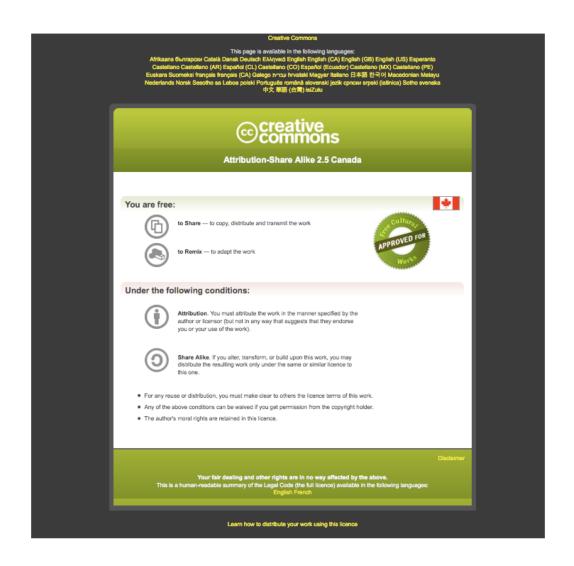


## Canadian Bioinformatics Workshops

www.bioinformatics.ca bioinformaticsdotca.github.io

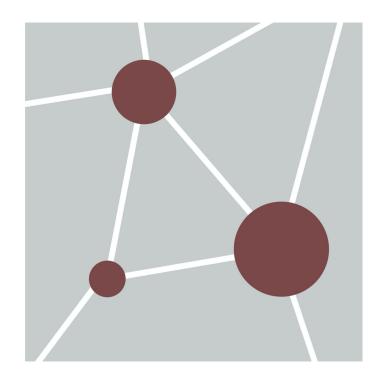




## Module 3 Enrichment Maps



Ruth Isserlin Pathway and Network Analysis of -omics Data
July 27-29, 2020





### Learning Objectives

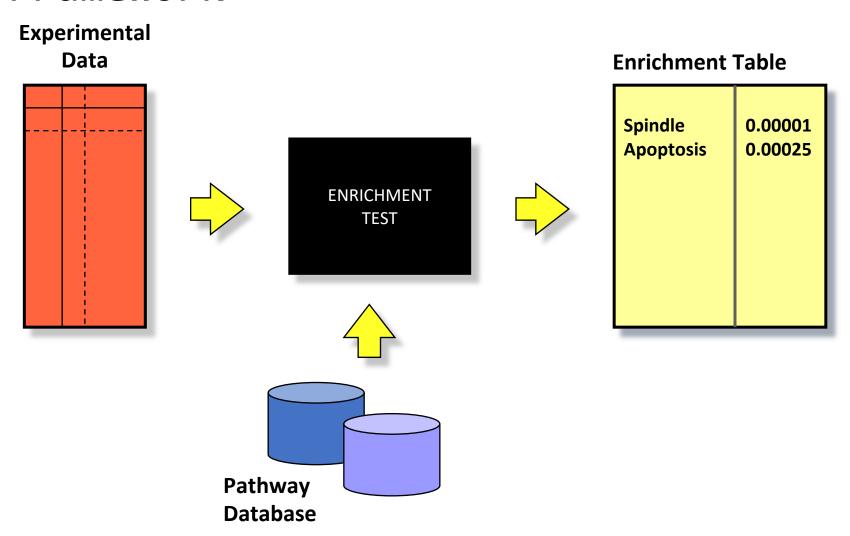
- By the end of this lecture, you will:
  - Understand how to transform your enrichment results from g:profiler, gsea or other enrichment algorithms to a Network
  - Understand the differences between a network and an Enrichment Map
  - Be able to summarize your enrichment results with annotation using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

### Results of Lab 2 - Enrichment Results

GO.ID	Description	p.Val	FDR	Phenotype	Genes						-				
GO:0060255	regulation of macromolecule metabolic proce	es 2.34E-30	2.34E-30	1	TP53,PTEN,	APC,VHL,KRA	S,ARID1A,EGF	R,NF1,PIK3R1	,CDKN2A,GA	TA3,RB1,NOTCH	1				
GO:0051171	regulation of nitrogen compound metabolic p	orc 6.44E-30	6.44E-30	1	TP53,PTEN,	APC,VHL,KRA	S,ARID1A,EGF	R,NF1,PIK3R1	,CDKN2A,GA	TA3,RB1,NOTCH	ĺ				
GO:0031323	regulation of cellular metabolic process	2.80E-29	2.80E-29	1	TP53,PIK3CA	A,PTEN,APC,\	/HL,KRAS,ARID	1A,EGFR,NF1	,PIK3R1,CDKI	N2A,GATA3,RB1					
GO:0019222	regulation of metabolic process	3.50E-29	3.50E-29	1	TP53,PIK3CA	A,PTEN,APC,\	/HL,KRAS,ARID	1A,EGFR,NF1	,PIK3R1,CDKI	N2A,GATA3,RB1					
GO:0080090	regulation of primary metabolic process	4.34E-29	4.34E-29	1	L TP53,PTEN,	APC,VHL,KRA	S,ARID1A,EGF	R,NF1,PIK3R1	,CDKN2A,GA	TA3,RB1,NOTCH	Ī				
GO:0051173	positive regulation of nitrogen compound me	ta 1.35E-28	1.35E-28	1	TP53,PTEN,	APC,VHL,KRA	S,ARID1A,EGF	R,PIK3R1,CDK	N2A,GATA3,	RB1,NOTCH1,FB					
GO:0010604	positive regulation of macromolecule metabo	1.51E-28	1.51E-28	1	L TP53,PTEN,	APC,VHL,KRA	S,ARID1A,EGF	R,PIK3R1,CDK	N2A,GATA3,	RB1,NOTCH1,FB					
GO:1903506	regulation of nucleic acid-templated transcrip	ot 5.40E-27	5.40E-27	1	L TP53,PTEN,	VHL,ARID1A,	EGFR,PIK3R1,0	DKN2A,GATA	3,RB1,NOTC	H1,FBXW7,CTNN	N				
GO:2001141	regulation of RNA biosynthetic process	5.40E-27	5.40E-27	1	L TP53,PTEN,	VHL,ARID1A,	EGFR,PIK3R1,0	DKN2A,GATA	3,RB1,NOTC	H1,FBXW7,CTNN	N				
GO:0009893	positive regulation of metabolic process NAM	ME							GS follo	GS DETAILS SIZE	E	ES	NES	NOM p-val	FDR g-val
GO:0097659	nucleic acid-templated transcription HAL	LMARK_INTERFE	RON_GAMMA	_RESPONSE%	MSIGDB_C2%I	HALLMARK_IN	TERFERON_GAI	MMA_RESPON	SE HALLMARK_	Details	195		-3.1241286	-	
GO:0032774	RNA biosynthetic process HAL	LMARK_INTERFE	RON_ALPHA_R	RESPONSE%N	1SIGDB_C2%HA	LLMARK_INTI	ERFERON_ALPH	A_RESPONSE	HALLMARK_	l Details	94	-0.871453	-2.969929	0	0
GO:0019219	regulation of nucleobase-containing compact	PONSE TO INTER	FERON-GAMM	1A%GOBP%G	O:0034341				RESPONSE T	Details	119	-0.7410603	-2.8620038	0	0
GO:0031325	bositive regulation of centual inetabolic L	ENSE RESPONSE							DEFENSE RE		102		-2.8492434	0	_
GO:0006355	regulation of transcription. DNA-template	ERFERON SIGNA							INTERFERO		169		-2.7944484	0	_
GO:0051252	regulation of RNA metabolic process	LMARK_ALLOGR	_	_			REJECTION		HALLMARK_		181				
	transcription DNA tomplated	LULAR RESPONS							CELLULAR R		50 81		-2.7440908 -2.7424092		
	l l l l c	ERFERON GAMM TIGEN PROCESSIN					EN VIA MUC CL	SCITAD DEDI			81		-2.7424092		-
	regulation of cellular macromolecule bios							ASS I, TAP-DEFE	INTERFEROI		64			0	
		E I INTERFERON					,5		TYPE I INTER		50		-2.6922164	0	-
	RES	PONSE TO TYPE	INTERFERONS	%GOBP%GO:	0034340				RESPONSE T	Details	54		-2.6604805	0	0
	ANT	TIGEN PROCESSIN	IG AND PRESE	NTATION OF	EXOGENOUS P	EPTIDE ANTIG	EN VIA MHC CLA	ASS I%GOBP%0	CANTIGEN PR	Details	77	-0.7585589	-2.5932803	0	0
	ER-	PHAGOSOME PA	THWAY%REACT	TOME%R-HSA	A-1236974.4				ER-PHAGOS	Details	81	-0.7237594	-2.589577	0	0
		STIMULATION BY							COSTIMULA		61				_
		DTEASOME DEGR			_		SAPIENS		PROTEASON		60		-2.5736508		
		GULATION OF LEU							REGULATIO		142				_
		TIGEN PROCESSIN					CLASS I%GOBP%	GO:0002474	ANTIGEN PR		94				
	ANI	TIGEN RECEPTOR	-IVIEDIATED SIG	INALING PAI	HWAY%GUBP9	%GO:0050851			ANTIGEN RE	Details	165	-0.6937604	-2.5514965	0	0

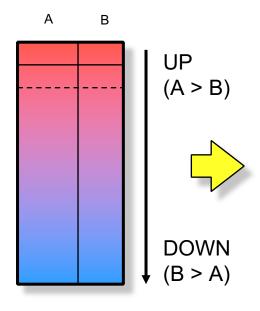
- Over the years thousands of papers have been translated into digital representations as pathways or functional annotation.
- Enrichment analysis allows us to interpret data from these papers
- But .... Major cognitive burden relating overlapping information

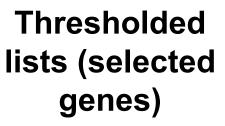
## Pathway Enrichment Test: General Framework



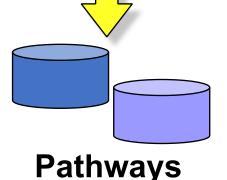
## Pathway Enrichment Analysis

### Ranked Gene List













Enrichment in Condition A vs. B

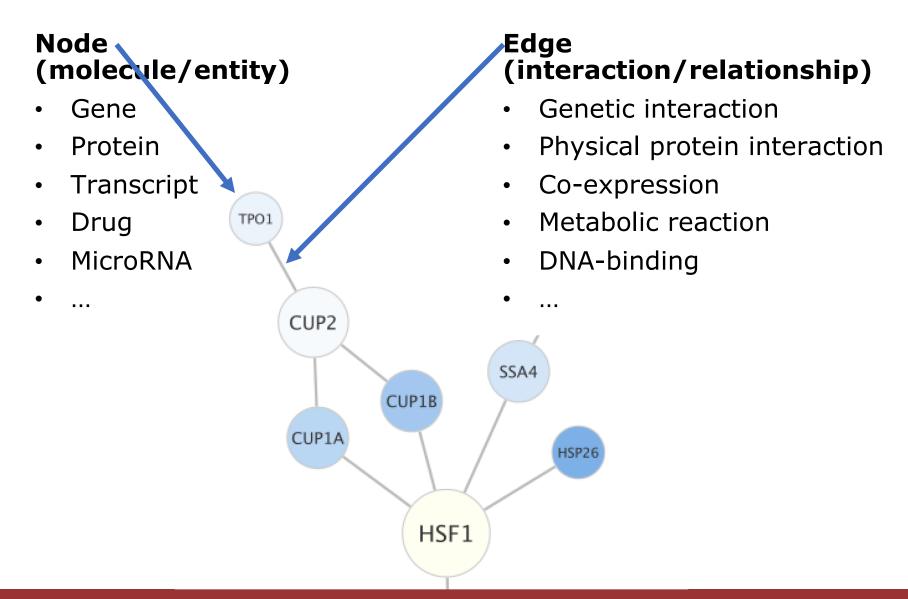
Gene-set	Significance
Cell Cycle EGF Pathway Spindle 	0.0001 0.003 0.007

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



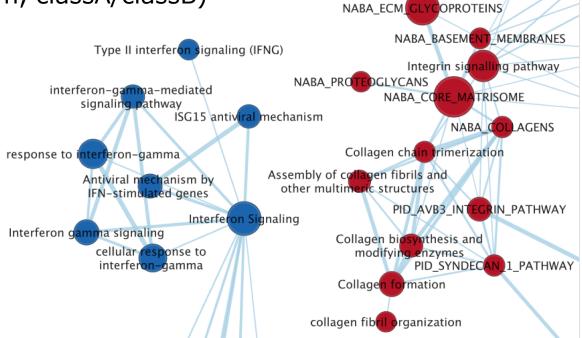
### Enrichment Map Basics

#### 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

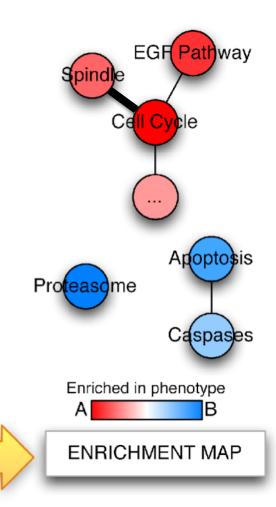
Enrichment in Condition A vs. B

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

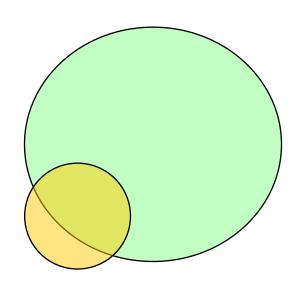
Enrichment in Condition B vs. A

Gene-set	Significance
Proteasome Apoptosis Caspase 	0.0002 0.005 0.009

**GENE-SET LIST** 

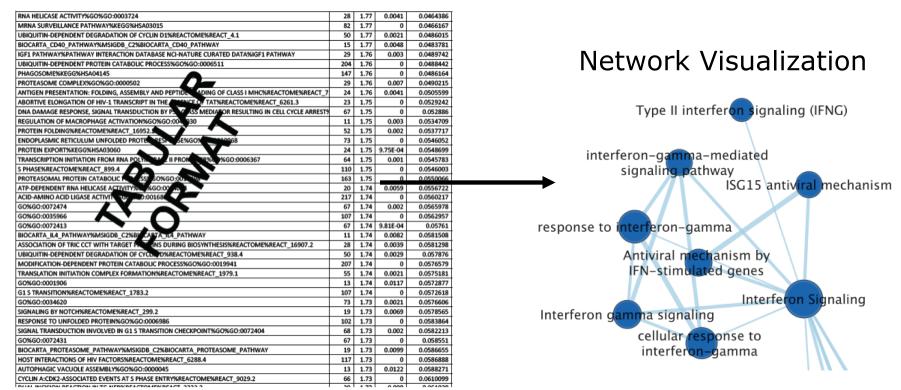


### <u>Overlap</u>



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

## 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

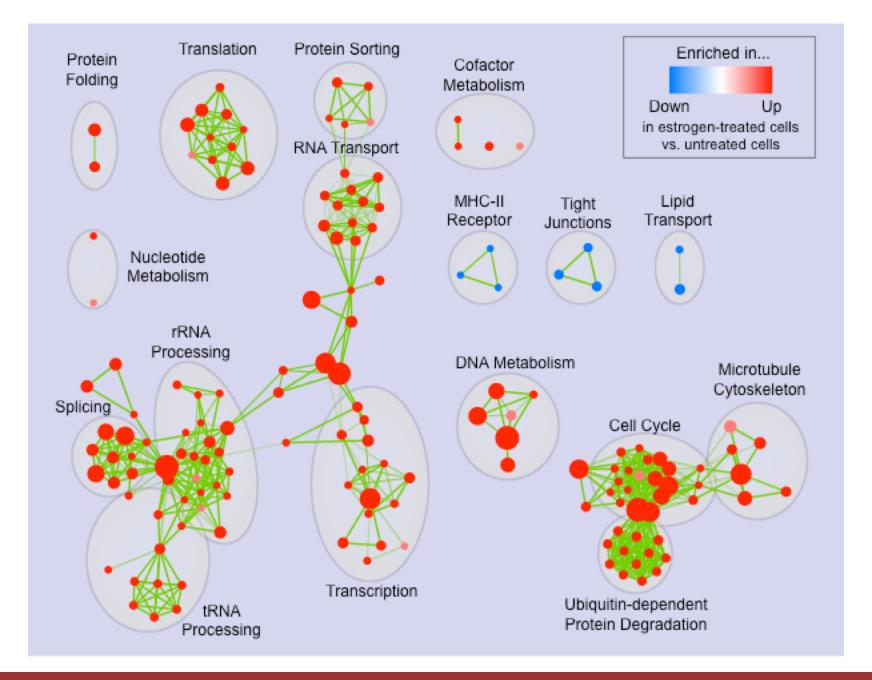
Design:2-time points, two-class

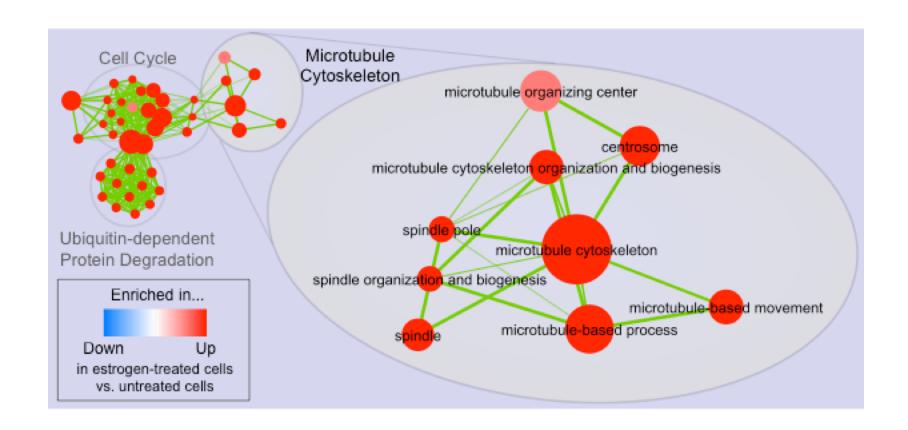
	12 hrs	24 hrs
Estrogen-treated	3	3
Untreated	3	3

• Gene set Database:

**Gene Ontology** 

Lin C-Y, Vega VB, Thomsen JS, Zhang T, Kong SL, et al. (2007) Whole-genome cartography of estrogen receptor alpha binding sites. PLoS Genetics 3:e87





# Enrichment Map: use case II Comparison of two enrichments

Estrogen treatment of breast cancer cells

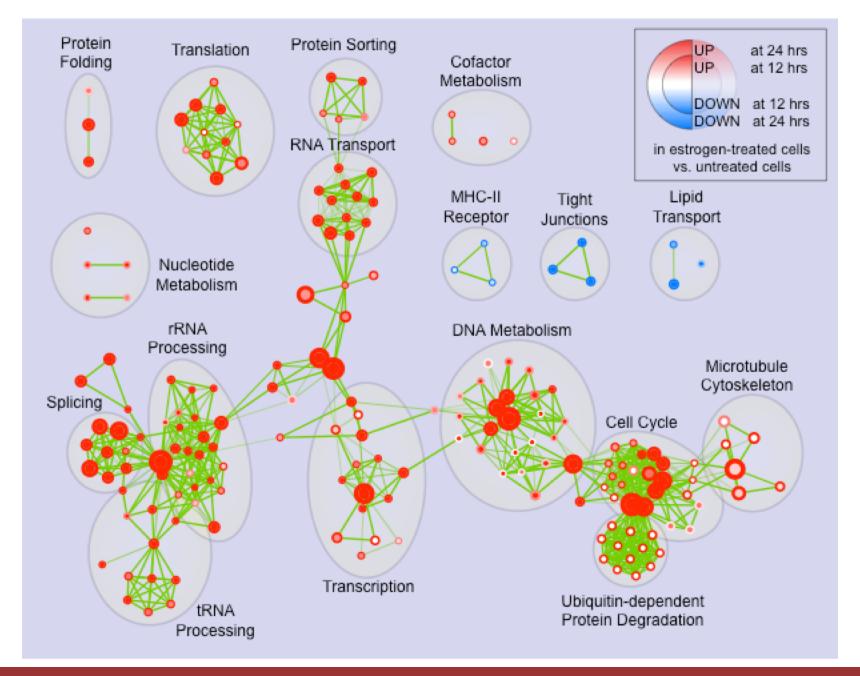
Design:

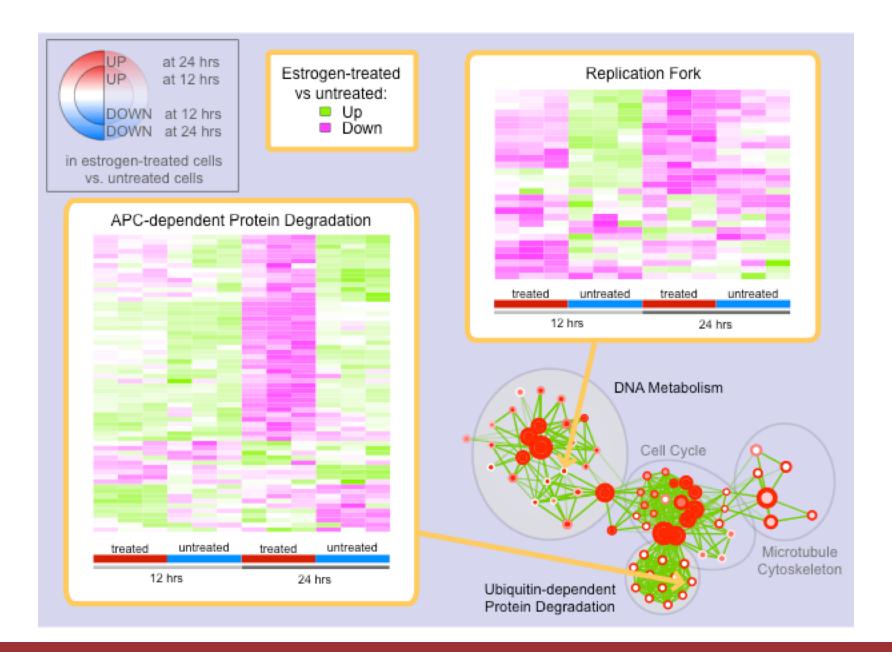
2-time points, two-class

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

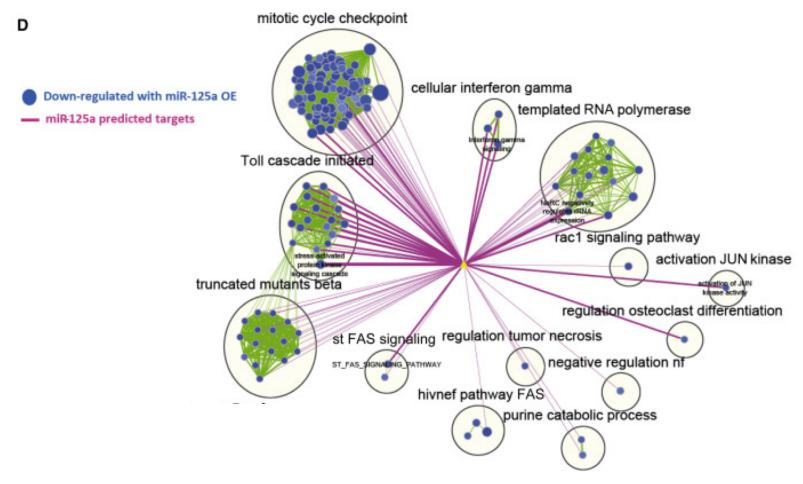
Gene set Database:

**Gene Ontology** 



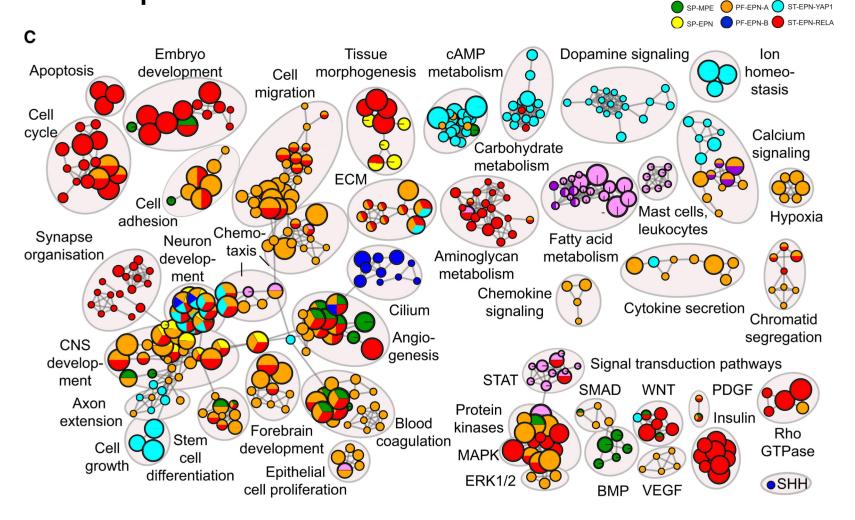


# Enrichment Map: use case III Query Set Analysis



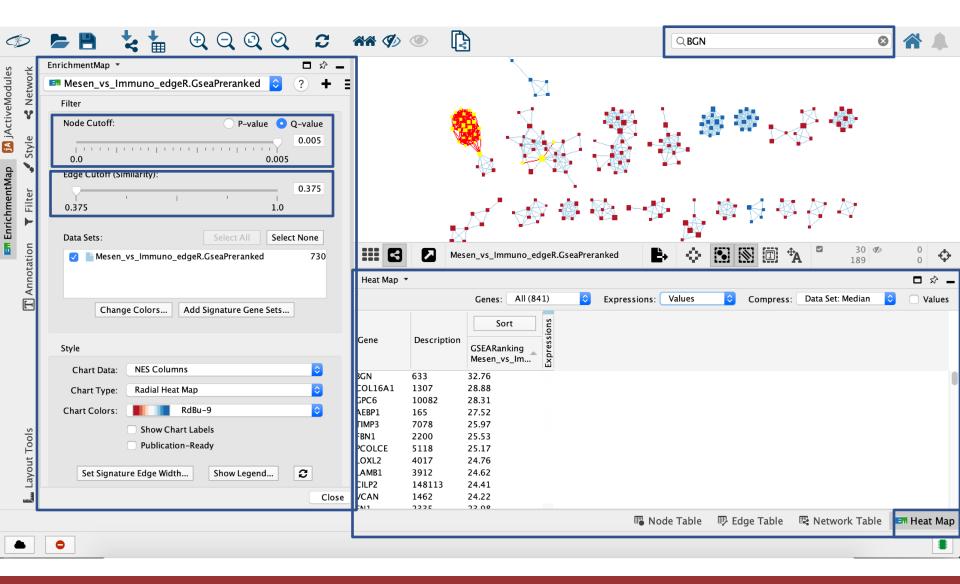
Wojtowicz EE, Lechman ER, Hermans KG, et al. Ectopic miR-125a Expression Induces Long-Term Repopulating Stem Cell Capacity in Mouse and Human Hematopoietic Progenitors. *Cell Stem Cell*. 2016;19(3):383-396

## Enrichment Map: use case IV Multiple Datasets

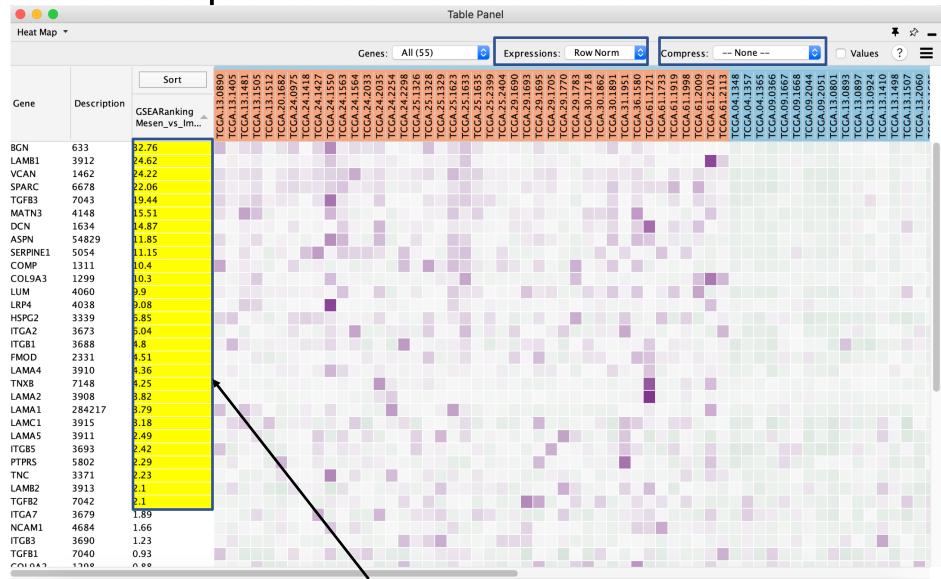


Pajtler KW, Witt H, Sill M, et al. Molecular Classification of Ependymal Tumors across All CNS Compartments, Histopathological Grades, and Age Groups. *Cancer Cell*. 2015;27(5):728-743

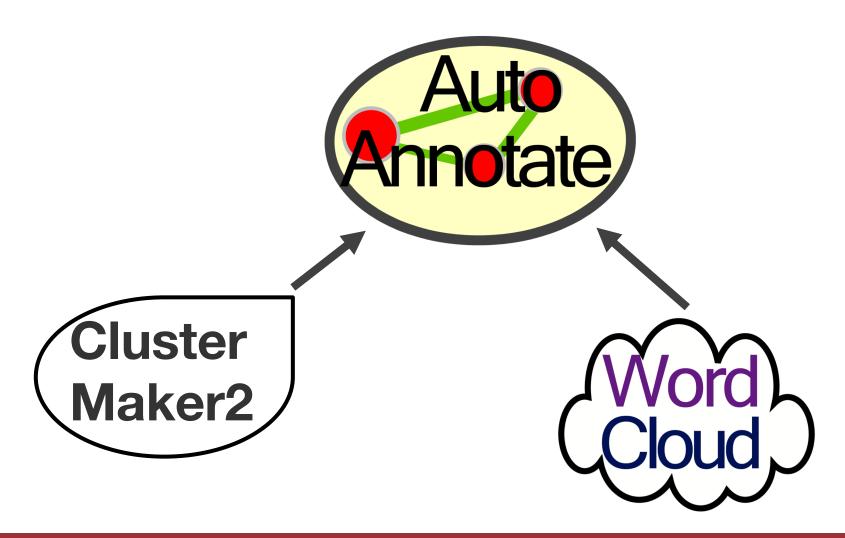
## Enrichment Map Features



### Heatmap Panel



## Enrichment Map Features



### AutoAnnotate

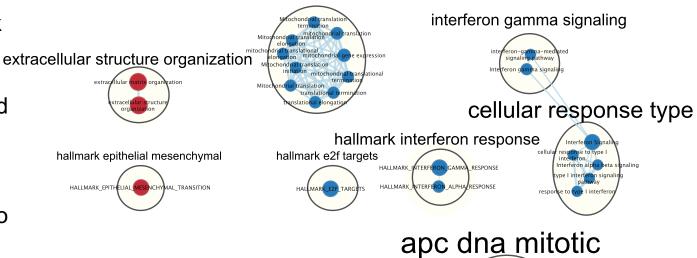
#### Steps:

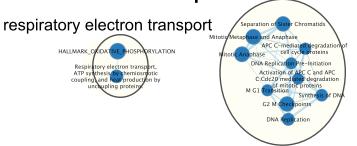
mitochondrial translational translation

Cluster the Network

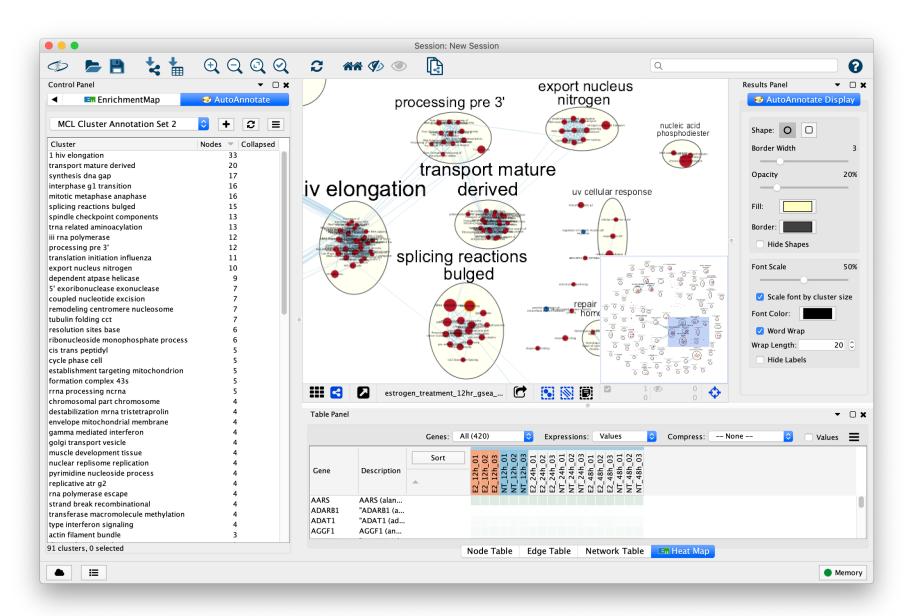
2. For each cluster, find frequent words in node labels

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

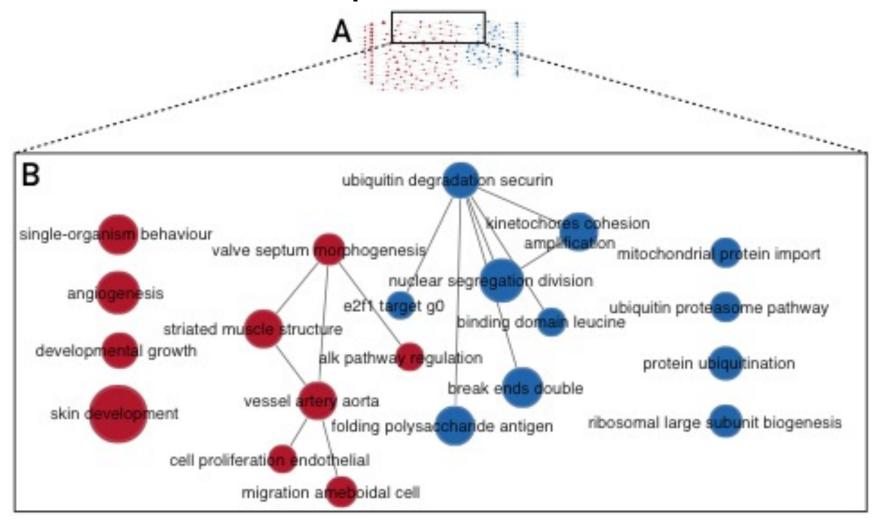


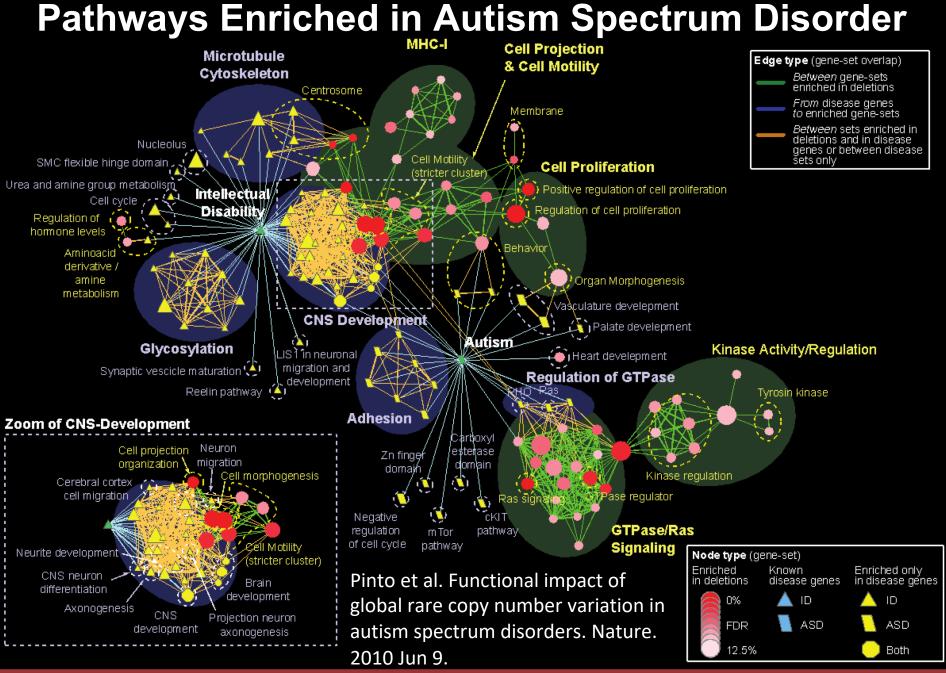


### AutoAnnotate



## 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



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