



Metabolic network analysis

Alexey Sergushichev

Sep 23, Nice



Outline

- Exploring gene expression datasets
- Simple analysis methods
- **Vorking with public datasets**



Let's open Cytoscape

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Metabolic&transcriptional profiling pipeline



Agios

Artyomov lab

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Macrophages polarization goes with high metabolic regulation



Wound healing and fibrosis

Murray&Wynn (2011), Nature Reviews Immunology

Integrating metabolic and transcriptional data



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M1 vs M2 module









Nucleic Acids Research

GAM: a web-service for integrated transcriptional and metabolic network analysis **a**

Alexey A. Sergushichev ➡; Alexander A. Loboda; Abhishek K. Jha; Emma E. Vincent; Edward M. Driggers; Russell G. Jones; Edward J. Pearce; Maxim N. Artyomov

Nucleic Acids Res (2016) 44 (W1): W194-W200. DOI: https://doi.org/10.1093/nar/gkw266 Published: 20 April 2016 Article history ▼

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https://doi.org/10.1093/nar/gkw266





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Let's go to Shiny GAM

https://artyomovlab.wustl.edu/shiny/gam/

Shiny GAM: integrated analysis of genes and metabolites

Work	Help	About									
Like Shiny GAM? Check out Phantasus where you can do differential expression and submit the results to Shiny GAM											
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	or										



Load the example data

Shiny GAM: integrated analysis of genes and metabolites

Work Help About

Like Shiny GAM? Check out Phantasus where you can do differential expression and submit the results to Shiny GAM

Reset all

✓ Example DE for genes

Example DE for metabolites

Organism: Mouse

Interpret reactions as

edges

Use RPAIRs

Run step 1, autogenerate FDRs and run step 2

or

Step 1: Make network

Differential expression for genes

- name : Ctrl.vs.MandLPSandIFNg.gene.de.tsv
- length : 16829
- ID type : RefSeq

Top DE genes:

ID	pval	log2FC	baseMean
NM_008730	2.89e-42	-12.39	490
NM_172621	3.85e-30	12.64	1388
NM_013653	2.16e-29	8.58	3164
NM_001004174	1.34e-26	8.07	3670
NM_011198	1.80e-26	7.98	1857
NM_021274	2.17e-26	8.02	3065

Differential expression for metabolites

- name : Ctrl.vs.MandLPSandIFNg.met.de.tsv
- length : 2119
- ID type : HMDB

Top DE metabolites:

ID	pval	log2FC	baseMean
HMDB00634	8.83e-34	3.12	17.1
HMDB00620	8.83e-34	3.12	17.1
HMDB02092	8.83e-34	3.12	17.1
HMDB00749	8.83e-34	3.12	17.1
HMDB10720	5.93e-31	2.51	16.0
HMDB03407	5.93e-31	2.51	16.0

Not mapped to KEGG: 570

Top unmapped metabolites: show

Network summary

There is no built network

Not mapped to Entrez: 73

Top unmapped genes: show

There are multiple ways to represent reactions as a graph



Reactions as edges

Reaction as nodes

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Create a network

Shiny GAM: integrated analysis of genes and metabolites

Work Help About Like Shiny GAM? Check out Phantasus where you can do differential expression and submit the results to Shiny GAM Differential expression for genes Differential expression for metabolites Reset all name : Ctrl.vs.MandLPSandIFNg.gene.de.tsv name : Ctrl.vs.MandLPSandIFNg.met.de.tsv Example DE for genes length : 16829 length : 2119 ID type : RefSeq ID type : HMDB Example DE for metabolites Top DE genes: Top DE metabolites: Organism: Mouse log2FC ID pval baseMean ID pval log2FC baseMean Interpret reactions as NM 008730 2 89e-42 -12.39 490 HMDB00634 8.83e-34 3 12 17 1 edges 72621 3.85e-30 12.64 1388 HMDB00620 8.83e-34 3.12 17.1 NM 013653 8.58 3164 17.1 -29 HMDB02092 8.83e-34 3.12 Use RPAIRs 8.07 NM 001004174 1.34e-26 3670 HMDB00749 8.83e-34 3 12 17 1 Run step 1, autogenerate NM 011198 7.98 1857 2.51 16.0 1.80e-26 HMDB10720 5.93e-31 FDRs and run step 2 HMDD02407 NM 021274 2.17e-26 8.02 3065 5.93e-31 2.51 16.0 or Reactions as edges Not mapped to Entrez: 73 Not mapped to KEGG: 570 Step 1: Make network Top unmapped genes: show Top unmapped metabolites: show Click "Make Download the network Network summary number of nodes : 1645 · number of edges : 1649 network" ▲ Download XGMML



Cytoscape: loading network

- Start New Session > From Network File > network.mmu.xgmml
 - or File > Import > Network > File
- Layout > Apply Preferred Layout (F5)





Cytoscape: editing style

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Cytoscape: set label style

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Cytoscape: set shape



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Cytoscape: set fill color



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Cytoscape: set node size



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Now it's better

File Edit View Select Layout Apps Tools Help



20



Cytoscape: loading style

Ownload style XML-file from server:

- <u>http://artyomovlab.wustl.edu/pu</u> <u>blications/supp_materials/GAM/</u> <u>GAM_VizMap.xml</u>
- File > Import > Styles > GAM_VizMap.xml
- Select "GAM" instead of "default"
- Styles > Options > Make Current Styles Default

Or download from here:

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🛓 Cytoscape VizMap



Cytoscape: loading style (2)

A



Memory



Zoom in to Itaconate and Irg1



Reaction ID: R02243



Look at Irg1 reaction (R02243) at KEGG

K CC	REACTION: R02243	
Entry	R02243 Reaction	All links
Name	cis-Aconitate carboxy-lyase	Optology (1)
Definition	cis-Aconitate <=> Itaconate + CO2	KEGG BRITE (1)
Equation	C00417 <=> C00490 + C00011	Pathway (2)
	$H \rightarrow H \rightarrow$	Chemical substance (3) KEGG COMPOUND (3) Chemical reaction (2) KEGG ENZYME (1) KEGG RCLASS (1) Gene (193) KEGG ORTHOLOGY (1) KEGG GENES (192) All databases (201) Download RDF
Reaction clas	s RC00667 C00417_C00490	
Enzyme	4.1.1.6	
Pathway	rn00660 C5-Branched dibasic acid metabolism	
Orthology	K17724 aconitate decarboxylase [EC:4.1.1.6]	
Other DBs	RHEA: 15256	

DBGET integrated database retrieval system

http://www.genome.jp/dbget-bin/www_bget?rn:R02243



The network is too big! Let's try to extract its most important part.

Jelly beans mining problem





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http://fineartamerica.com/featured/jelly-beans-background-ken-brown.html https://sysbio.curie.fr/bionetvisa/BioNetVisA2014_Presentations/BioNetVisA2014_Oral_Presentations/T alk08_Mohammed_ElKarib_BioNetVisa_2014.pdf

Jelly beans mining problem





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Generalized maximum-weight connected subgraph problem (GMWCS)

Definition 2. Given a connected undirected graph G = (V, E) and a weight function $\omega : (V \cup E) \to \mathbb{R}$, the Generalized Maximum-Weight Connected Subgraph (GMWCS) problem is the problem of finding a connected subgraph $\widetilde{G} = (\widetilde{V}, \widetilde{E})$ with the maximal total weight

$$\Omega(\widetilde{G}) = \sum_{v \in \widetilde{V}} \omega(v) + \sum_{e \in \widetilde{E}} \omega(e) \to max$$



http://dx.doi.org/10.1007/978-3-319-43681-4_17 https://github.com/ctlab/gmwcs-solver/

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Network scoring

- Small p-vaule -> positive score
- Close to one p-value -> negative score
- FDR thresholds control which genes and metabolites have positive score





Assign network scores





Loading module into Cytoscape

File > Import > Network > File >

Ctrl.vs.MandLPSandIFNg.gene.de.re.mf=-0.9.rf=-2.5.ams=-11.7.xgmml

Layout > Apply preferred layout



M0 vs M1 module

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Save the figure

File > Export as image...



PDFs can be opened in Inkscape or Illustrator

Running from Phantasus

Open GSE53986.Ctrl.vs.LPS.gct in Phantasus

Tools/Submit to Shiny GAM



Remember that only expressed genes are present

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The module

-6.1

🕹 PDF

Log2FC:





Exercise: upload data manually

- Open GSE53986.Ctrl.vs.LPS.gct in Excel
- Save differential expression table as a tab-separated file
- Upload to GAM

We got the M1 vs M2 module, what's next?



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UDP-GlcNAc and **Glutamine** modules



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Test glutamine involvement by using media without one



- complete RPMI 1640
- 10 mM glucose
- 2 mM L-glutamine,
- 100 U/mL of penicillin/streptomycin
- 10% FCS

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Test glutamine involvement by using media without one



- complete RPMI 1640
- 10 mM glucose
- 2 mM L-glutamine,
- 100 U/mL of penicillin/streptomycin
- 10% FCS



Glutamine withdrawal leads to defect in M2 polarization







Glutamine withdrawal affect TCA cycle in M2



Oxidative phosphorylation is impaired upon glutamine withdrawal in M2 cells

Glutamine feeds into UDP-GlcNAc

Major metabolic fluxes in M2 macrophages

N-glycosylation inhibition blocks M2 polarization

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Immunity

Volume 42, Issue 3, 17 March 2015, Pages 419-430

Article

Network Integration of Parallel Metabolic and Transcriptional Data Reveals Metabolic Modules that Regulate Macrophage Polarization

Abhishek K. Jha¹, Stanley Ching-Cheng Huang², Alexey Sergushichev^{2, 3}, Vicky Lampropoulou², Yulia Ivanova², Ekaterina Loginicheva², Karina Chmielewski¹, Kelly M. Stewart¹, Juliet Ashall², Bart Everts^{2, 5}, Edward J. Pearce^{2, 4}, Edward M. Driggers^{1, 4, 6}, **a**, **w**, Maxim N. Artyomov^{2, 4}, **a**, **w**

Show more

Highlights

- · Glutamine deprivation affects M2 polarization but not M1 polarization
- · UDP-GlcNAc biosynthesis and N-glycosylation are important for M2 polarization
- · There is no reverse or direct flow through Idh or malic enzyme in M1 macrophages
- · Aspartate-arginosuccinate shunt connects the NO and TCA cycles in M1 polarization

http://dx.doi.org/10.1016/j.immuni.2015.02.005

Atom transition network

http://maranas.che.psu.edu/research_pathways.htm

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Using atom transitions graph limits "bad" connections

Any path in an atom graph corresponds to a potential reaction flux

Trem2 KO in BMDM leads to decrease in energetic metabolism

Ulland et al, Cell 2017

GATOM

- <u>https://github.com/ctlab/gatom</u>
- Can be installed and ran in R
- ♥ To be available in Shiny GAM

We live in a high-throughput era, there can be many conditions!

Dendritic cells time-course: many conditions, how to analyze?

GSE59784, Jovanovic et al, 2015

Can we identify modules for the expression patterns?

GSE59784, Jovanovic et al, 2015

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Glycolysis transcriptionally goes up at 12h

Sterol synthesis/degradation goes down at 6h

Can find differences in metabolic regulation between cell populations

Artyomov et al. Seminars in Immunology 2016

Pipeline for single-cell RNA-seq

- Dataset for small intestinal epithelium (Haber et al, Nature 2017).
- Total: 7216 cells.

- Cells are fine-clustered into ~100 groups.
- Clustering is independent of cell type annotation.

- Gene expression in cell groups are averaged to form a "metasample".
- Enzyme are k-means clustered by expression in meta-samples.

- GAM for multiple conditions is applied to obtain final module patterns.
- In the intestine dataset we ended up with nine modules.

Glycolys is regulated in the immunotherapy dataset: macrophages getting activated

Pkm Aldoa Tpi1 Eno1 Gapdh Pgam1 Ldha Pgk1 Gpi1 Pfkl Ugp2 Pgm2

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- Network integration allows to highlight interplay between enzymes and metabolites
- Vetwork visualization (and some analysis) can be done in Cytoscape
- Modules need to be interpreted and validated

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