

Applying Information Visualization Principles to Biological Network Displays

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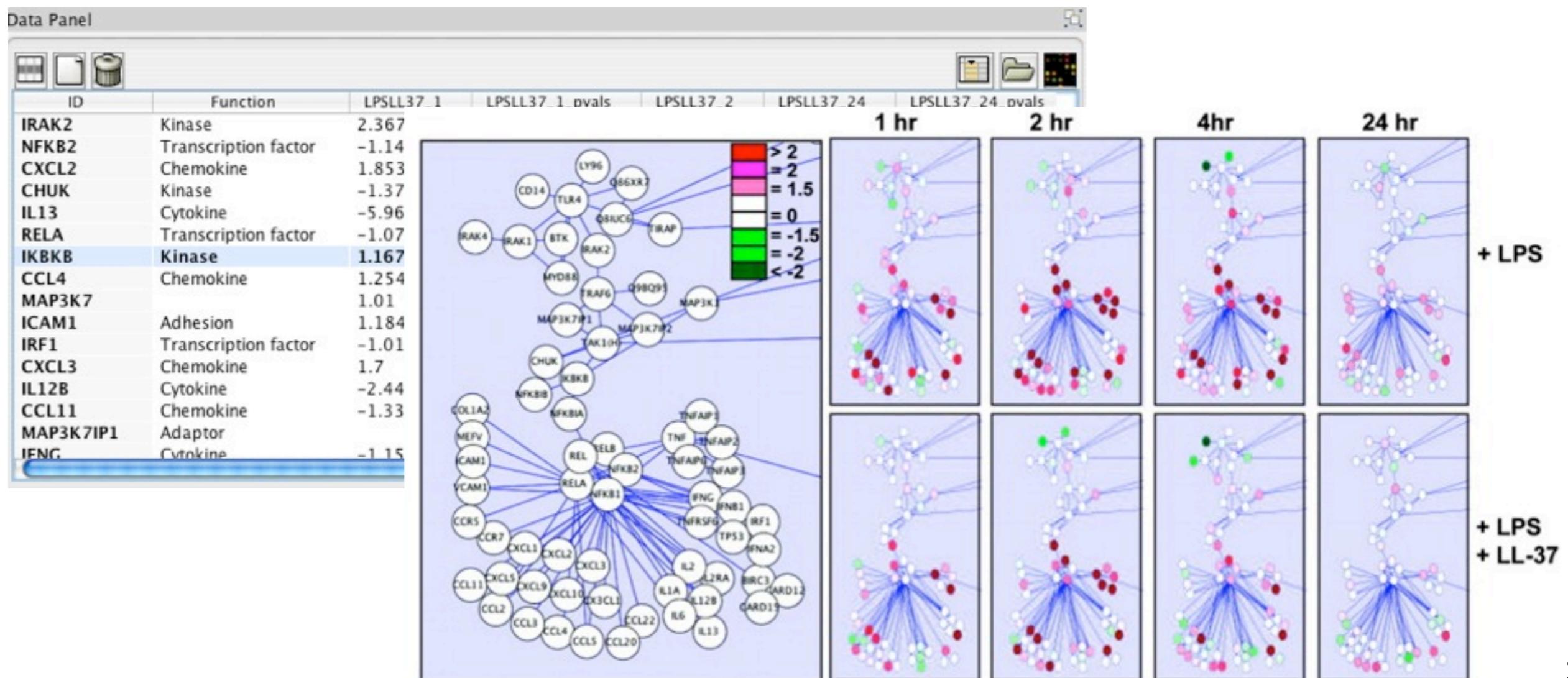
Human Vision and Electronic Imaging 2011
25 Jan 2011

Outline

- visualization principles
- Cerebral system
 - combining interaction networks with microarray data
- Pathline system
 - combining multiple genes, time points, species, and pathways

Why do visualization?

- pictures help us think
 - substitute perception for cognition
 - external memory: free up limited cognitive/memory resources for higher-level problems

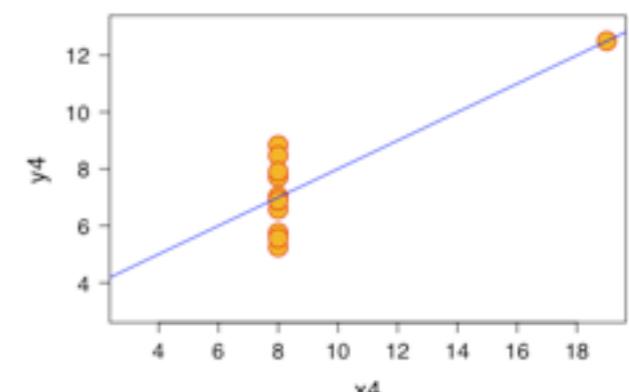
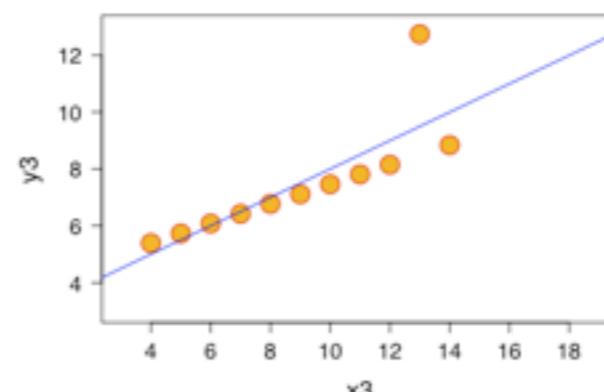
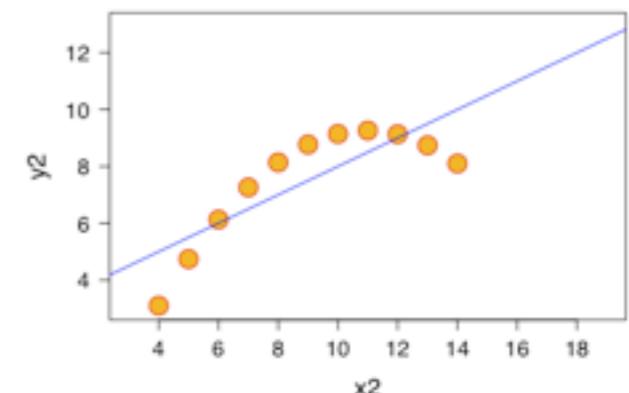
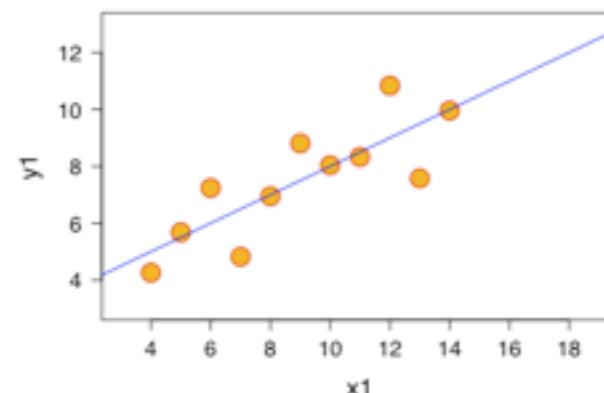


When should we bother doing vis?

- need a human in the loop
 - augment, not replace, human cognition
 - for problems that cannot be (completely) automated
- simple summary not adequate
 - statistics may not adequately characterize complexity of dataset distribution

Anscombe's quartet:
same

- mean
- variance
- correlation coefficient
- linear regression line



What does visualization allow?

- discovering new things
 - hypothesis generation, discovery, eureka moment
- confirming conjectured things
 - hypothesis confirmation
- contradicting conjectured things
 - especially (inevitably?) data cleansing
- novel capabilities
 - tool supports fundamentally new operations
- **speedup**
 - tool accelerates workflow (most common!)

Separate visualization concerns into four levels

problem

abstraction

encoding and interaction

algorithm

- different threats to validity at each level

Characterizing problems of real-world users

problem

 data/op abstraction

 encoding/interaction

 algorithm

- understanding domain concepts and current workflow
- finding gaps, breakdowns, slowdowns
 - where conjecture that vis would help
- threat to validity: users don't do that

Abstracting into operations on data types

problem

data/op abstraction

encoding/interaction

algorithm

- operations
 - sorting, filtering, browsing, comparison, characterizing trends and distributions, finding anomalies and outliers, finding correlation...
- data types
 - number tables, relational networks, spatial
 - transform into useful configuration: derived data
- threat to validity: you're showing them the wrong thing

Designing visual encoding and interaction tech

problem

data/op abstraction

encoding/interaction

algorithm

- **visual encoding**
 - marks: points, lines, areas
 - attributes: position, color, shape, size, orientation, ...
- **interaction**
 - selecting, navigating, ordering,...
- **threat to validity: the way you show it doesn't work**

LES VARIABLES DE L'IMAGE

XY
2 DIMENSIONS
DU PLAN

Z
TAILLE

VALEUR

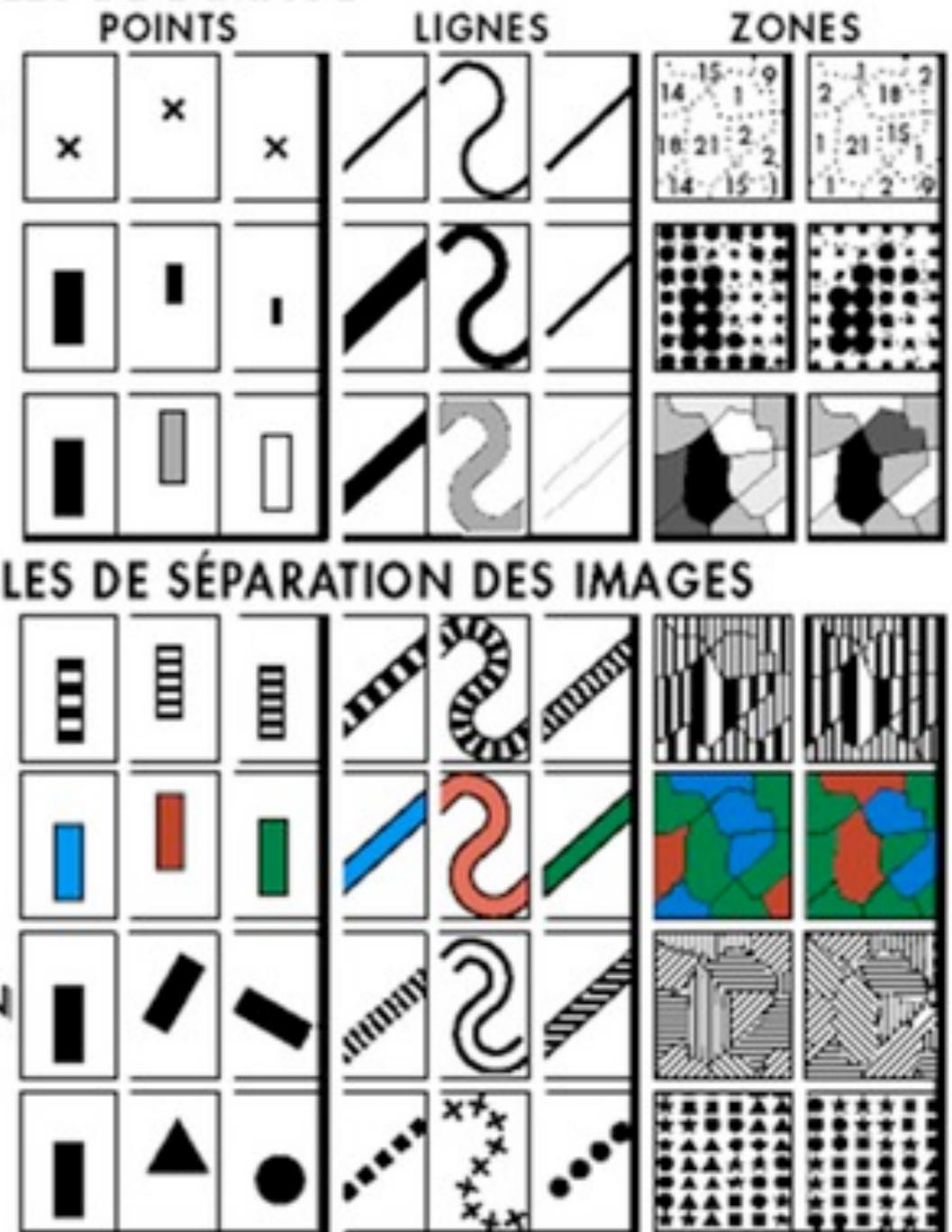
LES VARIABLES DE SÉPARATION DES IMAGES

GRAIN

COULEUR

ORIENTATION

FORME



Creating algorithms to execute techniques

problem

 data/op abstraction

 encoding/interaction

 algorithm

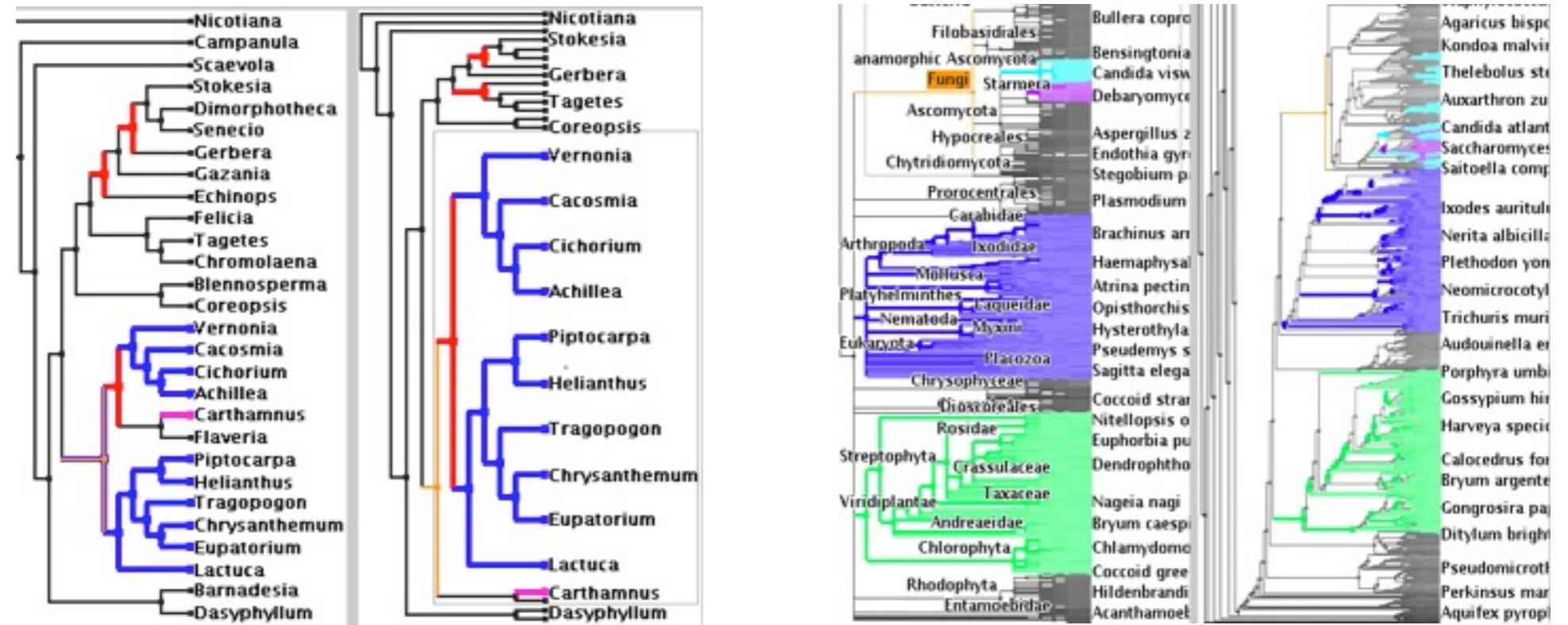
- classic computer science problem
 - create algorithm given clear specification
- threat to validity: your code is too slow

Design decisions

- huge space of design alternatives
- many choices are ineffective
 - wrong visual encoding can mislead, confuse
 - principled reasons to make choices usually not obvious to untrained people
 - conflicting tradeoffs
 - iterative refinement often necessary

Principles in action: walk through examples

- vis work in many domains
 - topology
 - computer networking
 - computational linguistics
 - web logs
 - large-scale system administration
 - ...
 - biology**



TreeJuxtaposer

Scalable Phylogenetic Tree Comparison

joint work with:

François Guimbretière, Serdar Tasiran, Li Zhang, Yunhong Zhou

<http://olduvai.sf.net/tj>

TreeJuxtaposer: Scalable Tree Comparison using Focus+Context with Guaranteed Visibility.
Munzner, Guimbretière, Tasiran, Zhang, Zhou. ACM SIGGRAPH 2003.

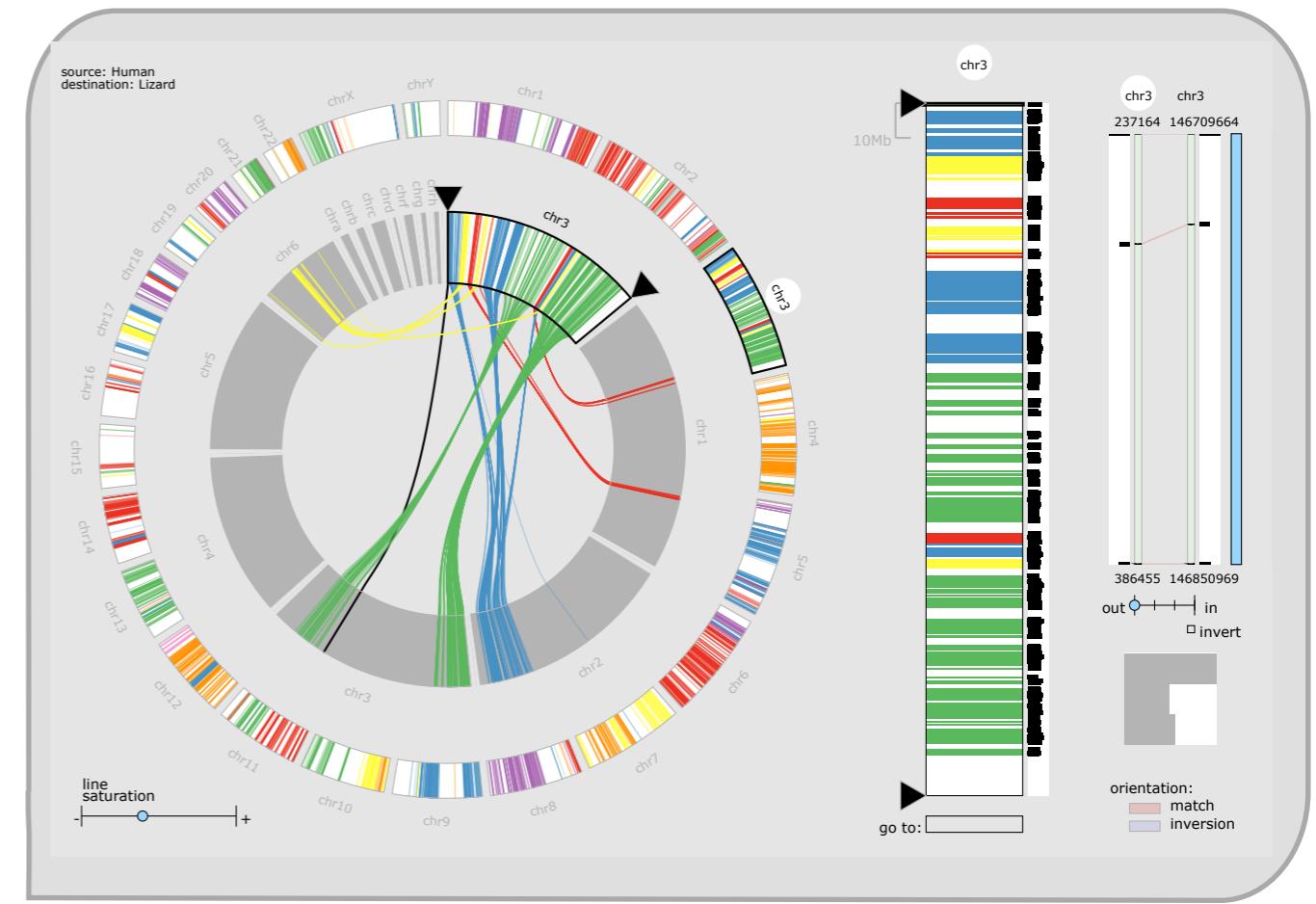
MizBee

A Browser for Comparative Genomics Data

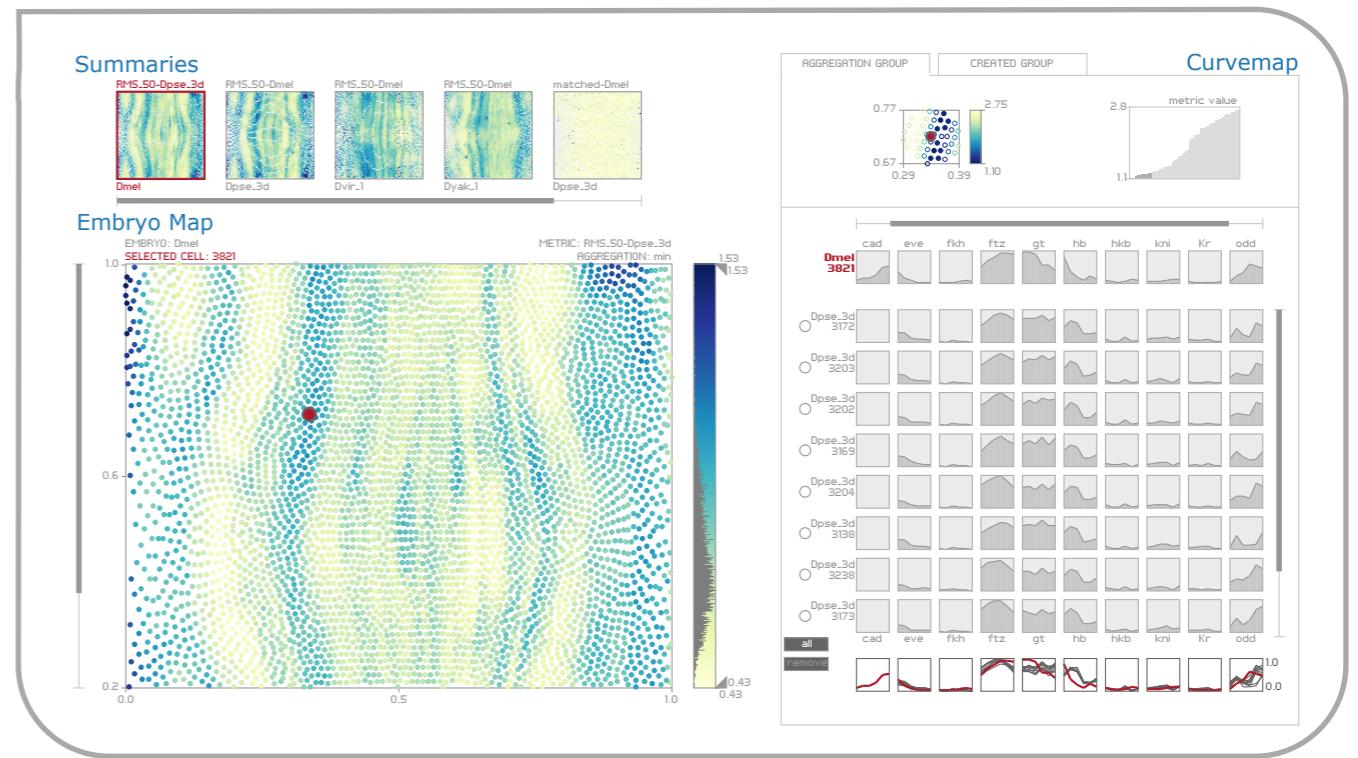
joint work with:

Miriah Meyer, Hanspeter Pfister

<http://www.mizbee.org>



MizBee: A Multiscale Synteny Browser.
Meyer, Munzner, Pfister, IEEE InfoVis 2009.



MulteeSum

A Tool for Exploring Space-Time Expression Data

joint work with:

Miriah Meyer, Angela DePace, Hanspeter Pfister

<http://www.multeesum.org>

MulteeSum: A Tool for Comparative Spatial and Temporal Gene Expression Data.
Meyer, Munzner, DePace, Pfister. IEEE InfoVis 2010.

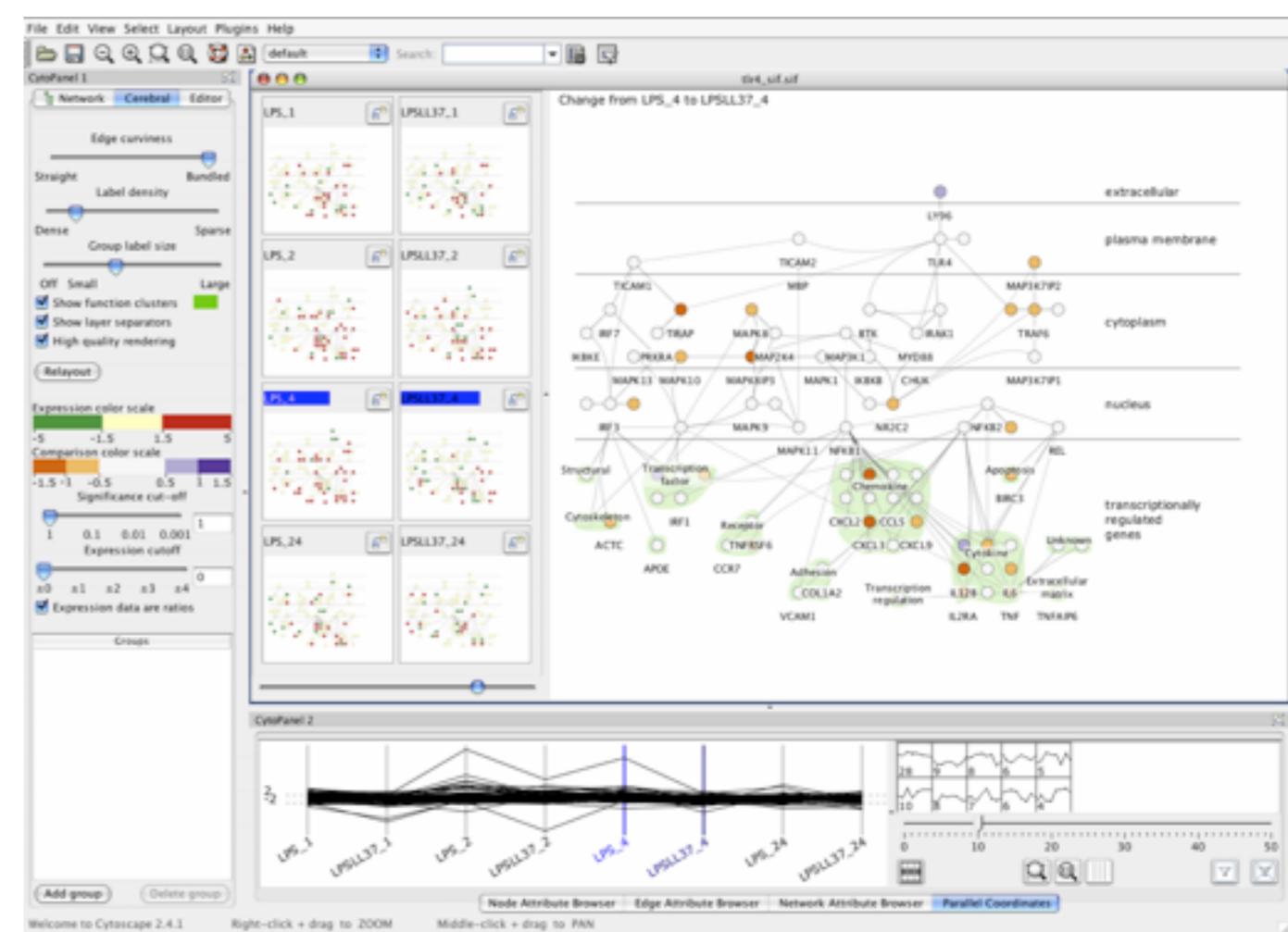
Cerebral

*Comparing Multiple Experimental Conditions
Within Biologically Meaningful Network Context*

joint work with:

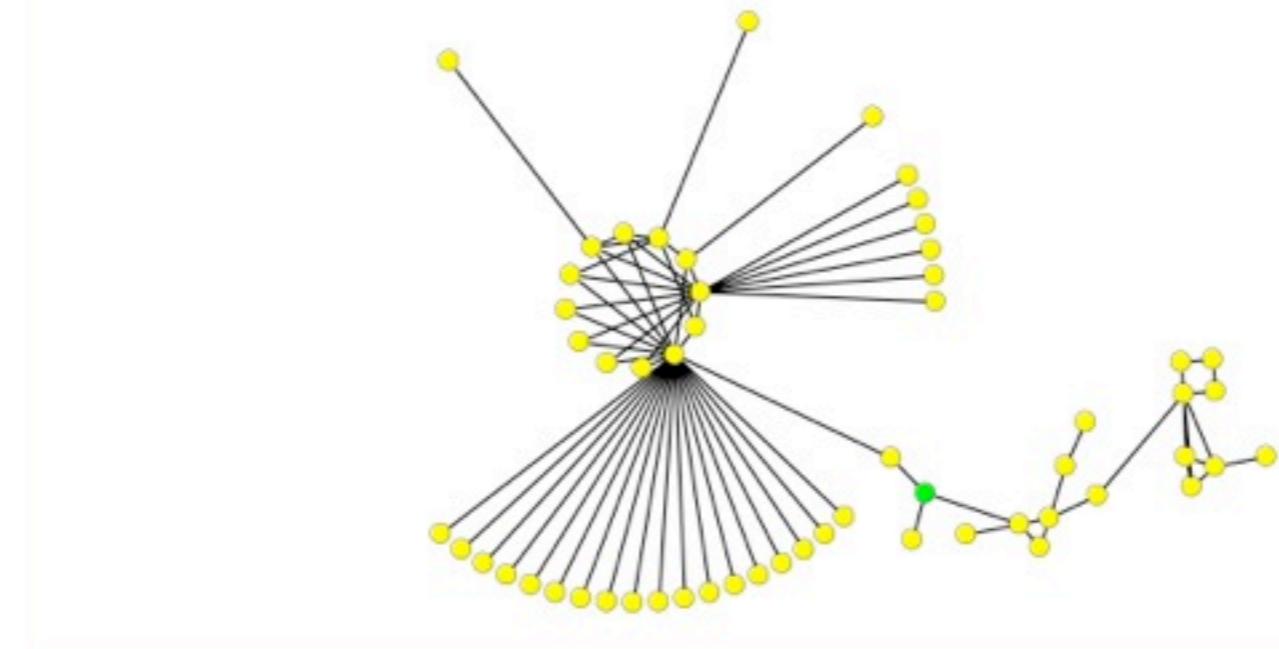
Aaron Barsky, Jennifer Gardy, Robert Kincaid

<http://www.pathogenomics.ca/cerebral/>

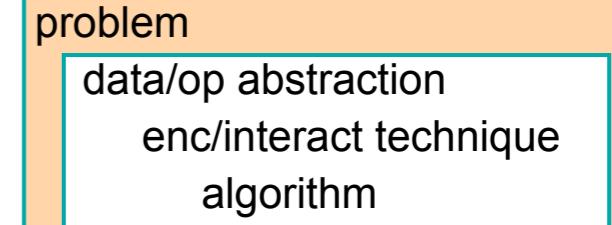


Systems biology model

- $\text{graph } G = \{\mathcal{V}, \mathcal{E}\}$
 - \mathcal{V} : proteins, genes, DNA, RNA, tRNA, etc.
 - metadata: labels, biological attributes
 - \mathcal{E} : interacting molecules
 - known from previous research



Cycle: model - experiment

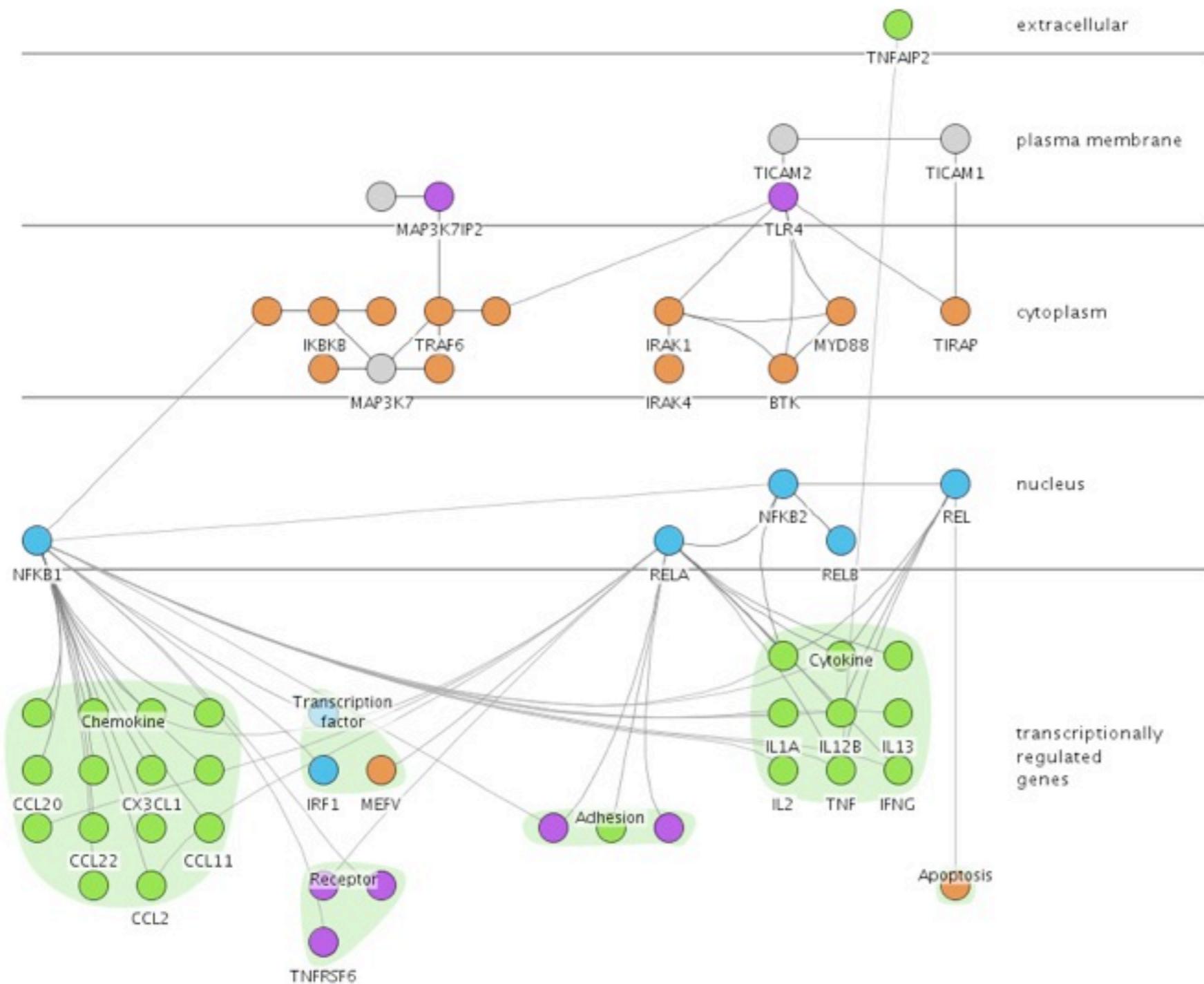


- conduct experiments on cells
 - microarrays
 - measurements for each vertex in graph
- interpret results in current graph model
- propose modifications to refine model
- vis tool to accelerate workflow
 - integrated tool to see graph and measurements together
 - choose scope for problem complexity

ID	Function	LPSLL37_1	LPSLL37_1_pvals	LPSLL37_2	LPSLL37_24	LPSLL37_24_pvals
IRAK2	Kinase	2.367	0.251	1.337	-1.553	0.807
NFKB2	Transcription factor	-1.14	0.972	-1.03	1.303	0.745
CXCL2	Chemokine	1.853	0.376	4.111	-1.019	0.387
CHUK	Kinase	-1.376	0.373	2.232	1.194	0.601
IL13	Cytokine	-5.961		2.139	-1.236	0.594
RELA	Transcription factor	-1.077	0.564	-1.169	1.943	0.761
IKBKB	Kinase	1.167	0.29	1.421	-1.907	0.286
CCL4	Chemokine	1.254	0.878	-1.052	1.499	0.8
MAP3K7		1.01	0.956	-1.096	1.222	0.671
ICAM1	Adhesion	1.184	0.669	1.537	1.392	0.995
IRF1	Transcription factor	-1.013	0.519	1.416	1.081	0.521
CXCL3	Chemokine	1.7	0.905	1.092	-1.598	0.08
IL12B	Cytokine	-2.448	0.042	-1.473	-2.109	0.129
CCL11	Chemokine	-1.338	0.349	-1.995	-1.785	0.521
MAP3K7IP1	Adaptor					
IFNG	Cytokine	-1.15	0.401	1.075	1.053	0.521

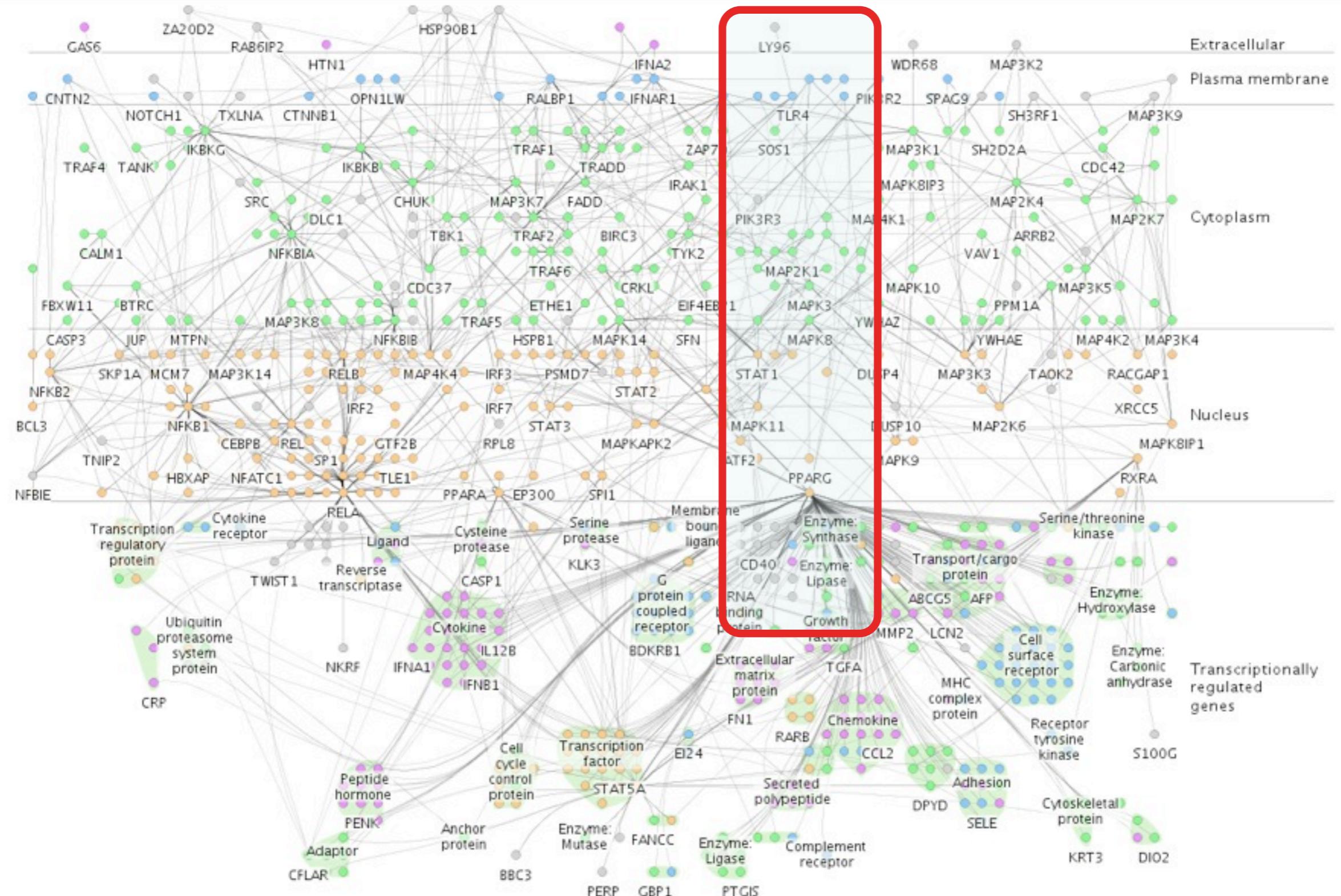
TLR4 biomolecule: E=74, V=54

- very local view

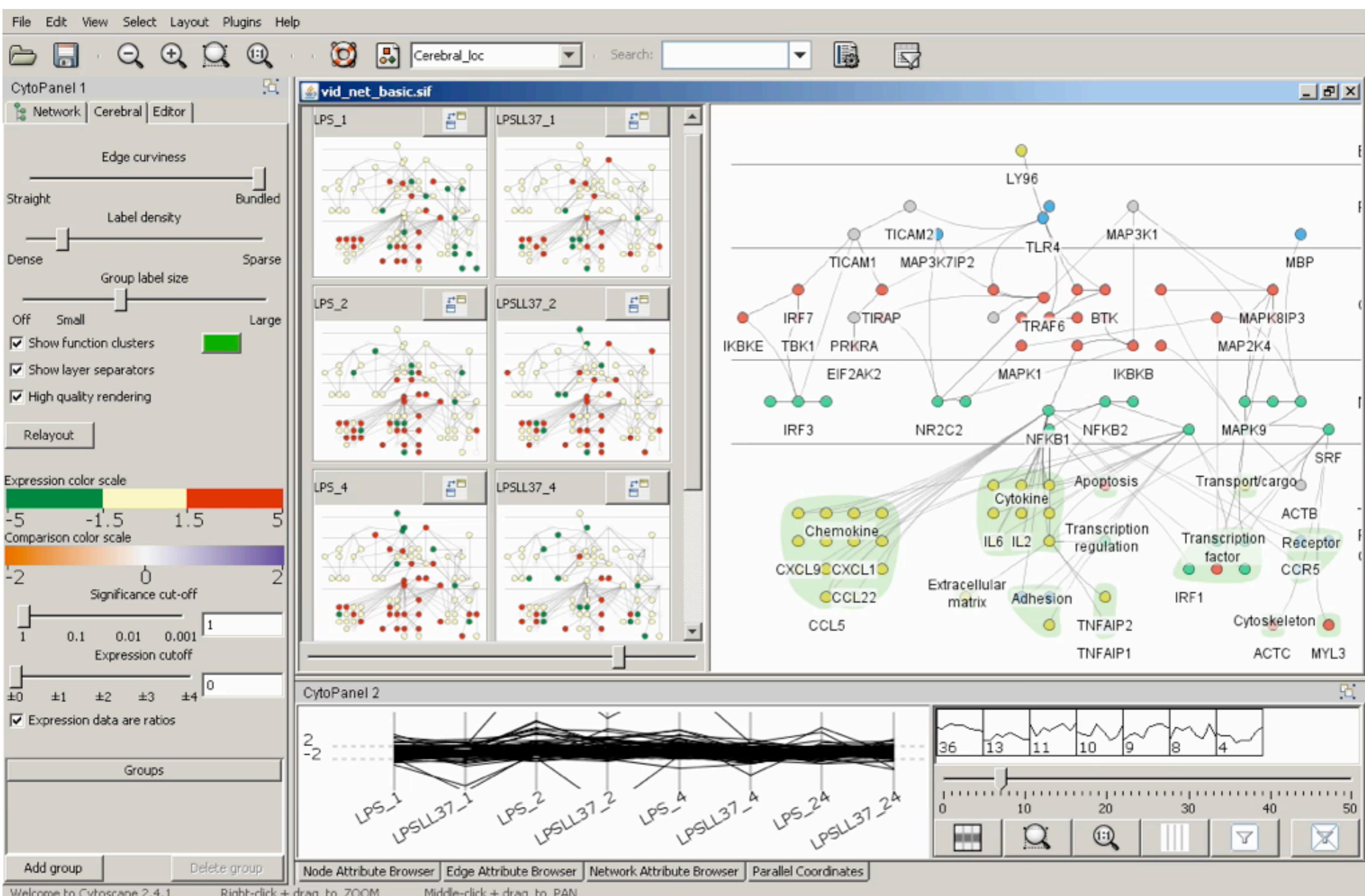


Immune system: E=1263, V=760

- bigger picture, target size for Cerebral



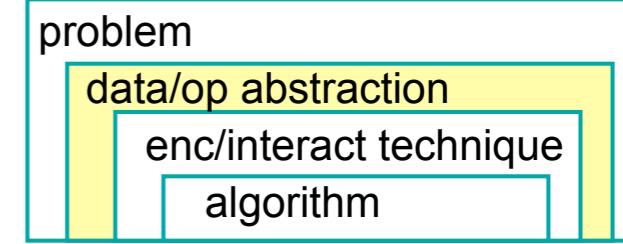
Cerebral video



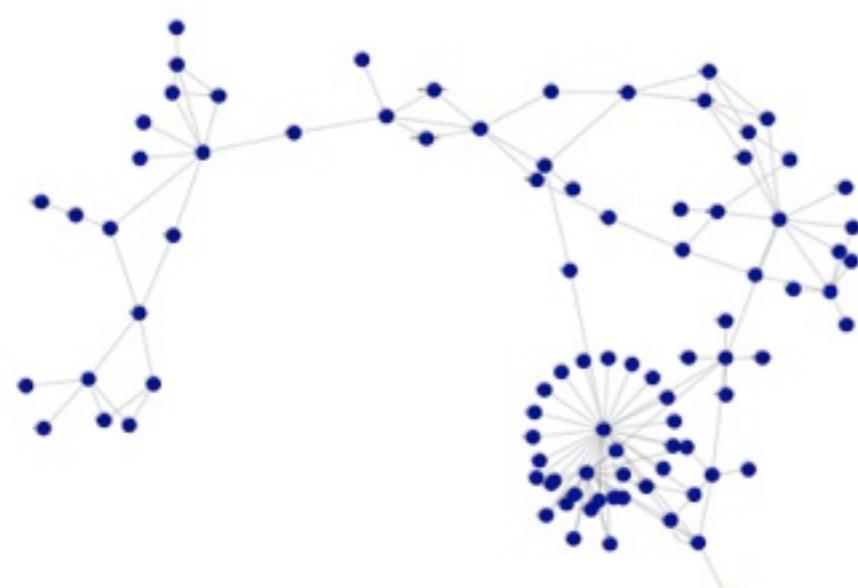
Encoding and interaction design decisions

- create custom graph layout
 - guided by biological metadata
- use small multiple views
 - one view per experimental condition
- show measured data in graph context
 - not in isolation

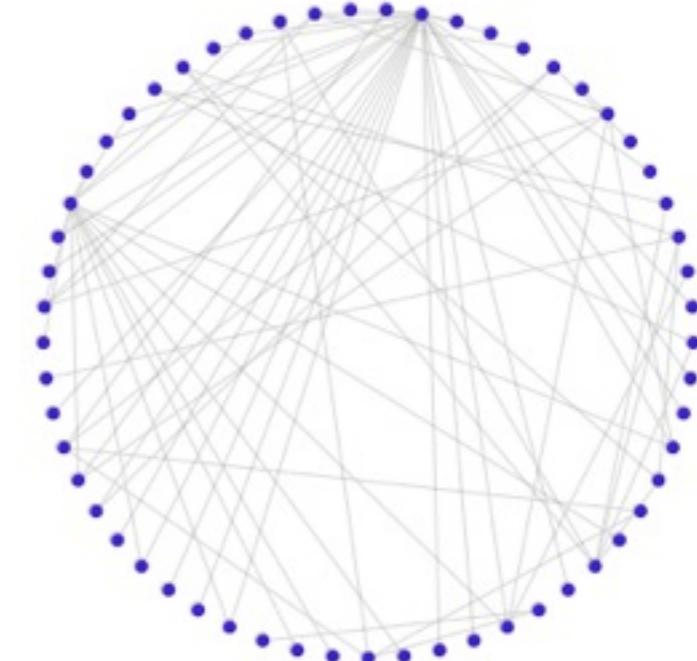
Choice: Create custom graph layout



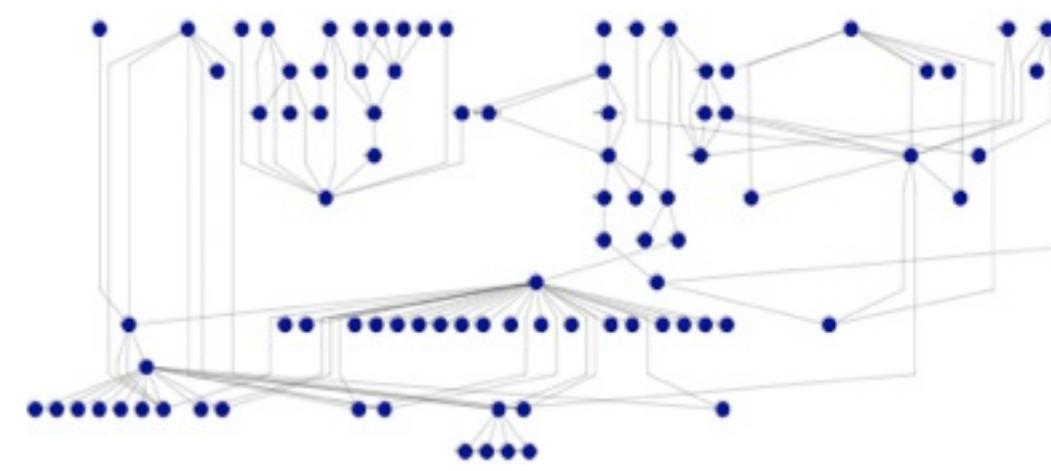
- graph layout heavily studied
 - given graph $G=\{V,E\}$,
 - create layout in 2D/3D plane
 - hundreds of papers
 - annual Graph Drawing conf.



Force-directed
(Fruchterman and Reingold, 1991)



Circular (Six and Tollis, 1999)



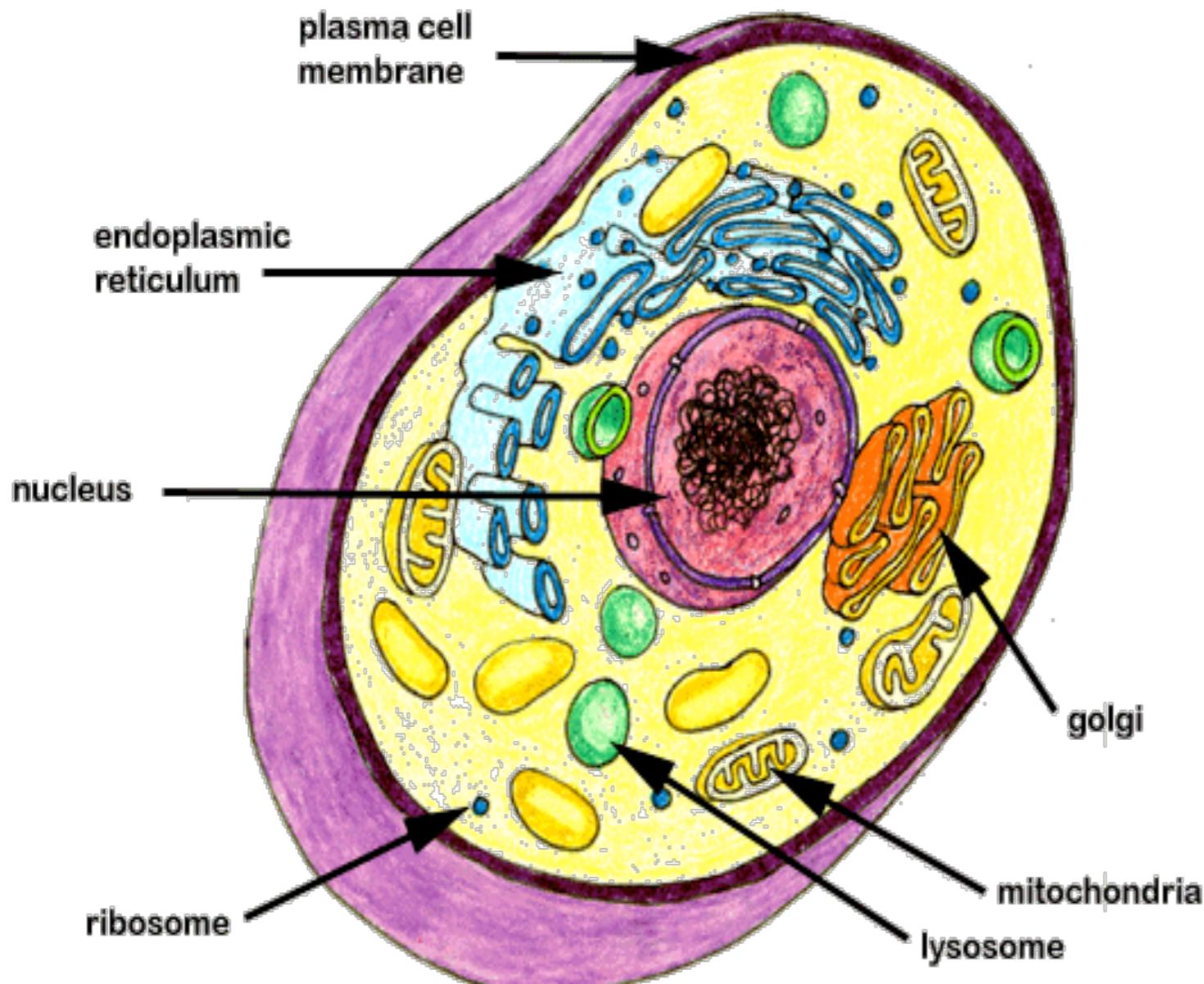
Hierarchical (Sugiyama 1989)

Existing layouts did not suit immunologists

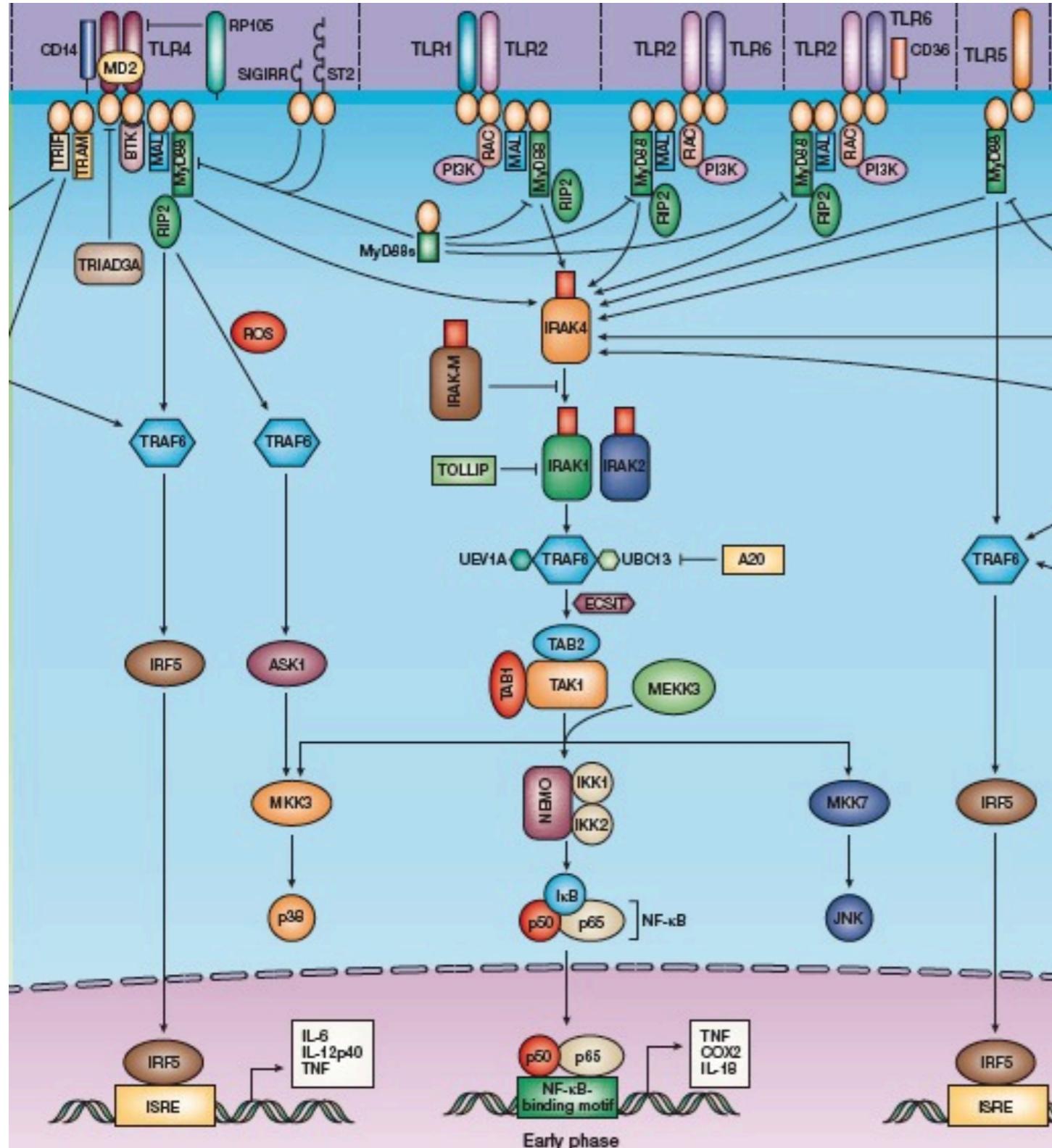
- graph drawing goals
 - visualize graph structure
- biologist goals
 - visualize biological knowledge
 - some relationships happen to form a graph
 - cell location also relevant

Biological cells divided by membranes

- interactions generally occur within a compartment
- interaction location often known as part of model



Hand-drawn diagrams



- cellular location spatially encoded vertically
- infeasible to create by hand in era of big data

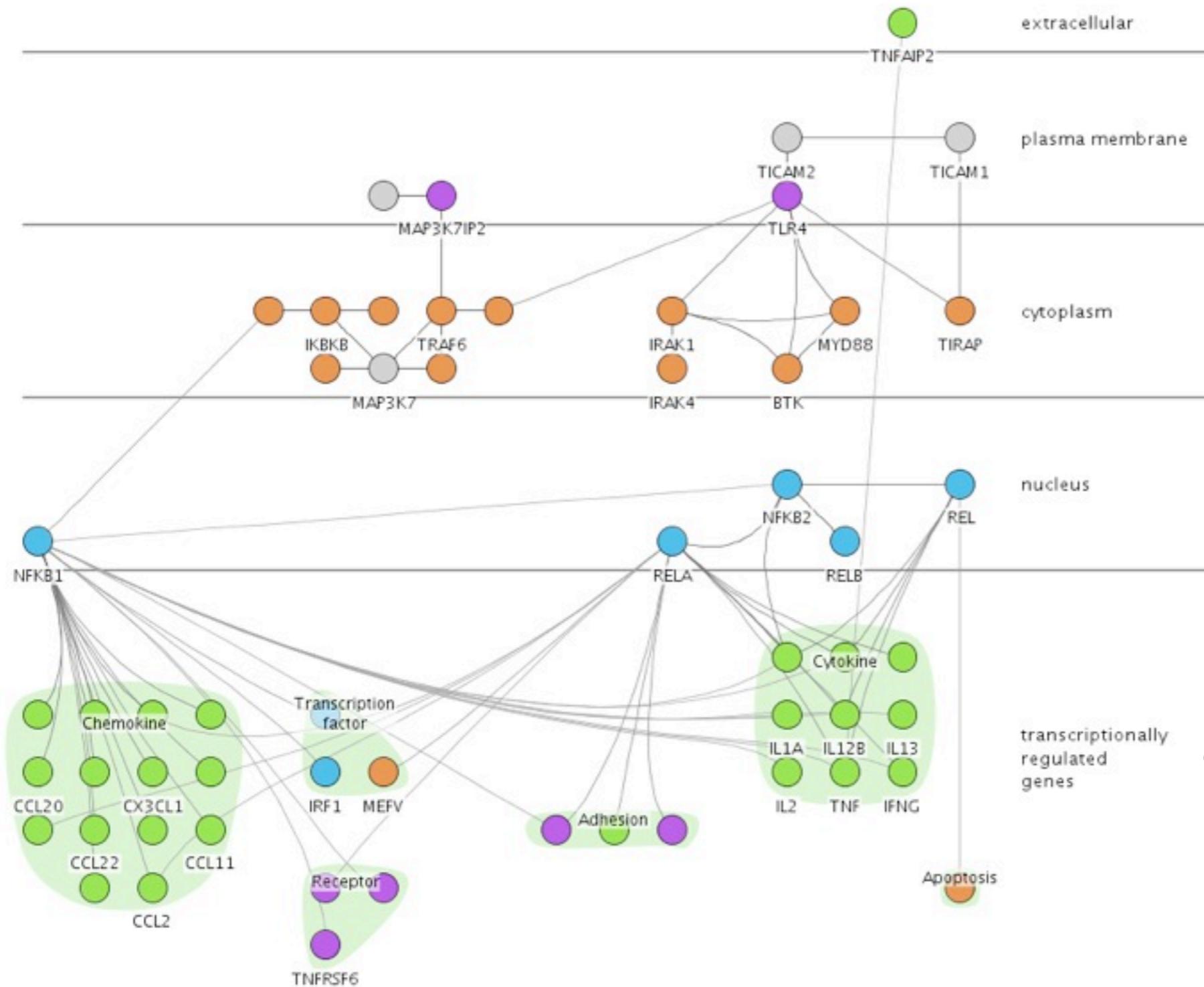
Lay out using biological metadata

problem

data/op abstraction

enc/interact technique

algorithm



- similar to hand-drawn: spatial position reveals location in cell

problem

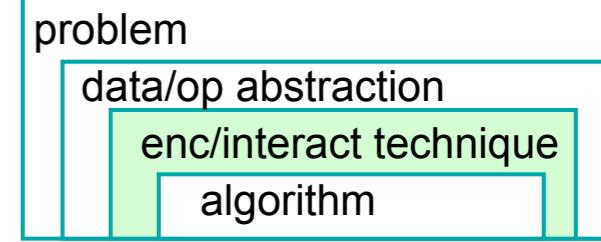
data/op abstraction

enc/interact technique

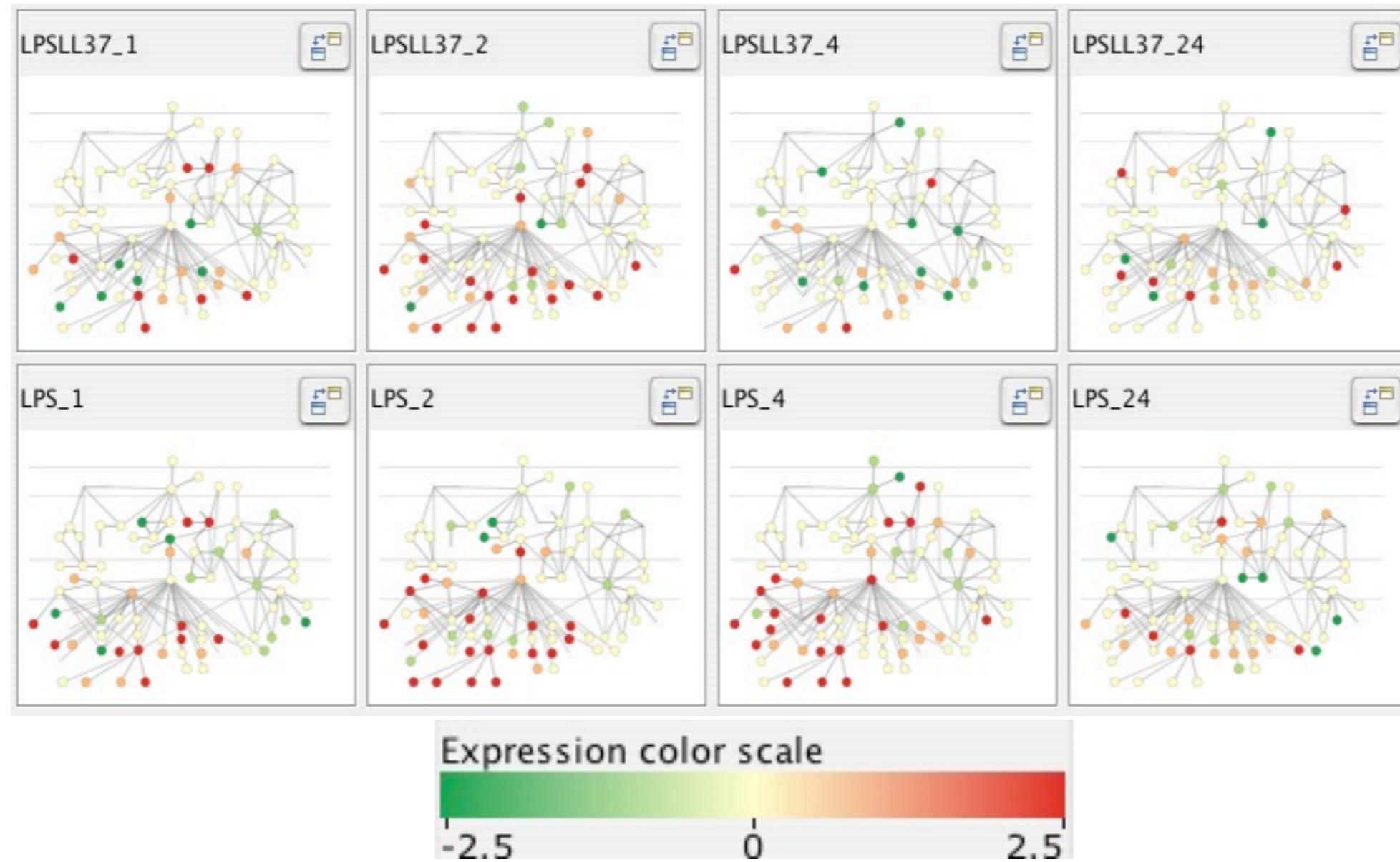
algorithm

- simulated annealing in $O(E\sqrt{V})$ vs. $O(V^3)$ time

Choice 2: Use small multiple views

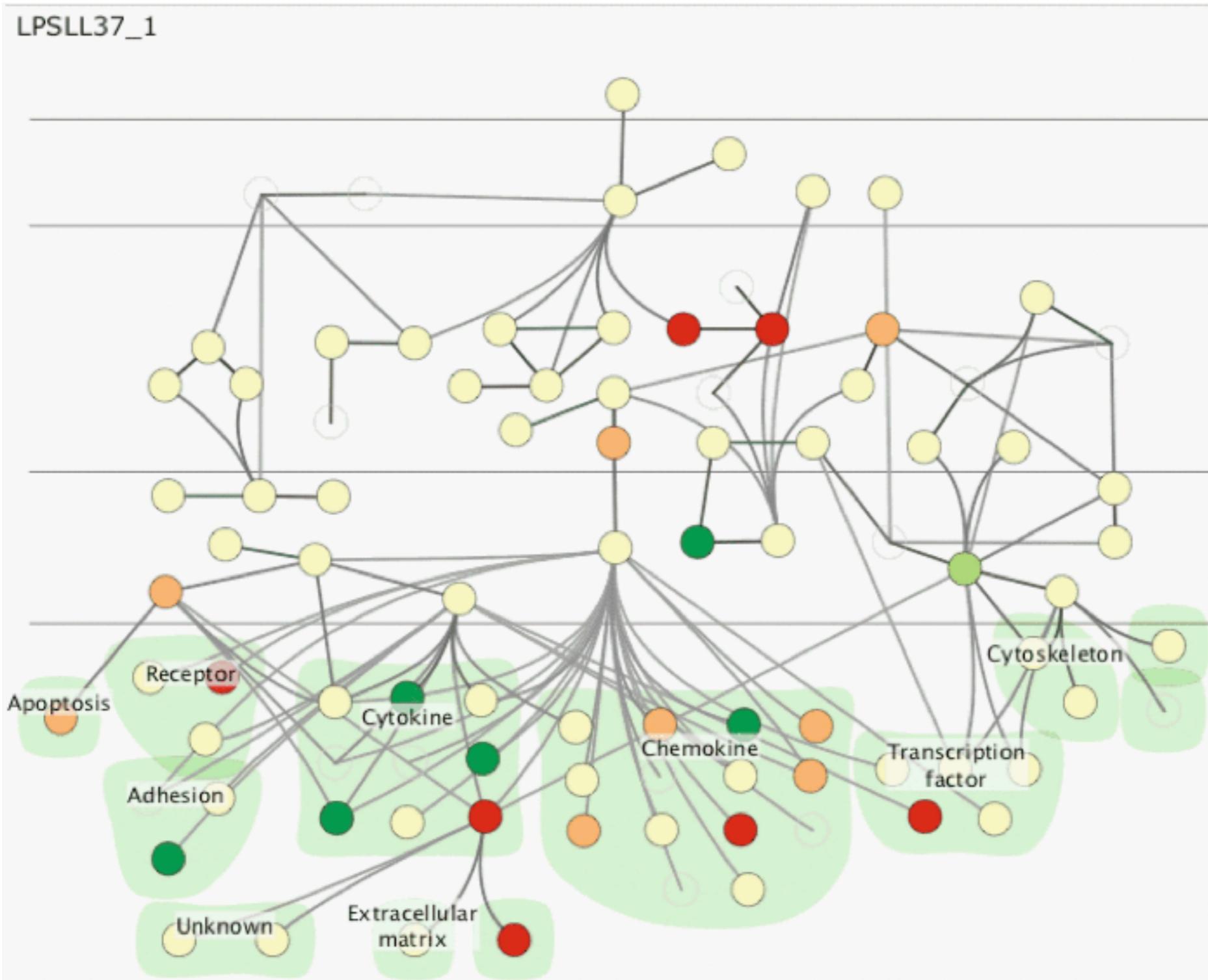


- one graph instance per experimental condition
 - same spatial layout
 - color differently, by condition



Why not animation?

- global comparison difficult

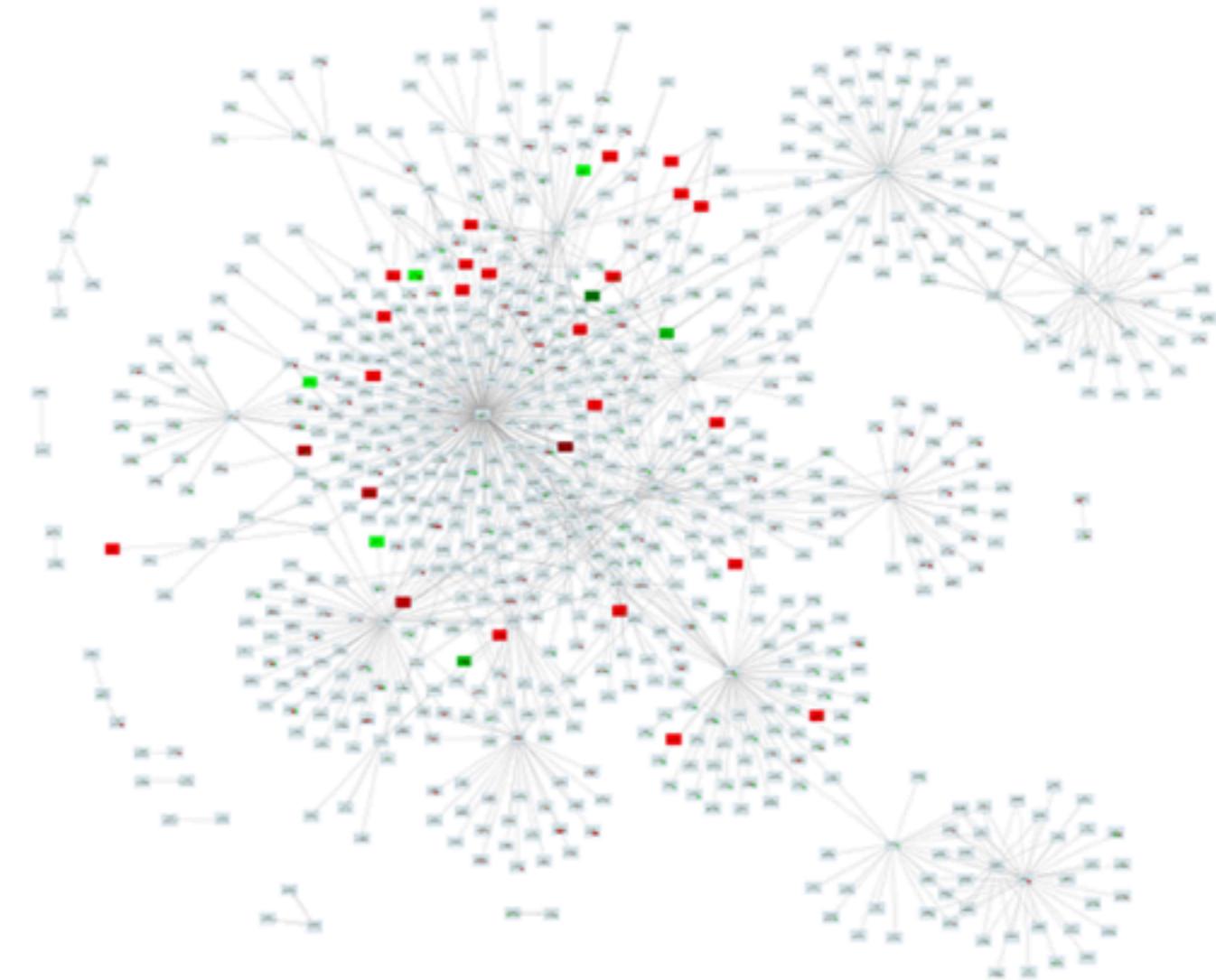
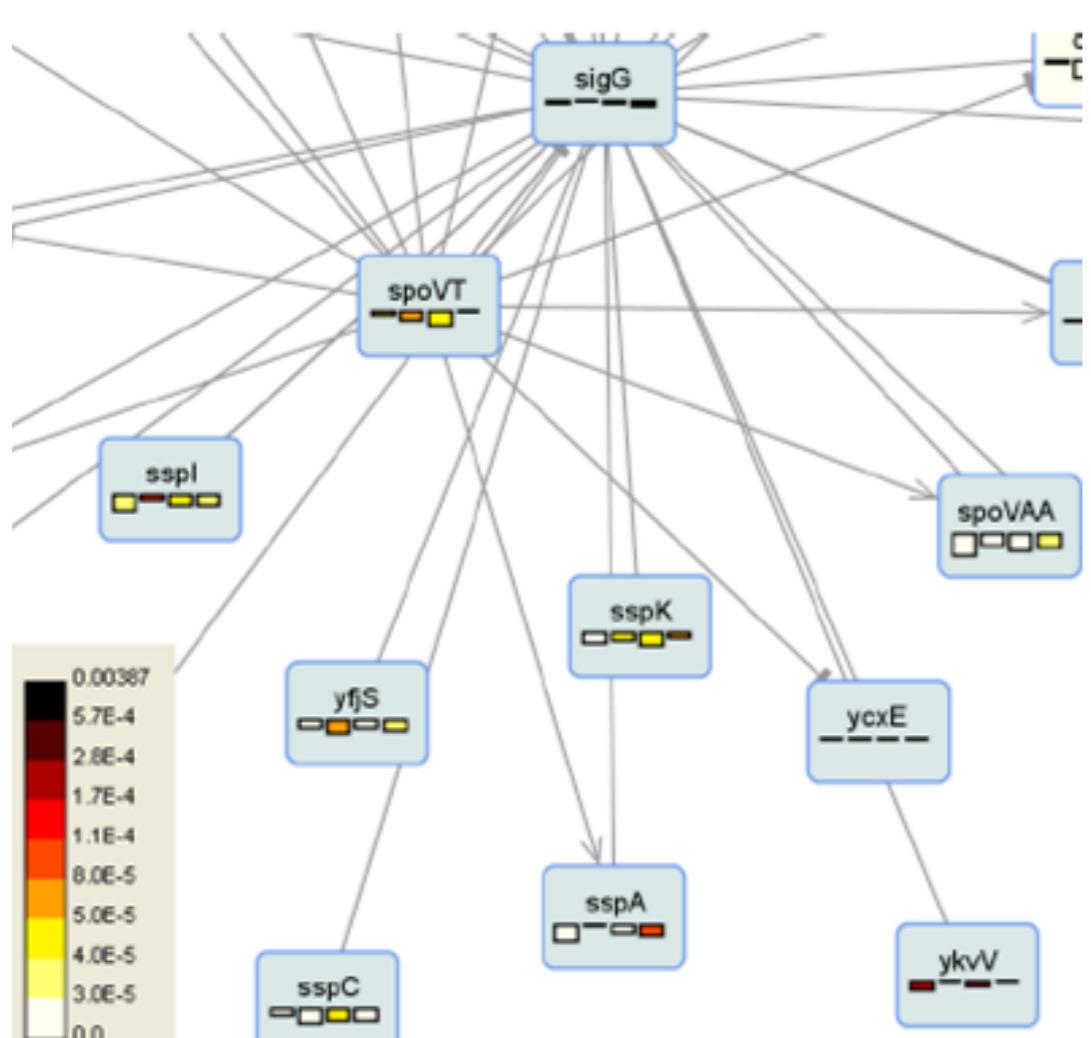


Why not animation?

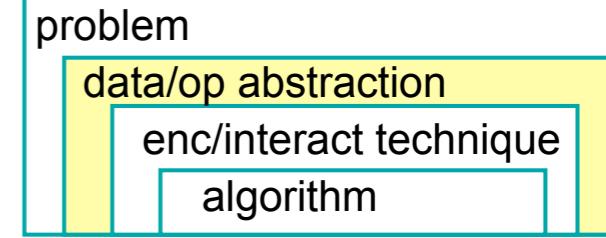
- limits of human visual memory
 - compared to side by side visual comparison
- Zooming versus multiple window interfaces: Cognitive costs of visual comparisons. Matthew Plumlee and Colin Ware. *ACM Trans. Computer-Human Interaction (ToCHI)*, 13(2):179-209, 2006.
- Animation: can it facilitate? Barbara Tversky, Julie Bauer Morrison, and Mireille Betrancourt. *International Journal of Human-Computer Studies*, 57(4):247-262, 2002.
- Effectiveness of Animation in Trend Visualization. George Robertson, Roland Fernandez, Danyel Fisher, Bongshin Lee, John Stasko. *IEEE Trans. Visualization and Computer Graphics* 14(6):1325-1332 (Proc. InfoVis 08), 2008.

Why not glyphs?

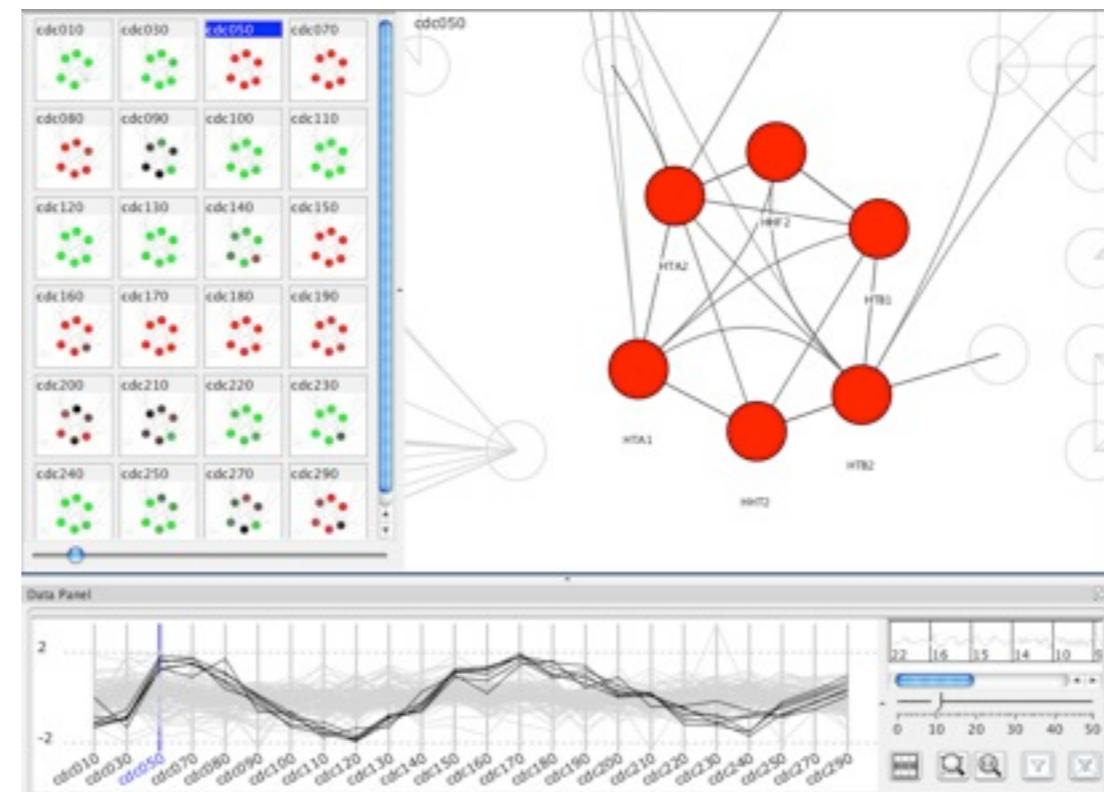
- embed multiple conditions as a chart inside node
- clearly visible when zoomed in
- but cannot see from global view
 - only one value shown in overview



Choice: Show measures and graph



- why not measurements alone?
 - data driven hypothesis: gene expression clusters indicate similar function in cell?
- clusters are often untrustworthy artifacts!
 - noisy data: different clustering alg. → different results
 - measured data alone potentially misleading
 - **show in context of graph model**



Contributions

- Cerebral
 - supports interactive exploration of multiple experimental conditions in graph context
 - provides familiar representation by using biological metadata to guide graph layout
- tool deployment
 - open source, Cytoscape plugin
 - used by target group of collaborators
 - 5 citations, showcased in <http://innatedb.ca>
 - many more independent adopters
 - 12+ bio lit citations with Cerebral diagrams so far

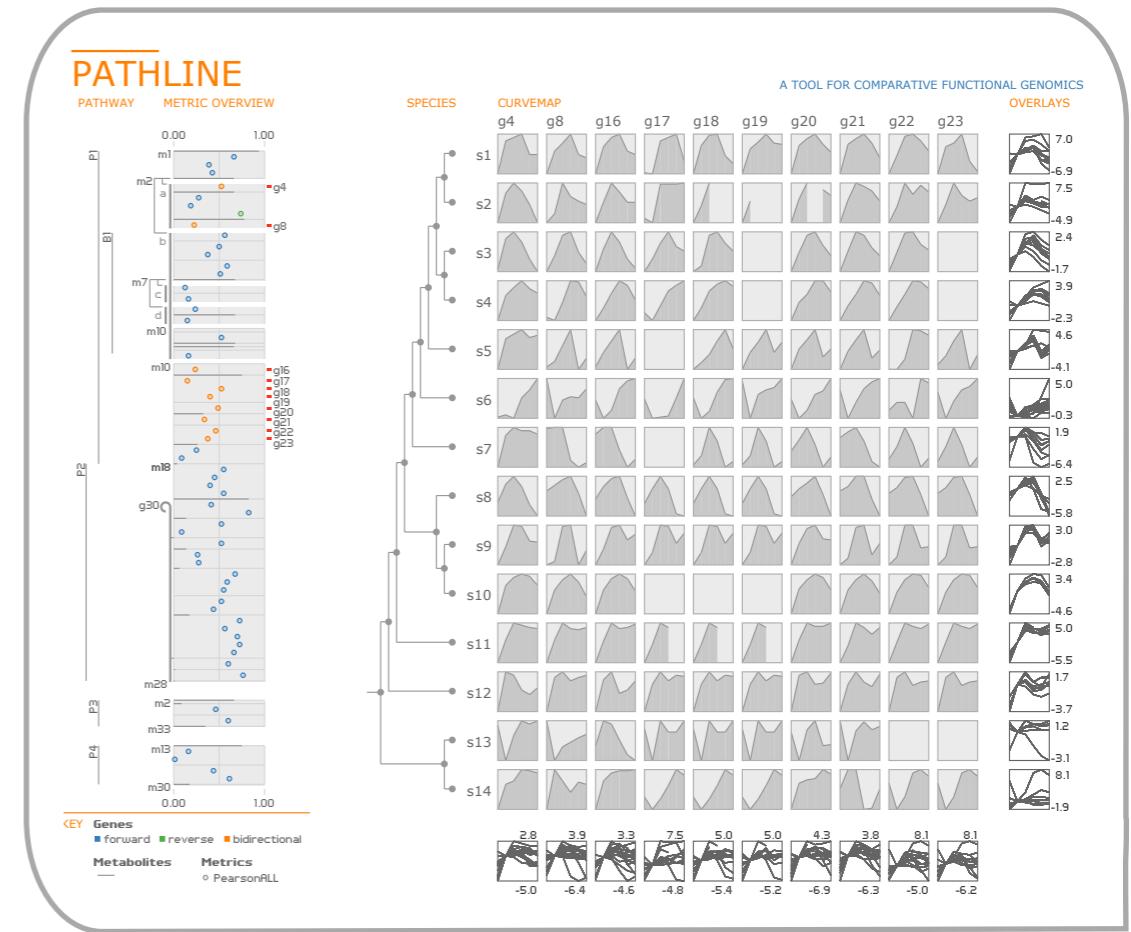
Pathline

A Tool for Comparative Functional Genomics Data

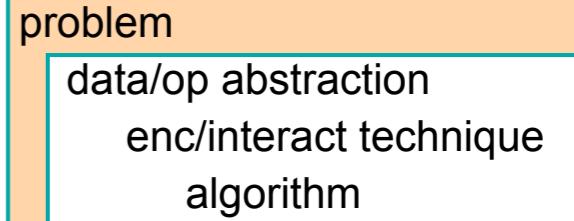
joint work with:

Miriah Meyer, Bang Wong, Mark Styczynski, Hanspeter Pfister

<http://www.pathline.org>



Pathline: A Tool for Comparative Functional Genomics
Meyer, Wong, Styczynski, Munzner, Pfister, IEEE/Eurographics EuroVis 2010.



problem: **functional genomics**

how do genes work together to perform different functions in a cell?

functional genomics data

gene expression

molecular pathways

gene expression is ...

... the measured level of how much a gene is on or off

... a single quantitative value

biologists measure it ...

... for many genes

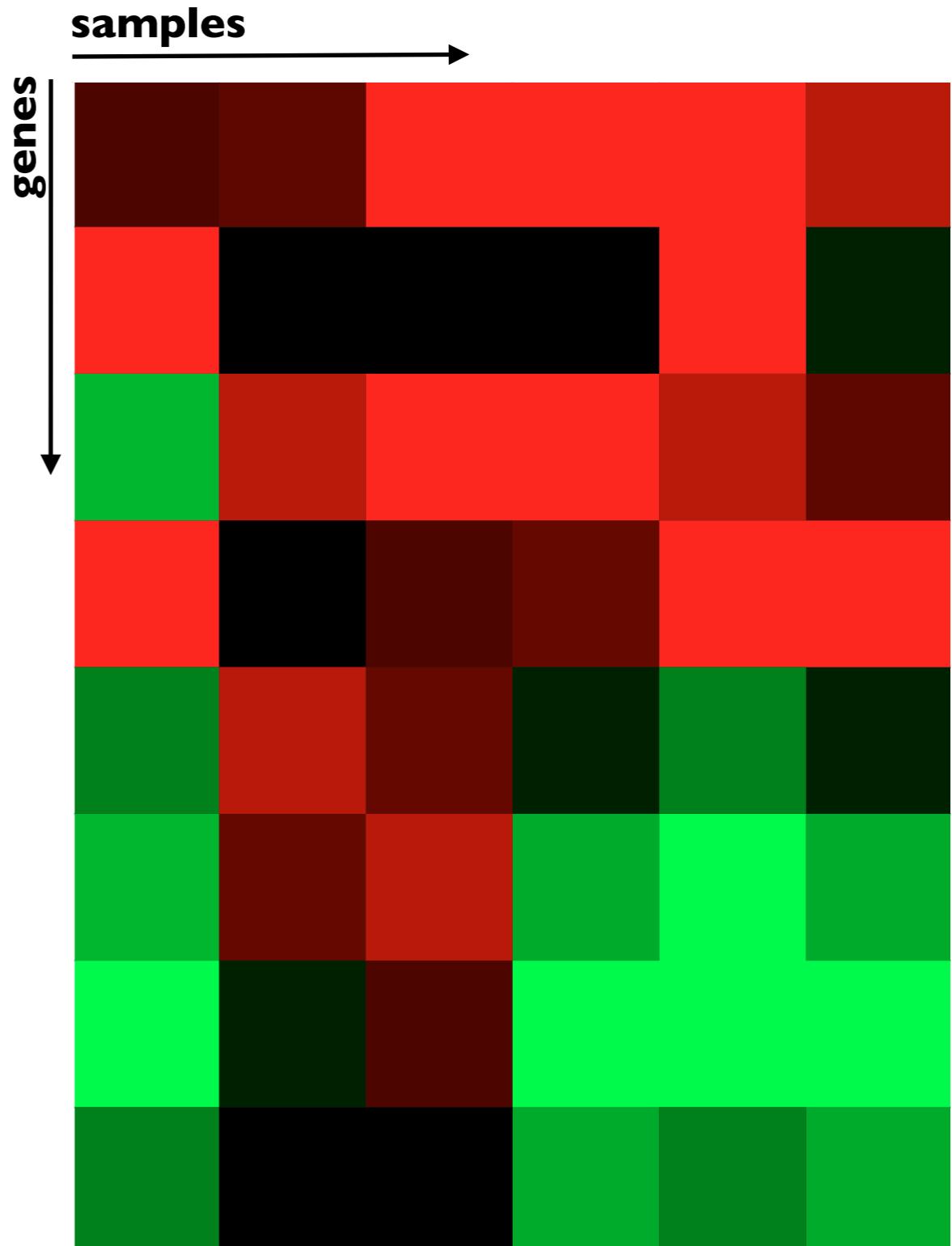
... in many samples (time points, tissue types, species)

visualized with heatmaps

[Wilkinson09] [Saldanha04] [Seo02] [Eisen98]

[Gehlenborg10] [Weinstein08]

encode value with color



gene expression is ...

... the measured level of how much a gene is on or off

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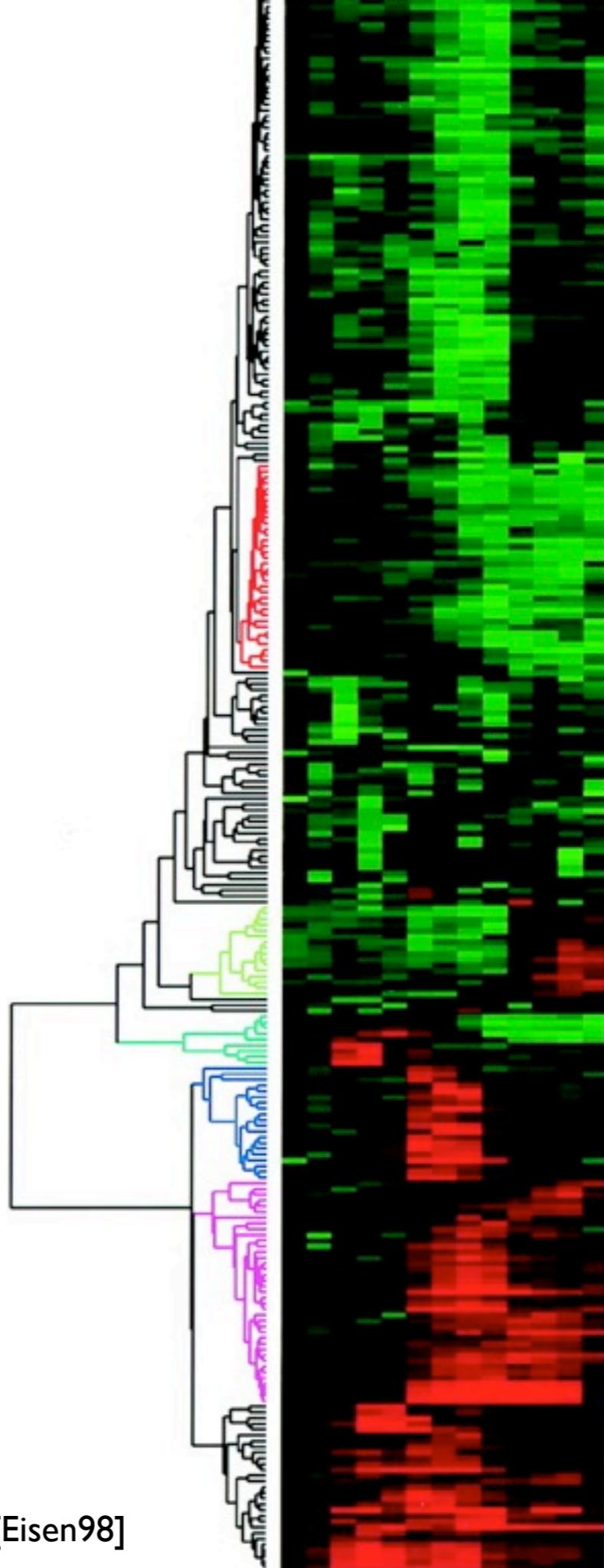
visualized with heatmaps

[Wilkinson09] [Saldanha04] [Seo02] [Eisen98]

[Gehlenborg10] [Weinstein08]

encode value with color

augmented with clustering



[Eisen98]

functional genomics data

gene expression

molecular pathways

the functioning of a cell is controlled by
many interrelated chemical reactions
performed by genes



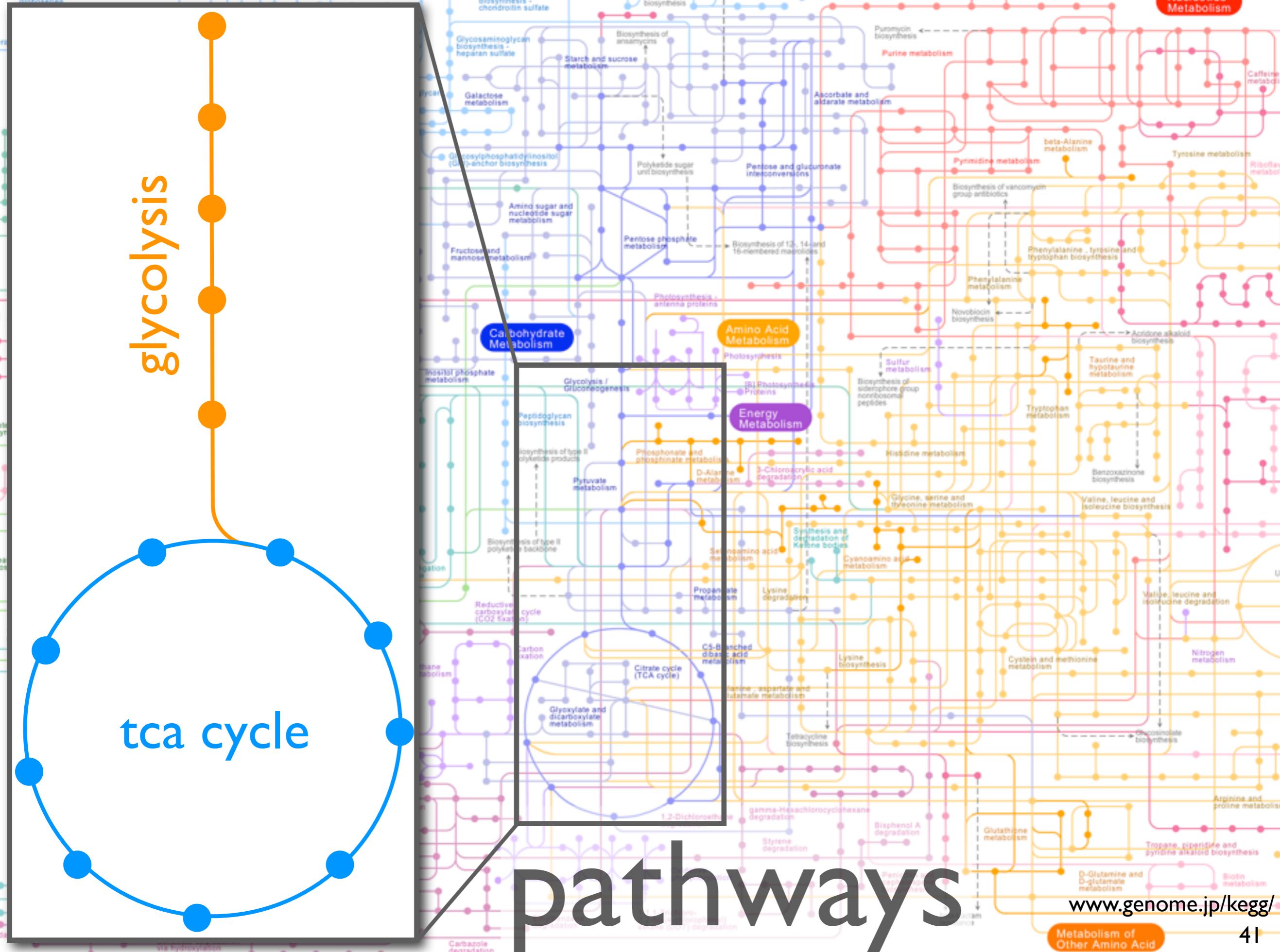
tca cycle

glycolysis

pathways

www.genome.jp/kegg/

41



functional genomics:

how do genes work together to perform different functions in a cell?

comparative functional genomics:

how do the gene interactions vary across different species?

collaborators: Regev Lab at the Broad Institute

biology: metabolism in yeast

data: multiple genes

- multiple time points

- multiple related species

- multiple pathways

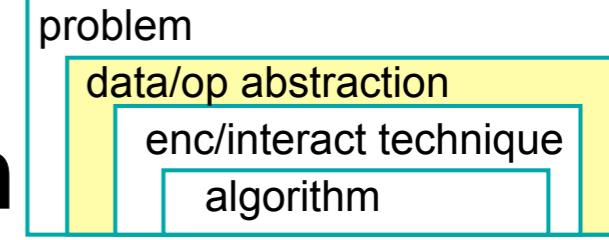
problem: *existing tools can only look at a subset of this data*

comparative functional genomics

how do the gene interactions vary across different species?

metabolic pathways

gene expression



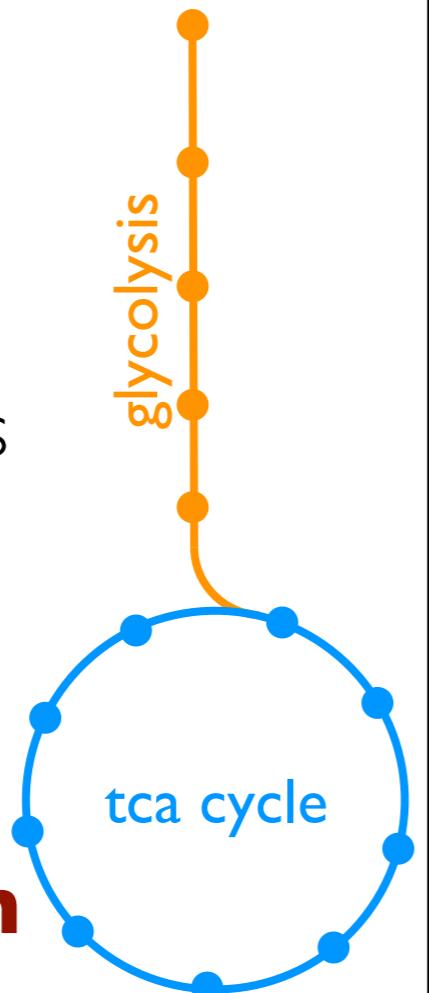
Data

similarity scores

phylogeny

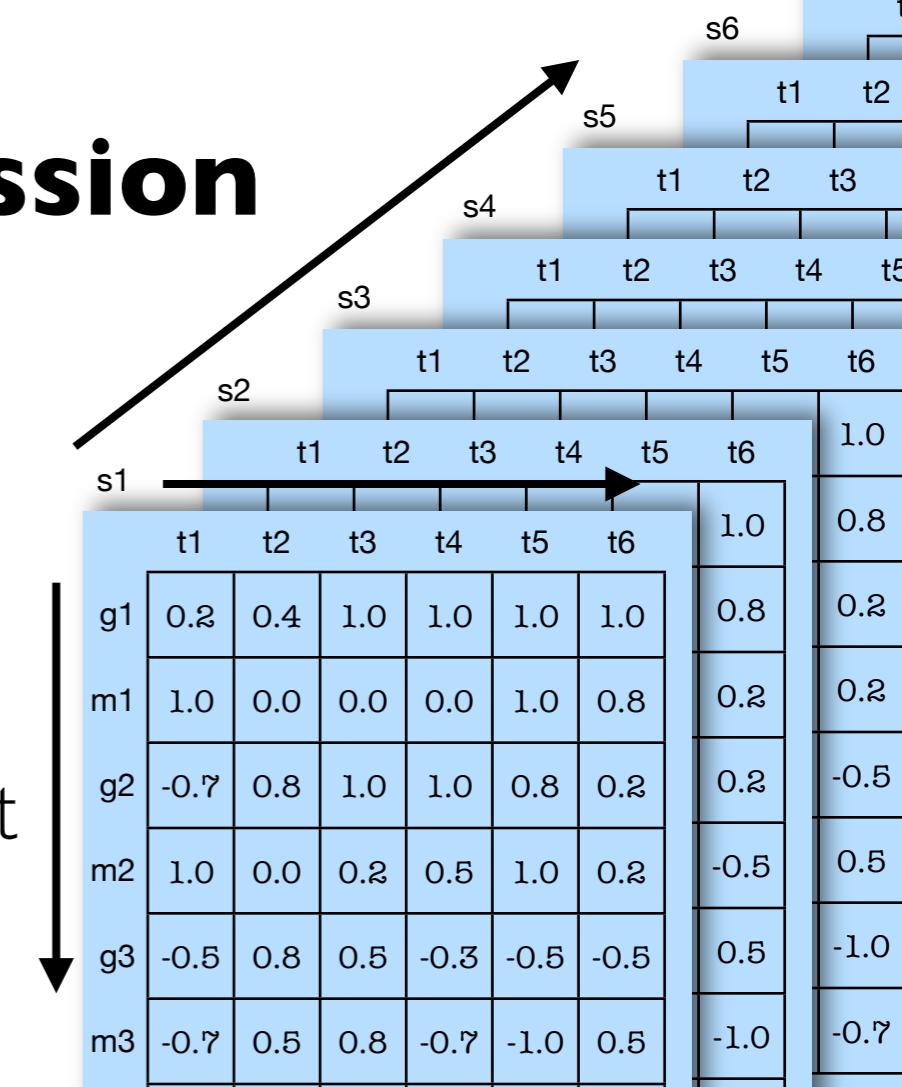
metabolic pathways

- 10 to 50 pathways of interest
- inputs/outputs called metabolites
- **directed graph**



gene expression

- 6000 genes and 140 metabolites
 - 6 time points
 - 14 species of yeast
- **3D table**

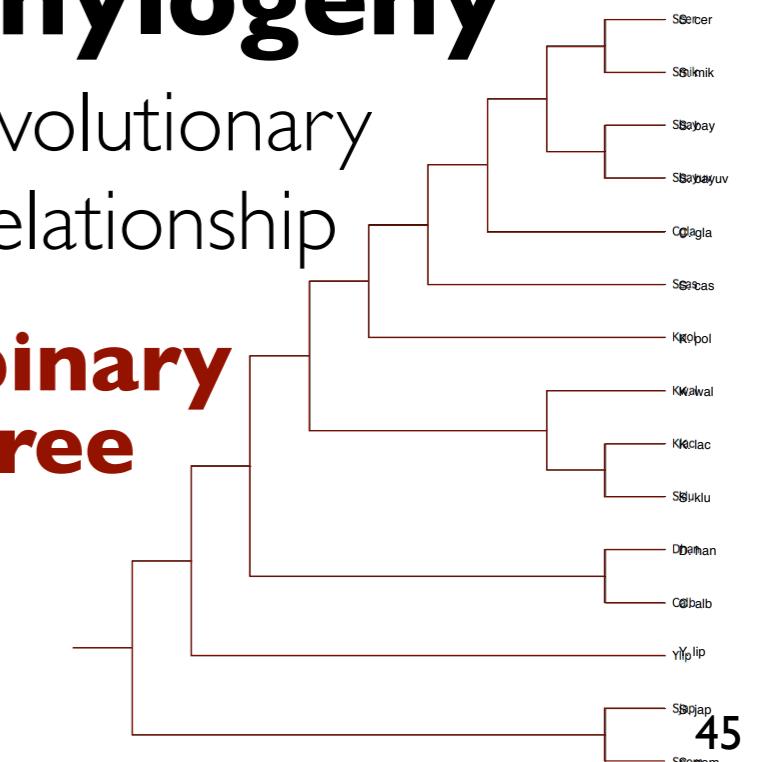


similarity scores

phylogeny

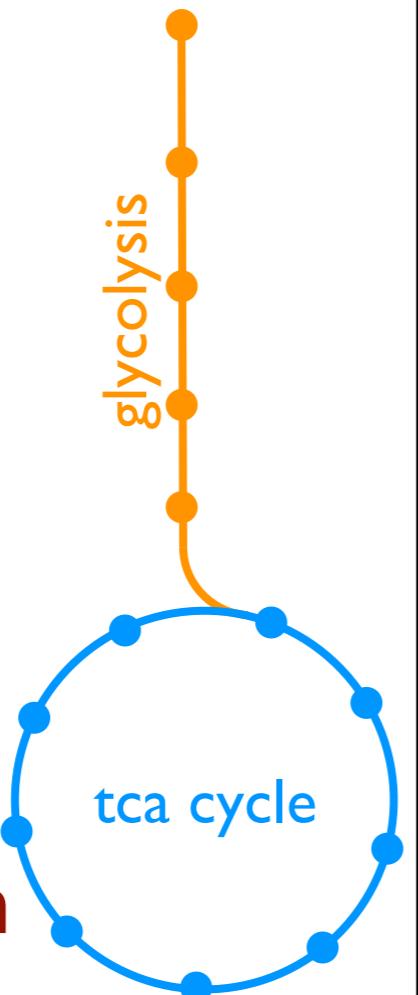
- evolutionary relationship

• **binary tree**



metabolic pathways

- 10 to 50 pathways of interest
- inputs/outputs called metabolites
- **directed graph**



gene expression

- 6000 genes and 140 metabolites
- 6 time points
- 14 species of yeast
- **3D table**

		t					
		t1	t2	t3	t4	t5	t6
s	s6	1.0					
		0.8					
s	s5	0.2					
		-0.5					
s	s4	0.5					
		-1.0					
s	s3	-1.0					
		0.2					
s	s2	0.2					
		-0.5					
s	s1	0.5					
		-1.0					
		t1	t2	t3	t4	t5	t6
g	g1	0.2	0.4	1.0	1.0	1.0	1.0
m	m1	1.0	0.0	0.0	0.0	1.0	0.8
m	g2	-0.7	0.8	1.0	1.0	0.8	0.2
m	m2	1.0	0.0	0.2	0.5	1.0	0.2
m	g3	-0.5	0.8	0.5	-0.3	-0.5	-0.5
m	m3	-0.7	0.5	0.8	-0.7	-1.0	0.5

similarity scores

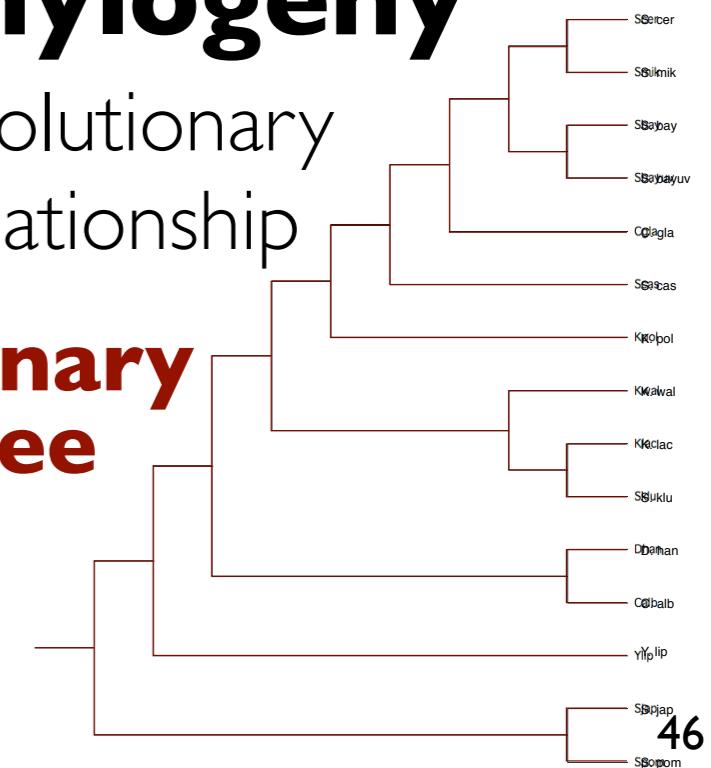
$$\text{aggregate} \left(s_1, s_2, s_3, \dots, \right) = 0.83$$

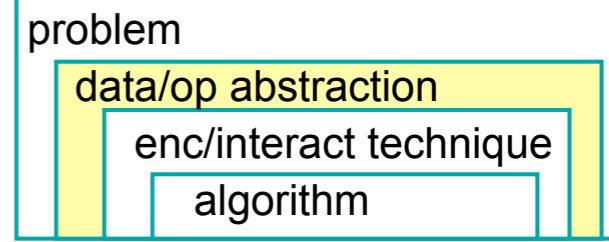
- aggregate time series for a gene/metabolite over species
- similarity of expression across species
- aggregate: Pearson, Spearman, others
- **quantitative value**

phylogeny

- evolutionary relationship

• binary tree





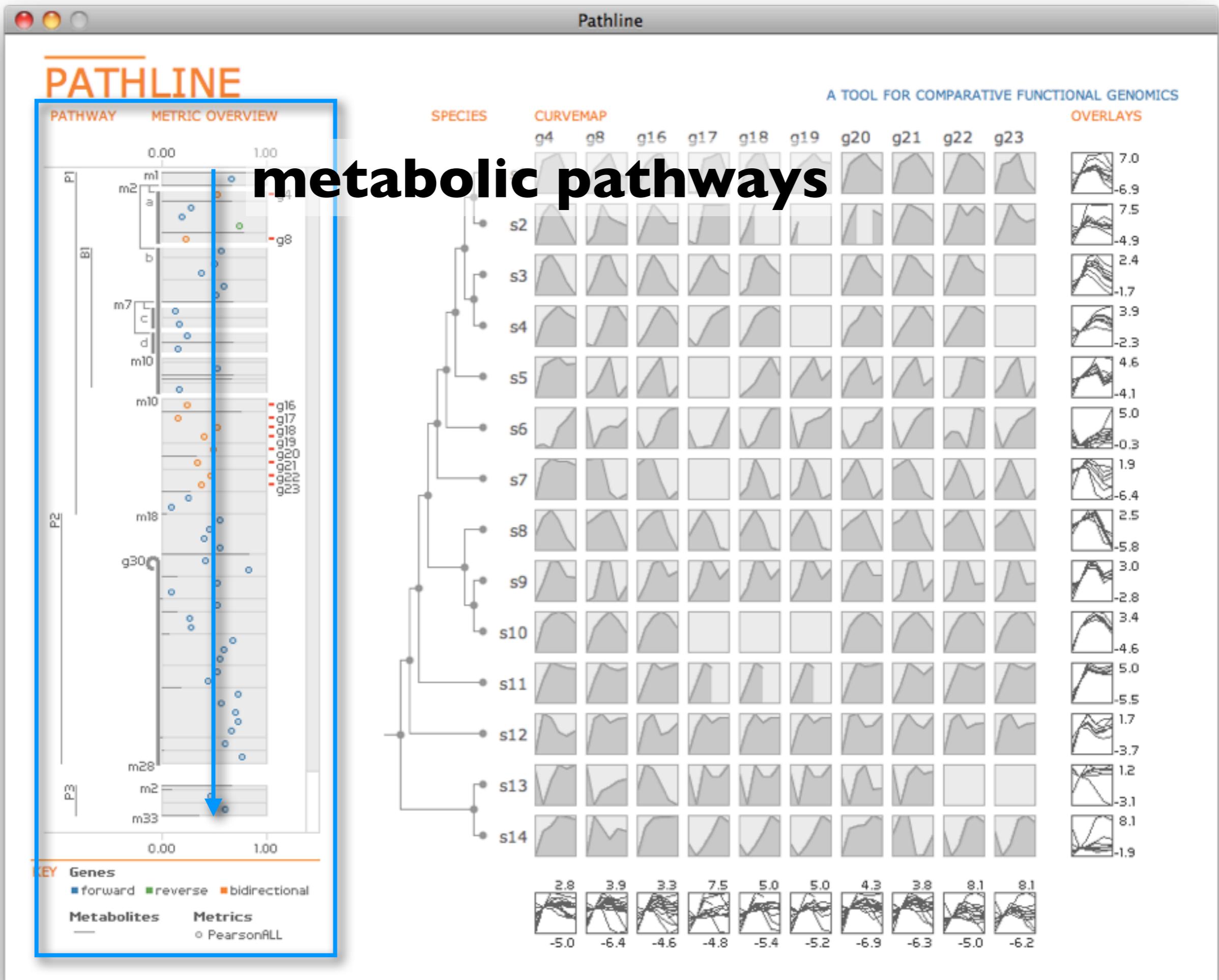
Tasks

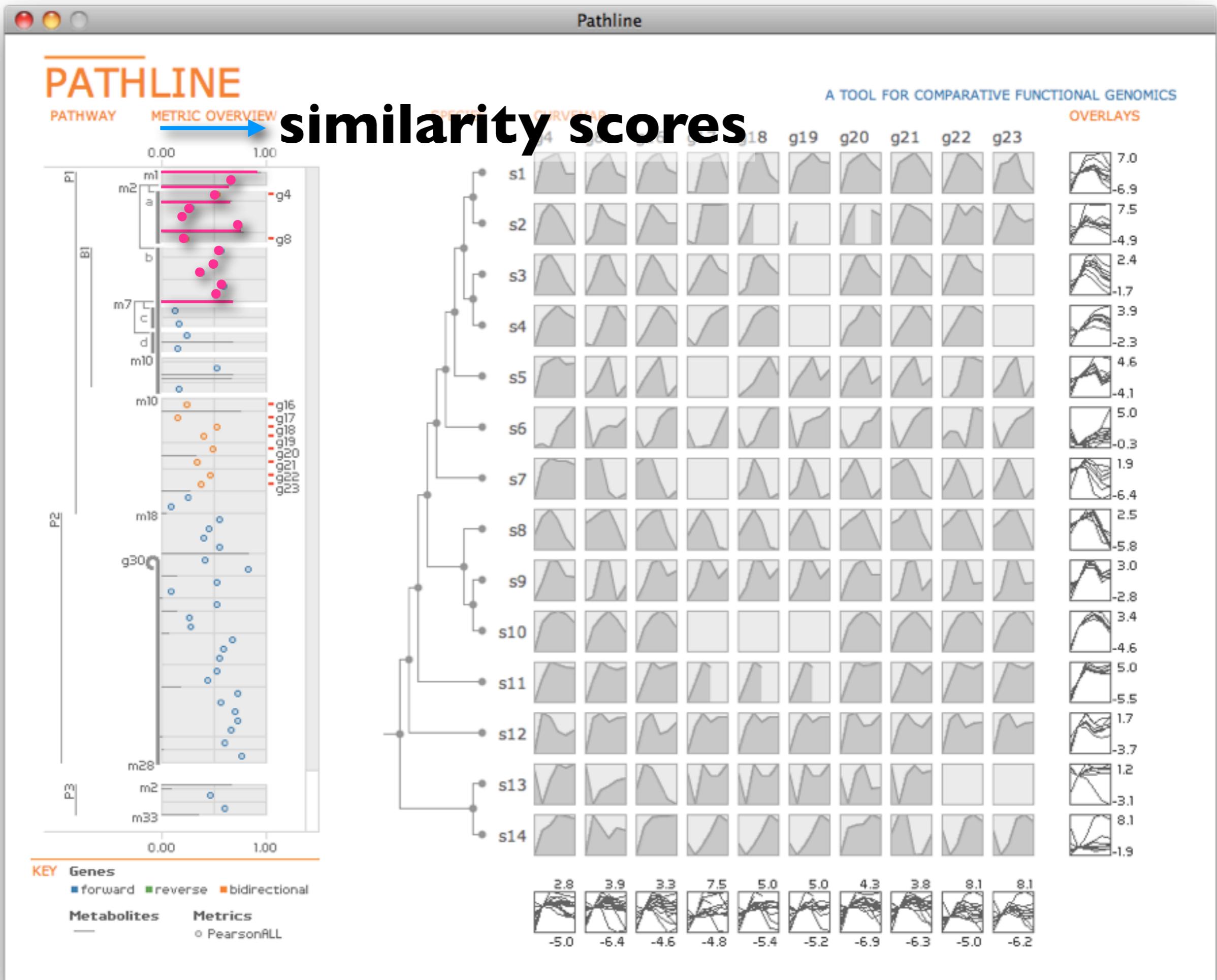
study expression data as a time series

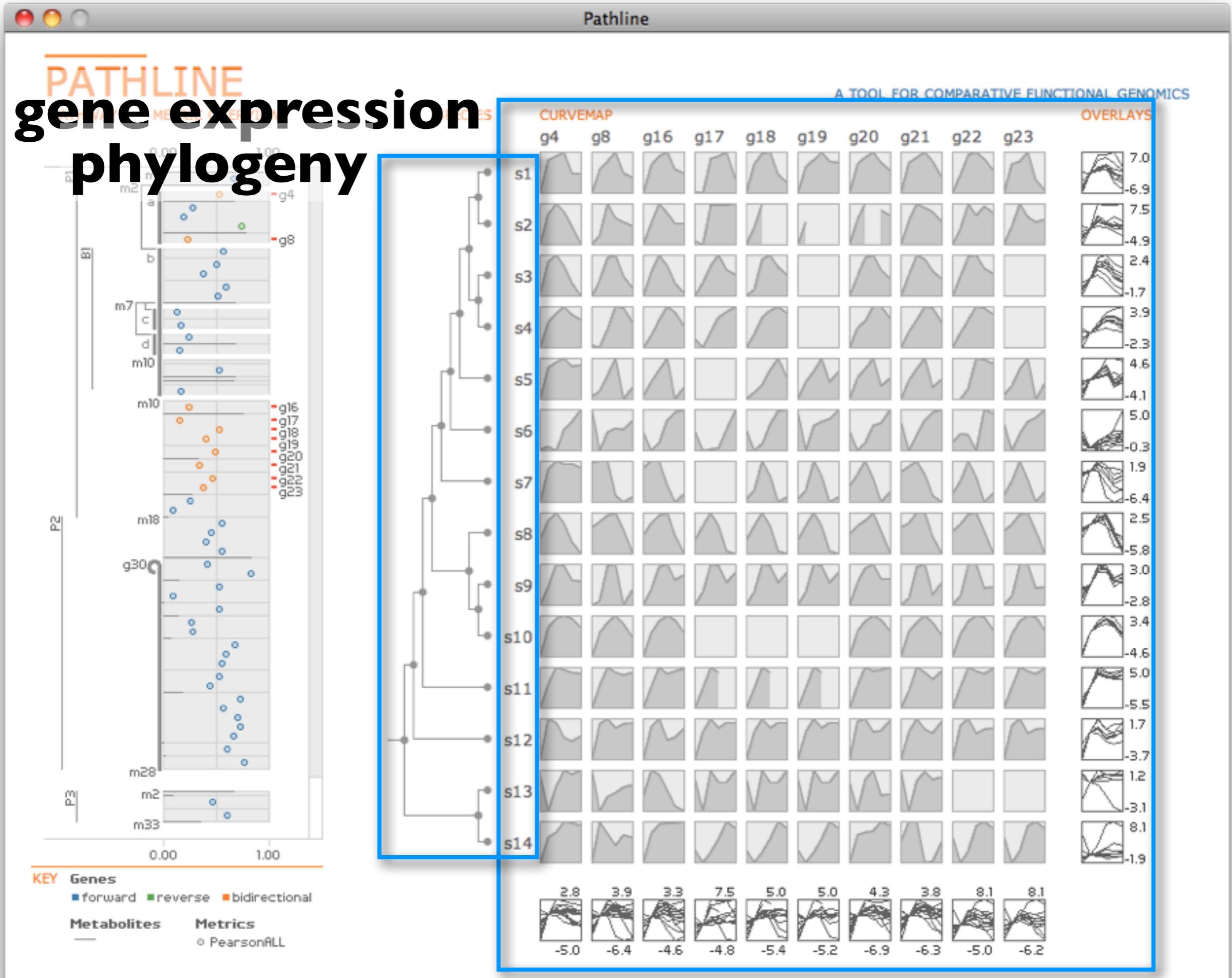
compare a limited number of time series

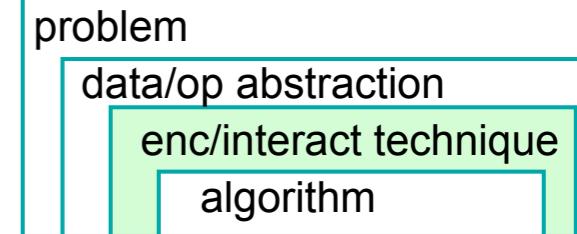
compare similarity scores along a pathway(s)

comparison of multiple similarity scores





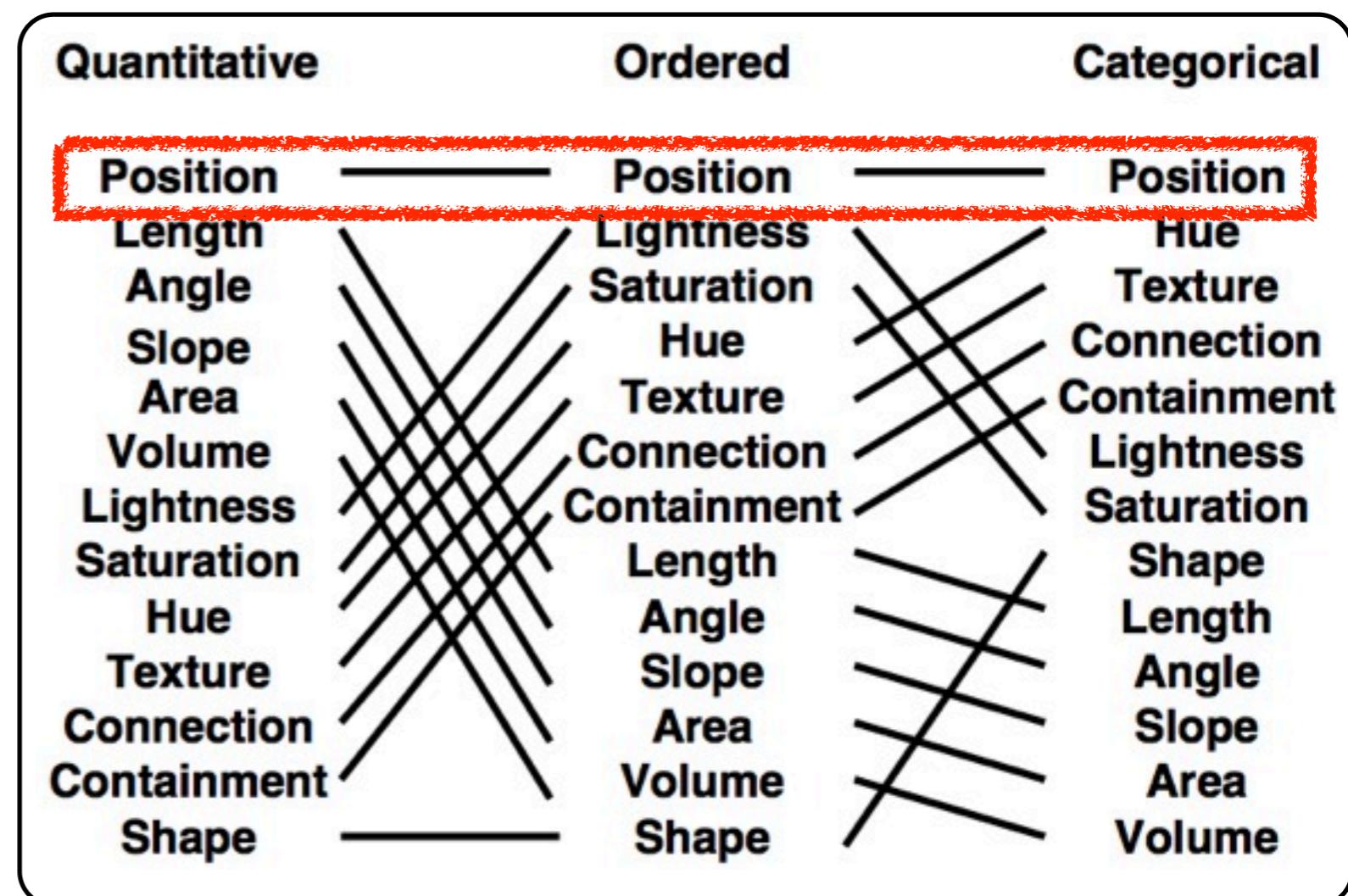




Principle: spatial position is visual channel most accurately perceived for all data types

LES VARIABLES DE L'IMAGE			
	POINTS	LIGNES	ZONES
XY 2 DIMENSIONS DU PLAN	x x x	/ \ 2	14 15 9 2 16 21 1 21 15 1 14 15 1 1 2 9
Z TAILLE			
VALEUR			

LES VARIABLES DE SÉPARATION DES IMAGES			
	GRAIN	COULEUR	ORIENTATION
GRAIN			
COULEUR			
ORIENTATION			
FORME			

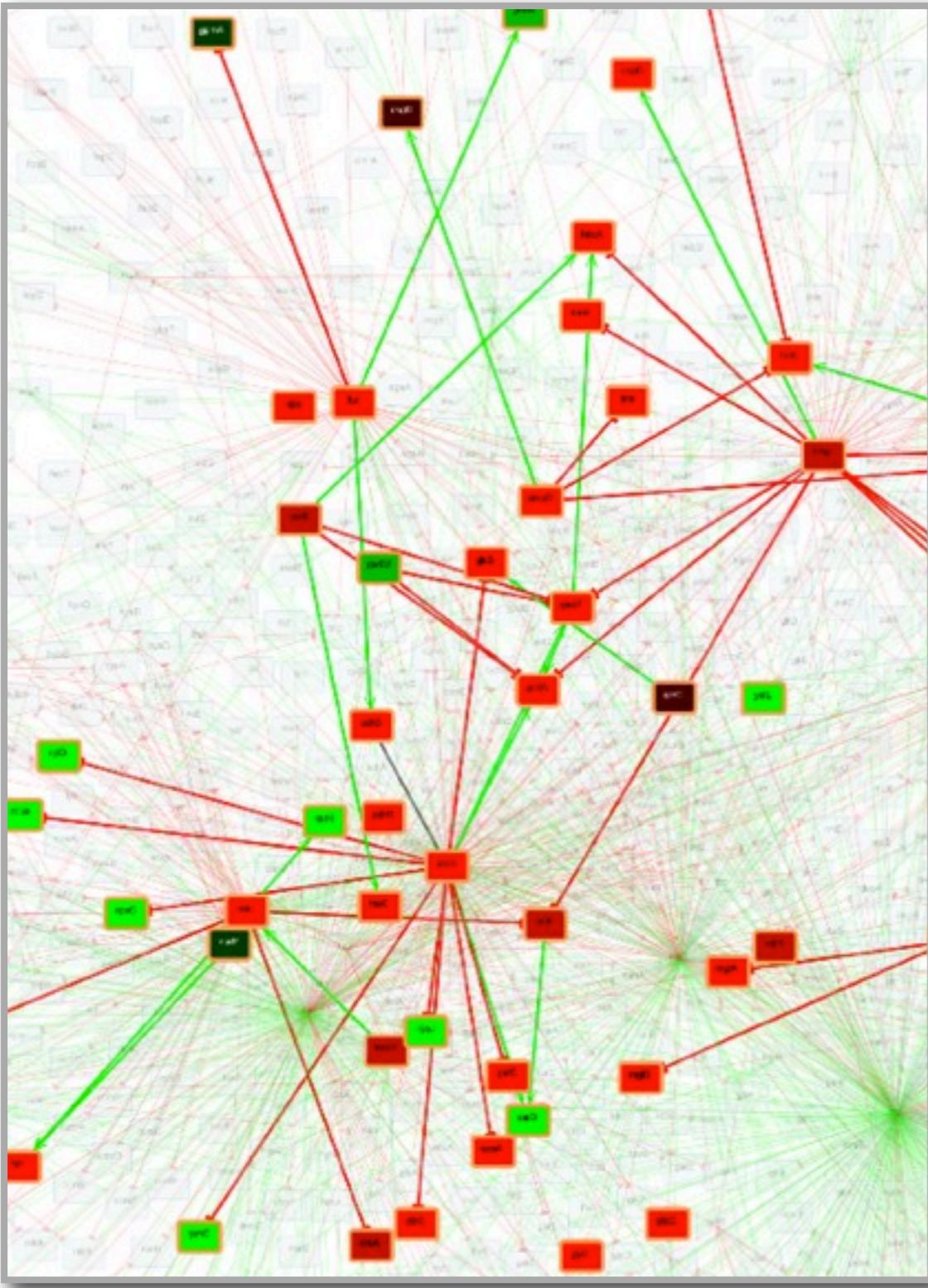


Semiology of Graphics
Bertin, 1967

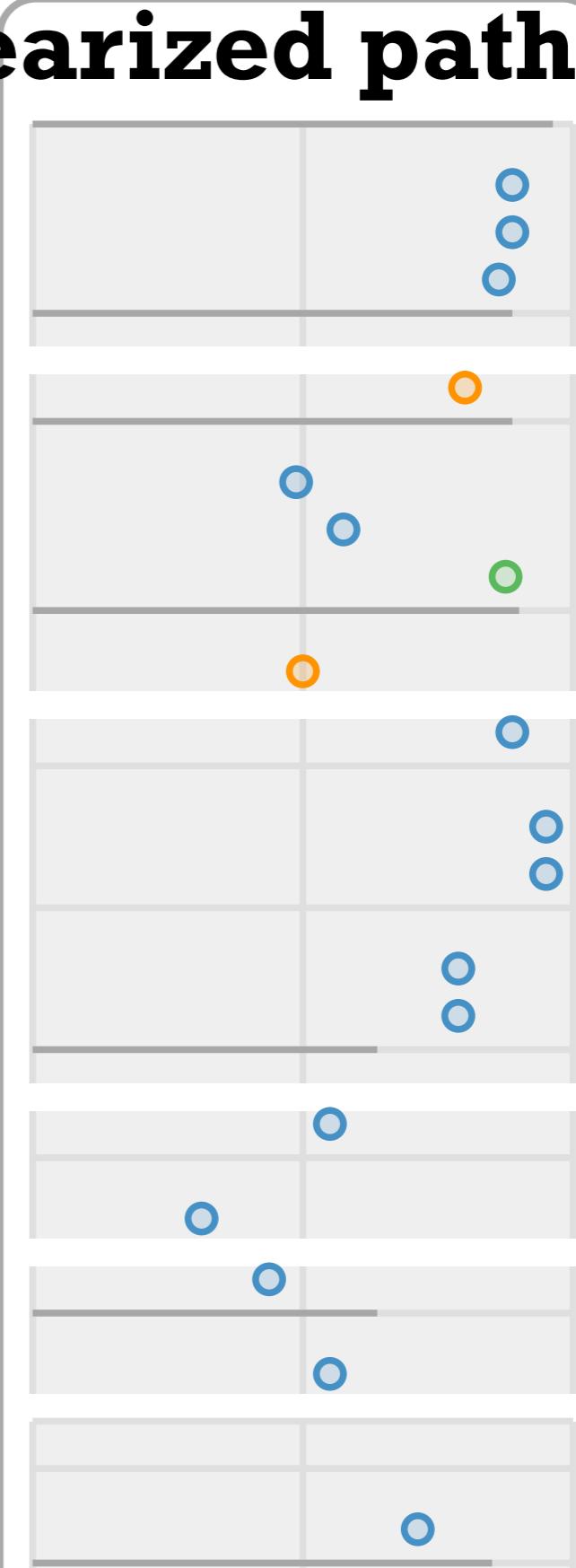
Automating the Design of Graphical Presentations of Relational Information
Mackinlay, 1986

Encode quantitative values with spatial position

topological layout

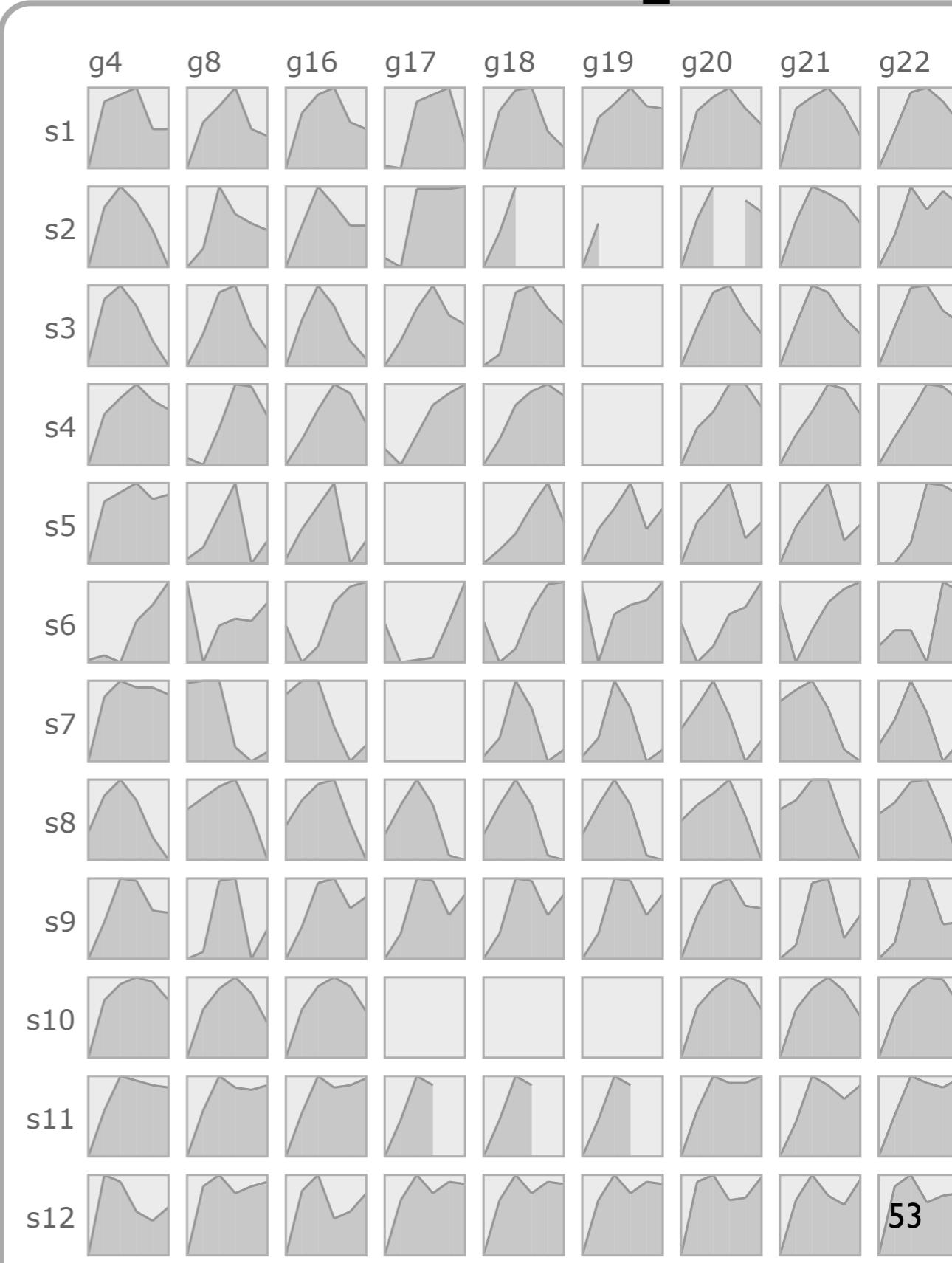
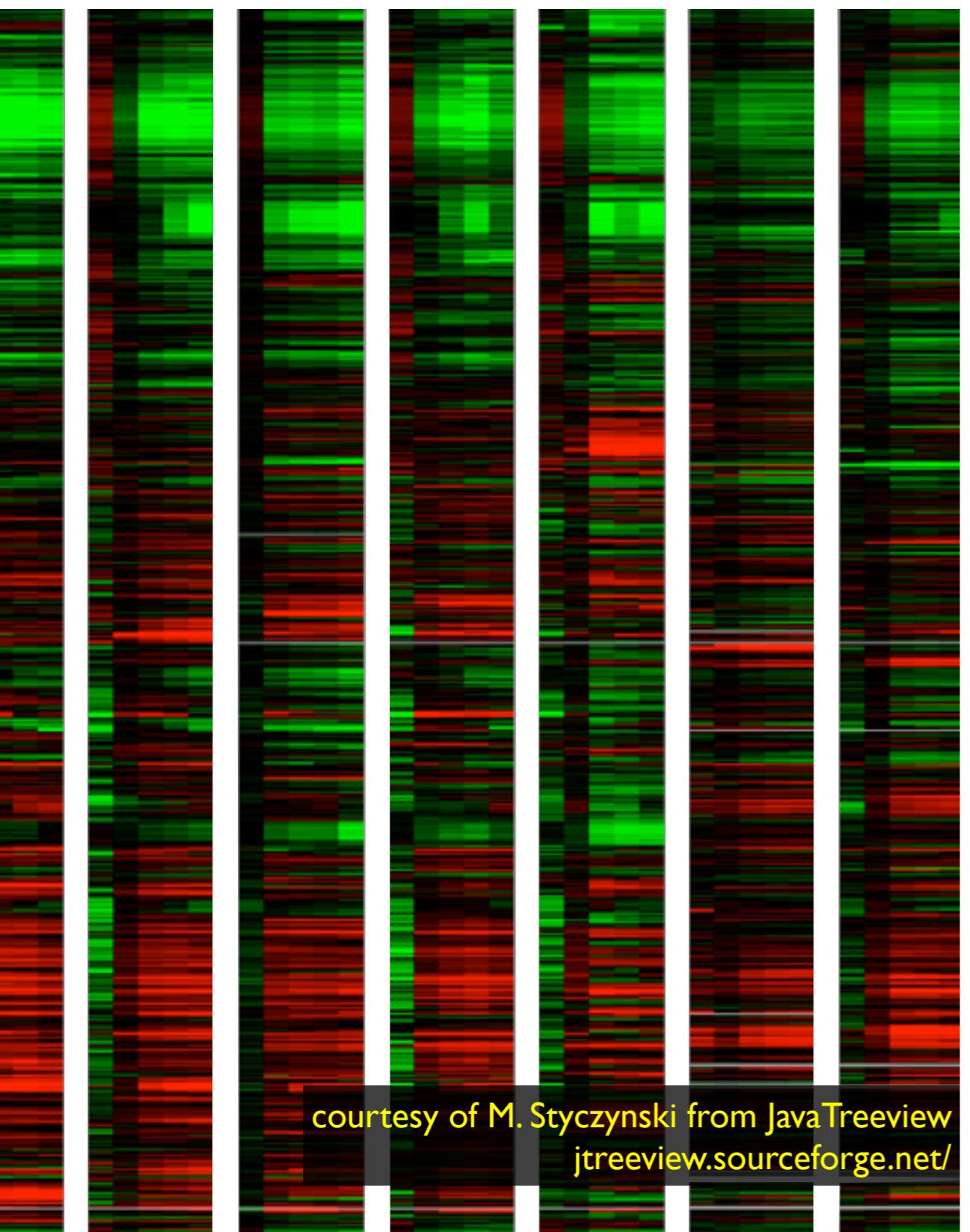


linearized pathway



Encode quantitative values with spatial position heatmap

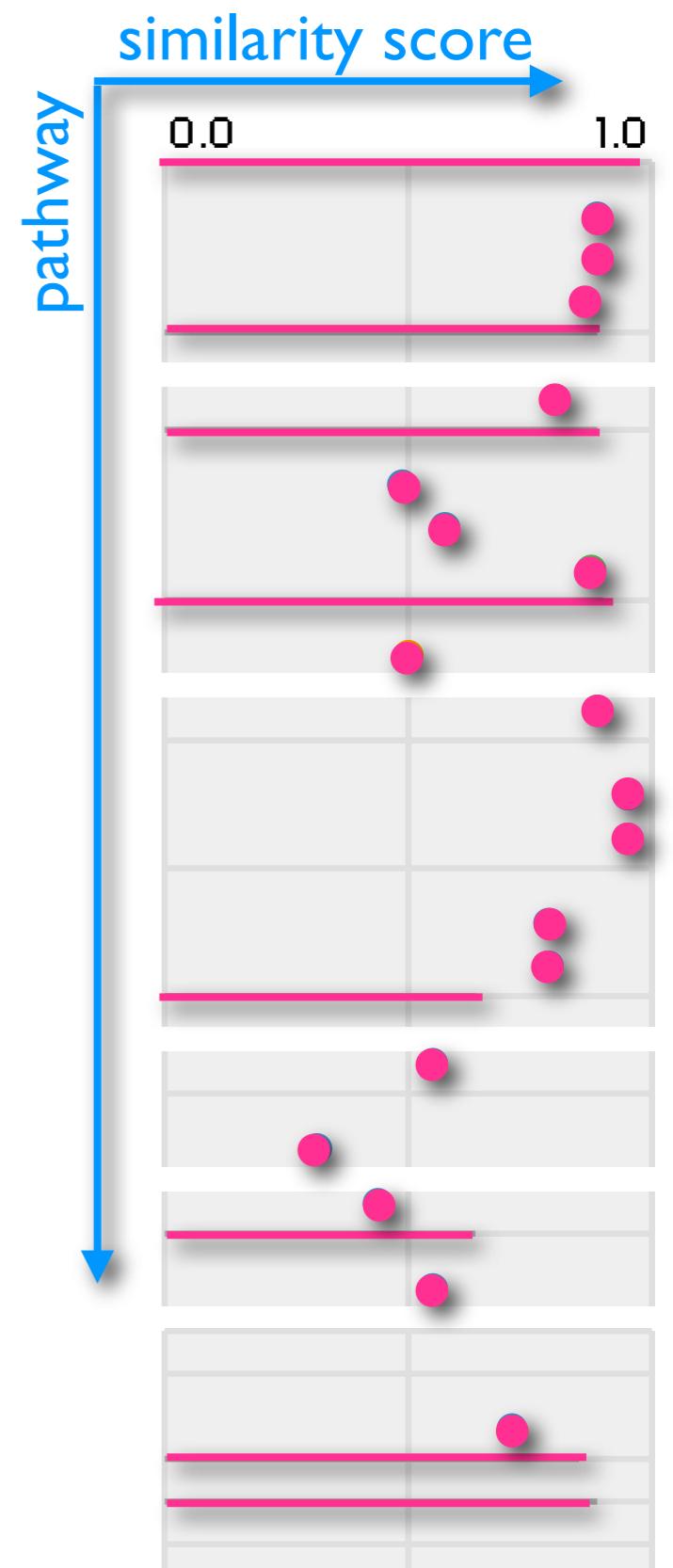
curvemap



Linearized pathway

common axes to compare similarity scores

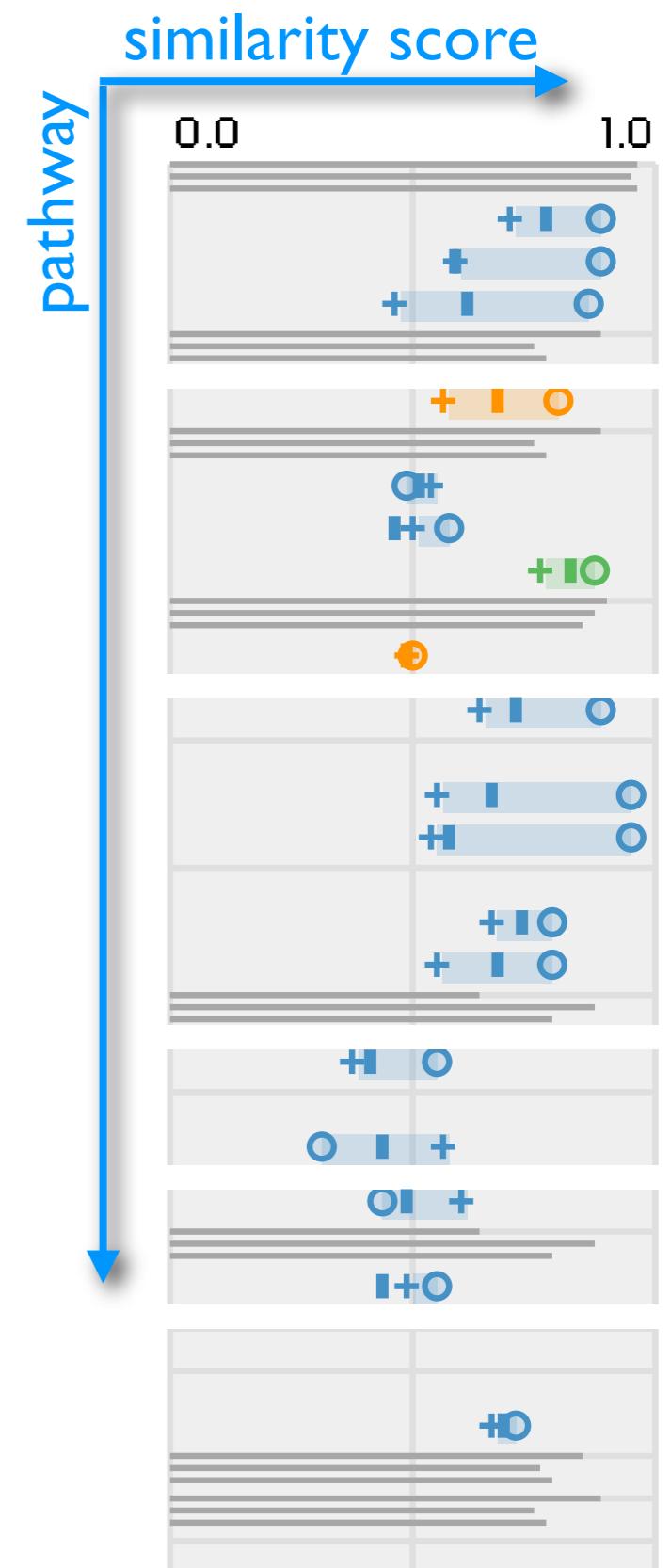
- bars and circles
 - visual layer for selective attention
 - color-code gene direction



Linearized pathway

common axes to compare similarity scores

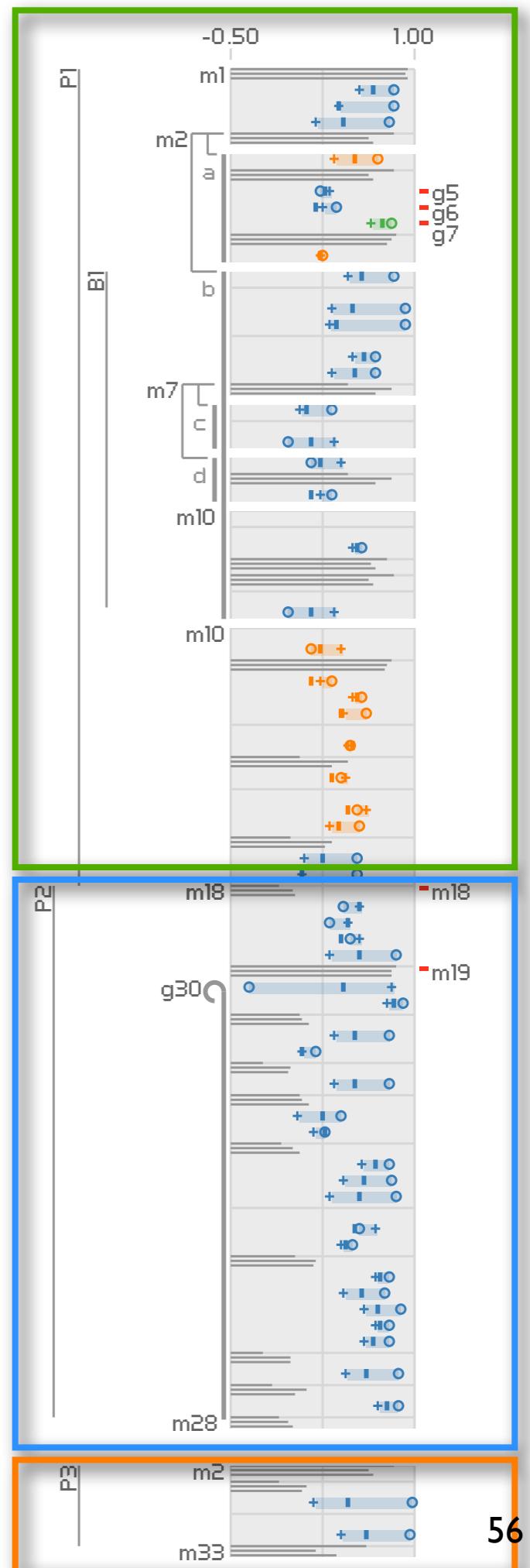
- bars and circles
 - visual layer for selective attention
 - color-code gene direction
- multiple similarity scores



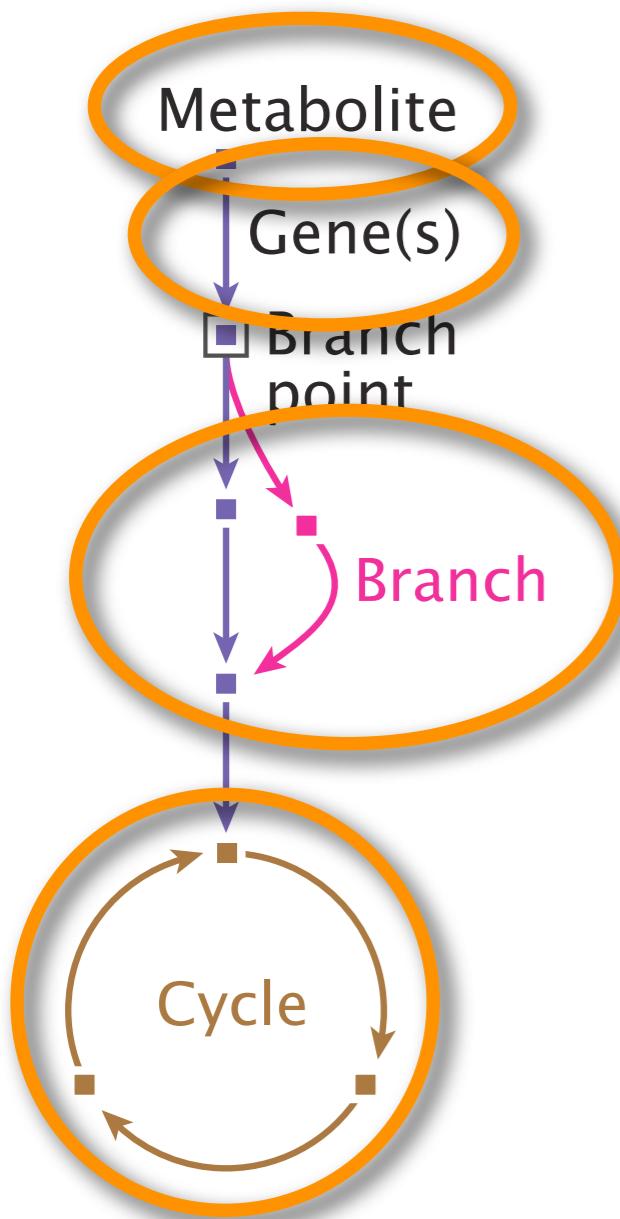
Linearized pathway

common axes to compare similarity scores

- bars and circles
 - visual layer for selective attention
 - color-code gene direction
- multiple similarity scores
- multiple pathways



Pathway to ordered list of nodes



**unroll
and cut**

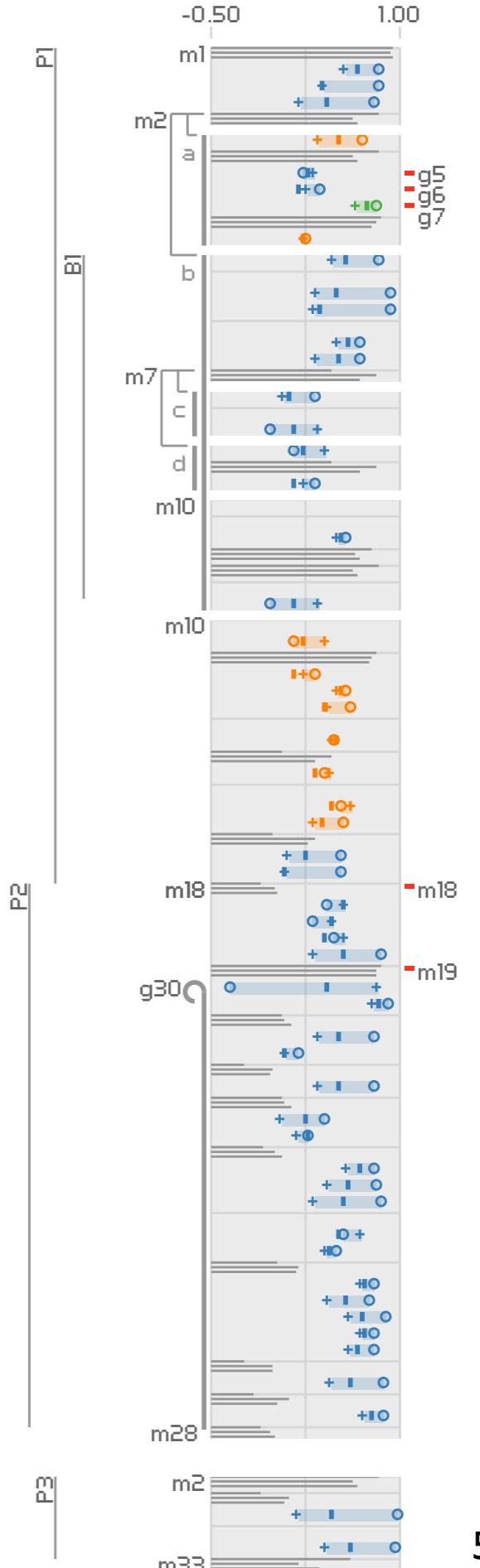
reinsert

**shared
coordinate
frame and
stylized marks**

Linearized pathway

putting it together . . .

- use spatial position for similarity scores
- topology is secondary



Curvemap

alternative to heatmaps

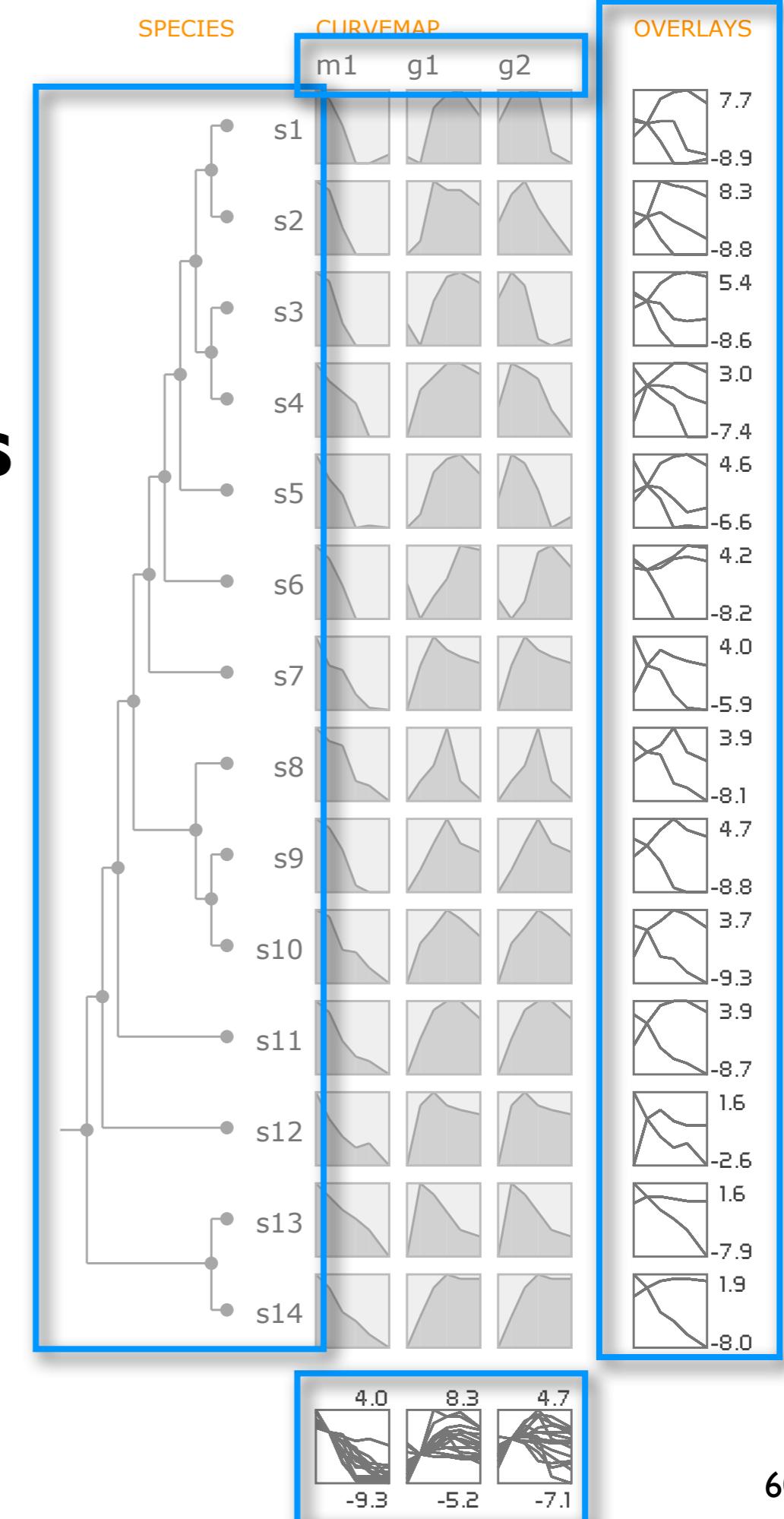
- base visual unit is a curve
- filled, framed line charts to enhance shape perception

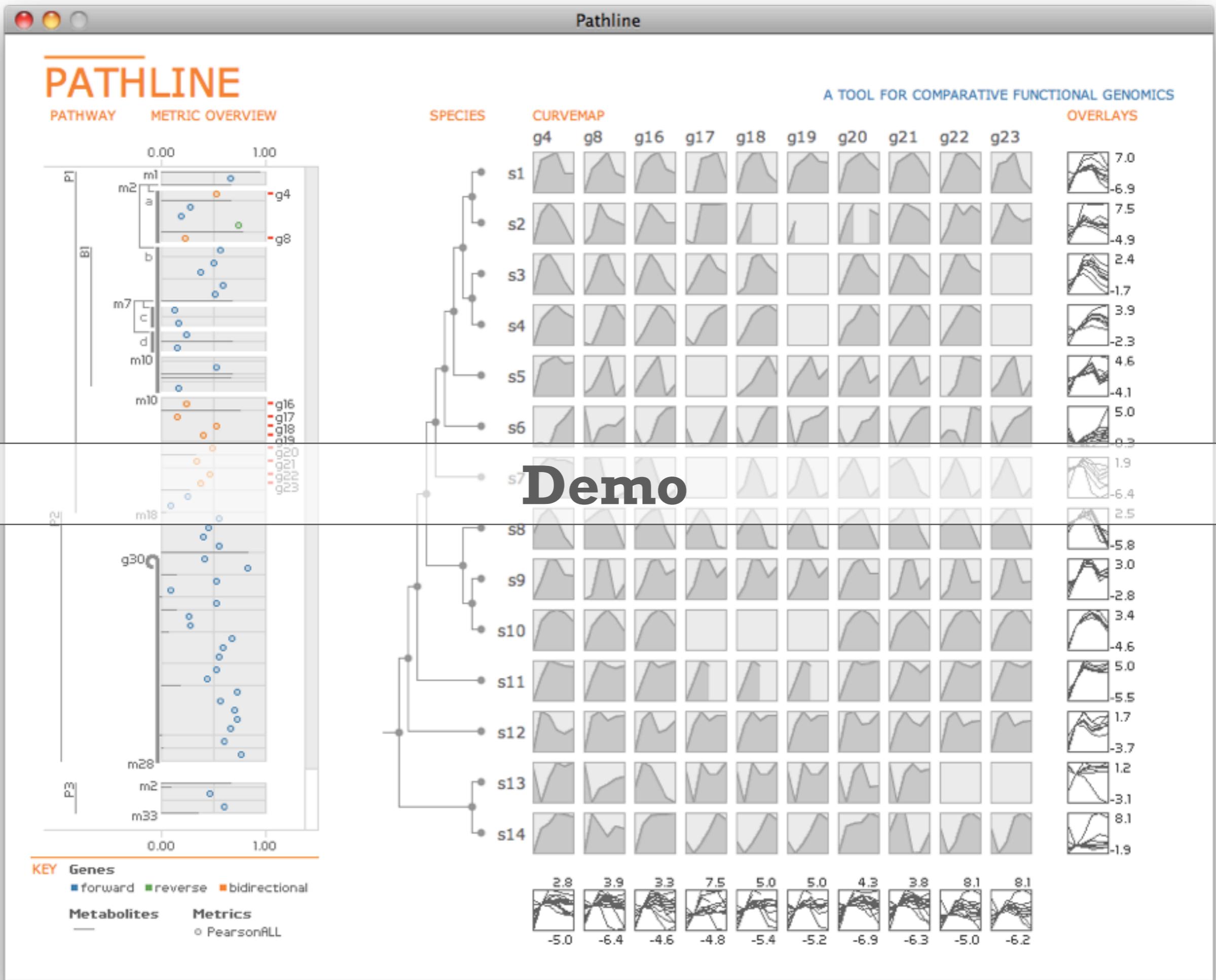


Curvemap

alternative to heatmaps

- base visual unit is a curve
- filled, framed line charts to enhance shape perception
- rows are species
- columns are genes/metabolites
- overlays to enhance trends





Contributions

- Pathline
 - multiple genes, time points, species, and pathways
- new visual encoding techniques based on infovis principles and biology needs
 - linearized pathway representation
 - curvemap
- tool deployment
 - open source
 - used daily by several collaborators

Principle: use validation methods tuned to level

- is target problem really solved?
 - what have we learned about tradeoffs in design space?

validate: observe target users

validate: justify design wrt alternatives

validate: measure system time

validate: lab study, qualitative results analysis

validate: observe real usage in field

A Nested Model for Visualization Design and Validation.

Munzner. IEEE InfoVis 2009.

More information

- principles in more depth: vis intro book chapter
<http://www.cs.ubc.ca/~tmm/papers.html#akpchapter>
- papers, talks, videos, courses
<http://www.cs.ubc.ca/~tmm>
- this talk
<http://www.cs.ubc.ca/~tmm/talks.html#hveill>