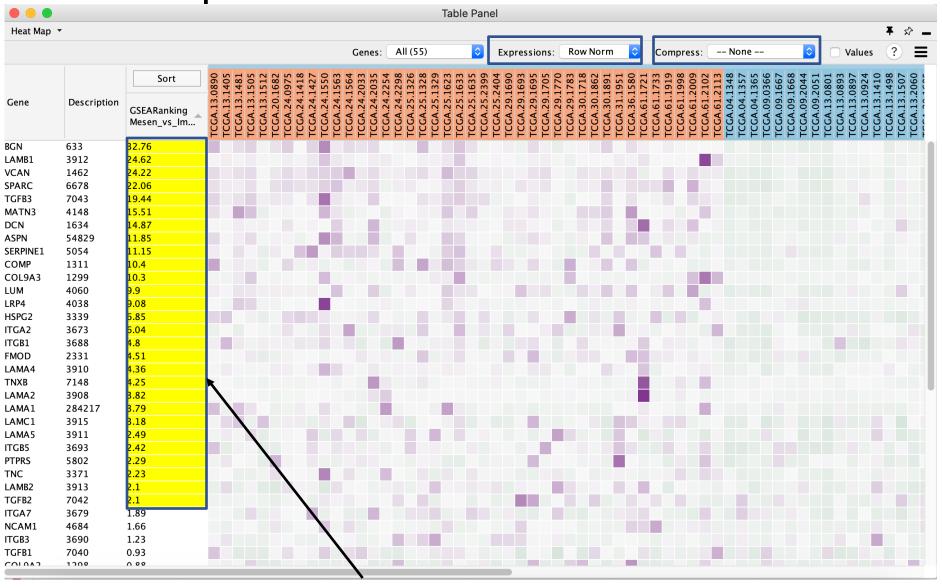


Pajtler KW, Witt H, Sill M, et al. Molecular Classification of Ependymal Tumors across All CNS Compartments, Histopathological Grades, and Age Groups. *Cancer Cell*. 2015;27(5):728-743

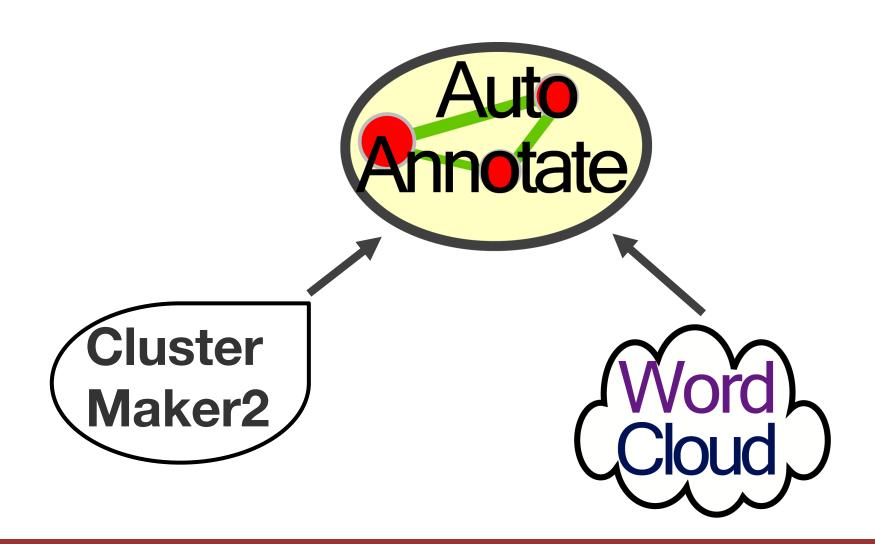
Enrichment Map Features



Heatmap Panel



Enrichment Map Features



AutoAnnotate

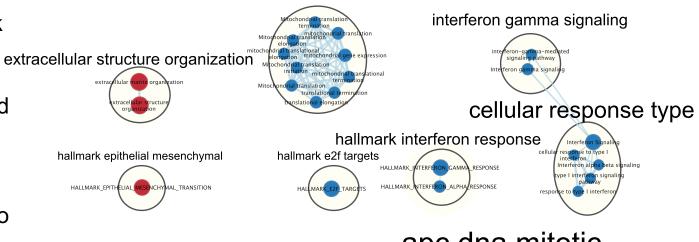
Steps:

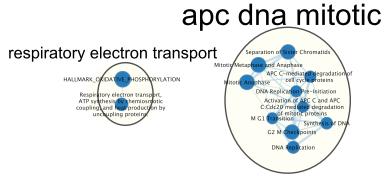
mitochondrial translational translation

1. Cluster the. Network

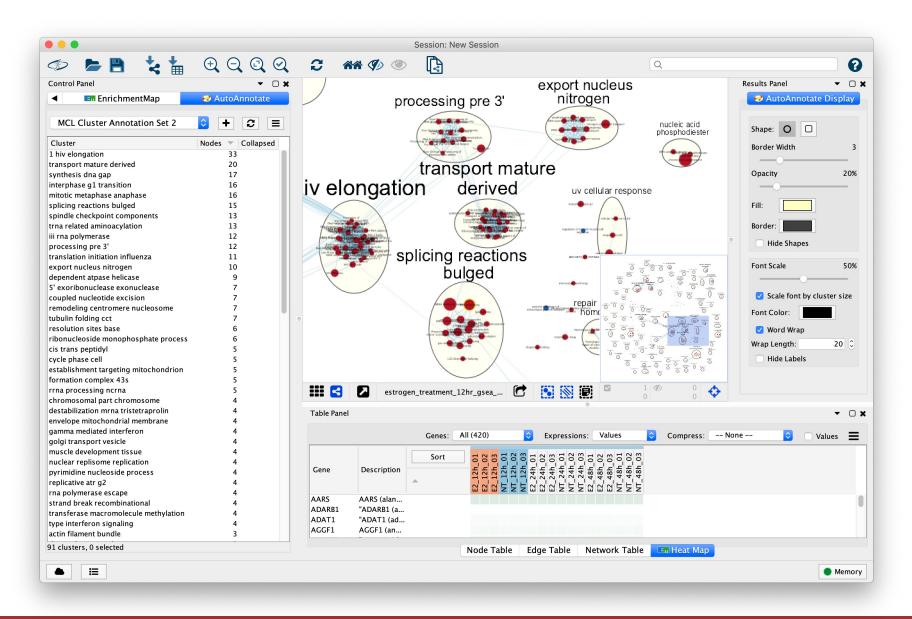
2. For each cluster, find frequent words in node labels

- 3. Select top 3 words to create labels
- 4. Possibility to move or edit labels to make figure publication ready.

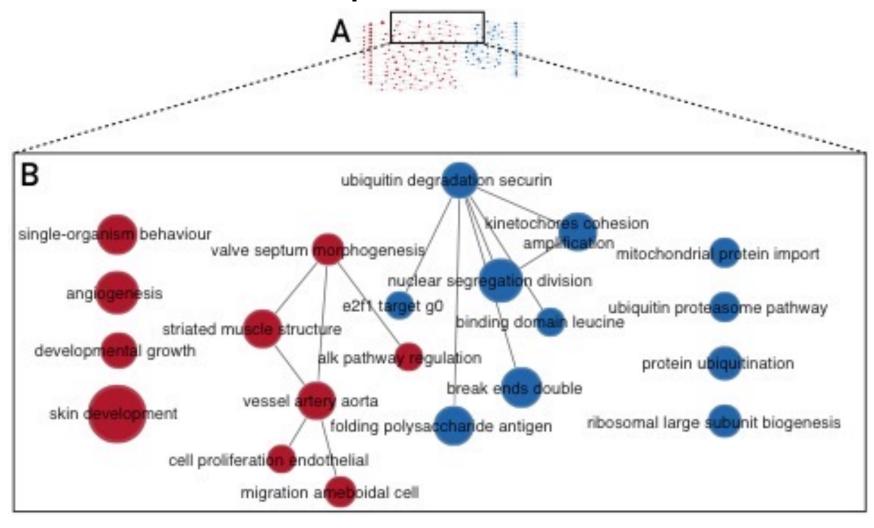




AutoAnnotate



Collapsed network



Practical Lab

- Experiment with Cytoscape
- Try out Enrichment Map
- Load your g:Profiler results and load the GSEA enrichment analysis files that you generated in Module 2 lab
- Try out all the different Enrichment Map features
- Try AutoAnnotate to annotate your networks.

We are on a Coffee Break & Networking Session

Workshop Sponsors:







