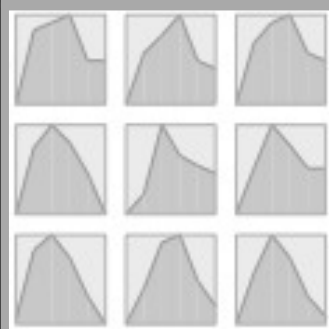
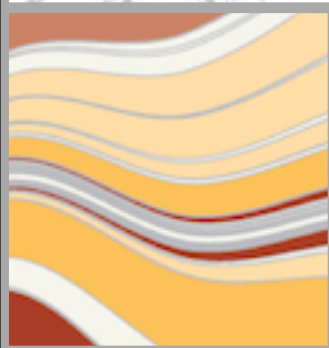
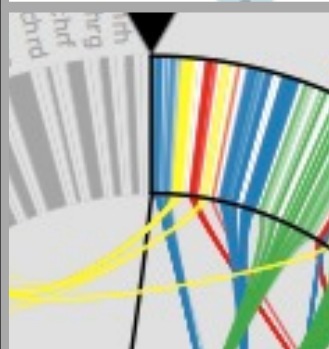
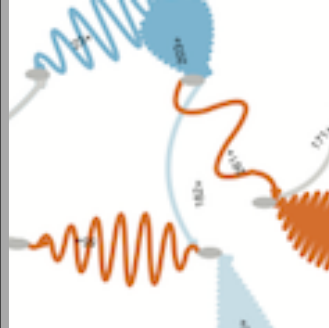


cs6964 | January 17 2012

PROCESS

Miriah Meyer
University of Utah



ADMINISTRIVIA

LAST TIME

Tufte's integrity principles

Clear, detailed, and thorough labeling should be used to defeat graphical distortion and ambiguity.

The representation of numbers, as physically measured on the surface of the graphic itself, should be directly proportional to the numerical quantities represented. *aka The Lie Factor*

Show data variation, not design variation.

Tufte's design principles

- maximize the data-ink ratio**
- avoid chart junk (*sometimes*)**
- use multifunctioning elements**
- layer information**
- maximize the data density**
 - shrink the graphics
 - maximize the amount of data shown (*sometimes*)

Williams's design principles

Contrast

Repetition

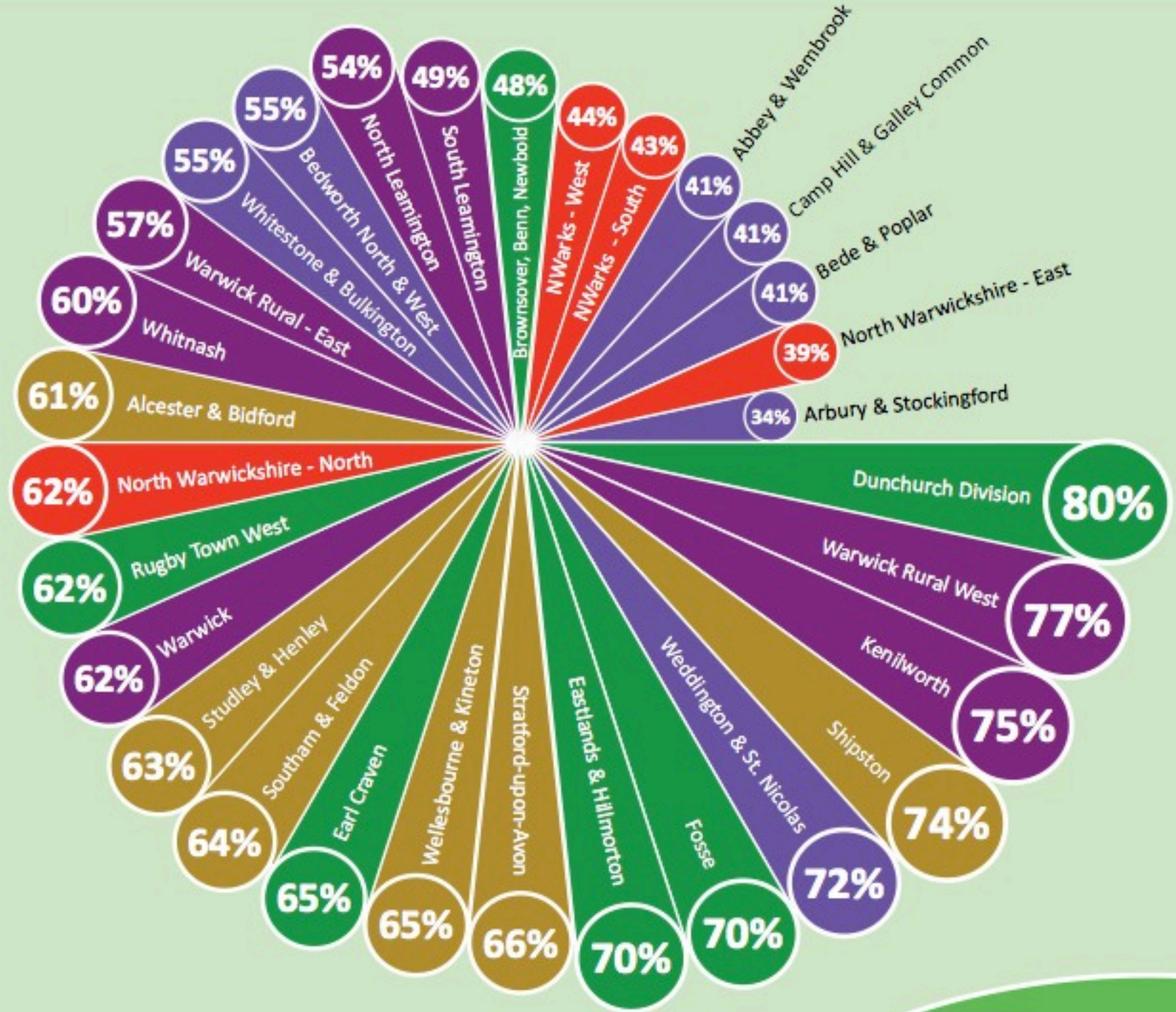
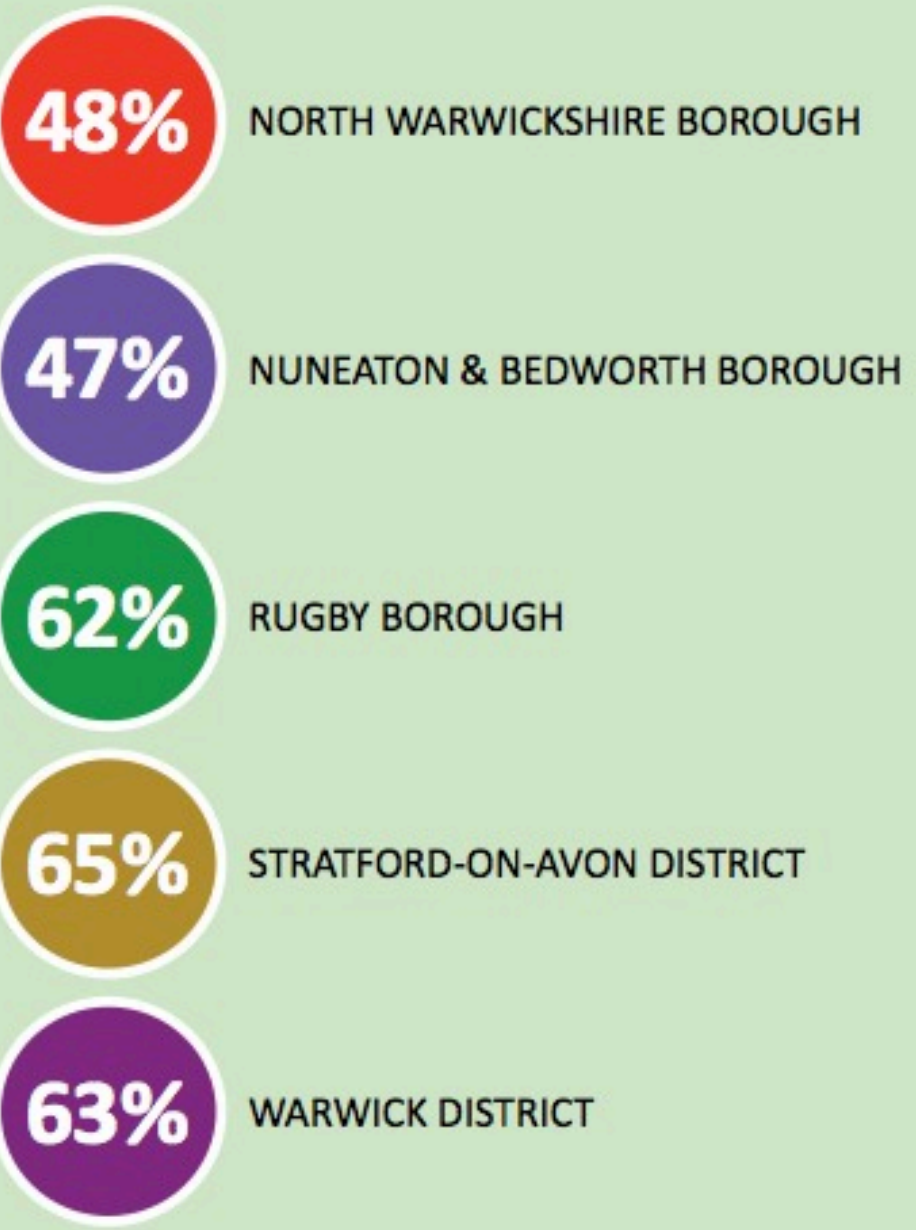
Alignment

Proximity

critique & redesign exercise . . .

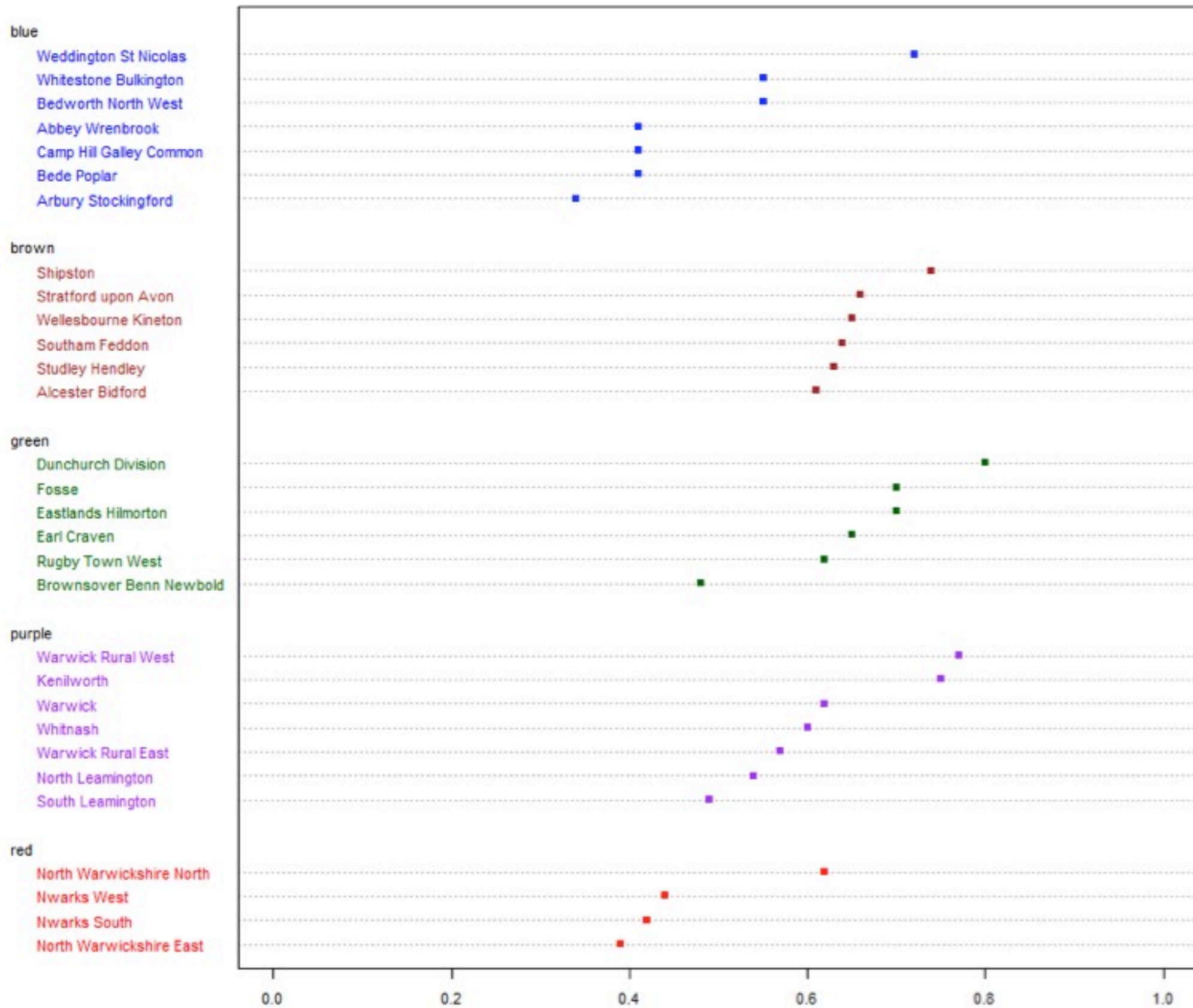
PERCENTAGE OF PUPILS GAINING 5 OR MORE GCSEs AT GRADES A*-C, INCLUDING ENGLISH AND MATHS, IN 2010 BY LOCALITY

Source: Warwickshire County Council (CYPF Directorate), Warwickshire Observatory



Based on residence, not school location

Proportion of pupils gaining good GCSEs



- visualization design process**
- types of research contributions**

COMMENTS ON READINGS

-visualization design process

-types of research contributions

target

target a ...

... specific domain

... set of users

learn about goals of
users and kinds of data

acquire and clean data

How do you select and foster a good collaboration?

target



translate

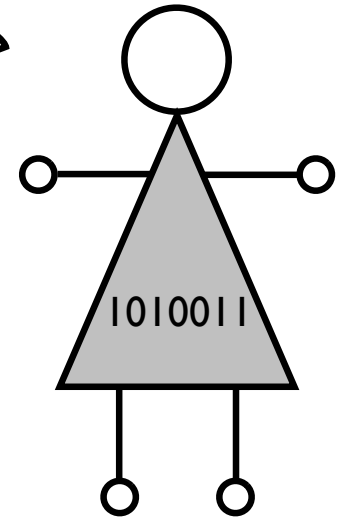
translate goals into data
analysis tasks

structure and characterize data

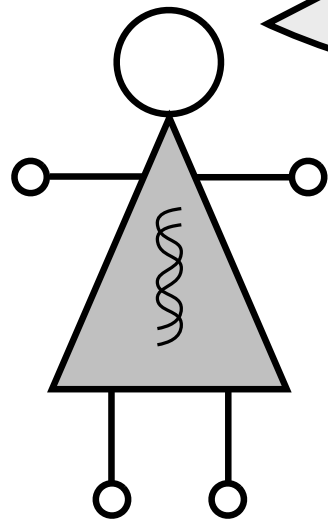
create abstraction of problem



What do you want to visualize?

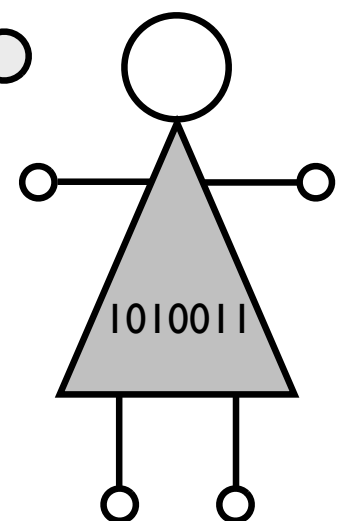


From patterns of conservation in gene expression profiles we want to visualize the evolutionary mechanisms that influence gene regulation.



*blah blah blah blah blah blah
blah blah **visualize** blah blah
blah blah blah blah*

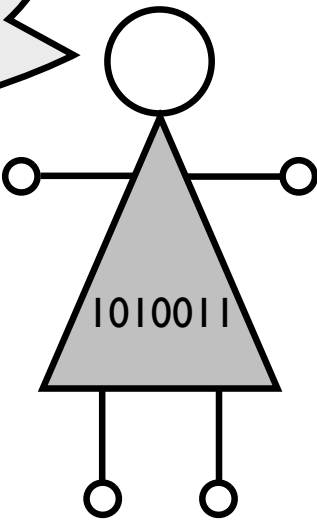
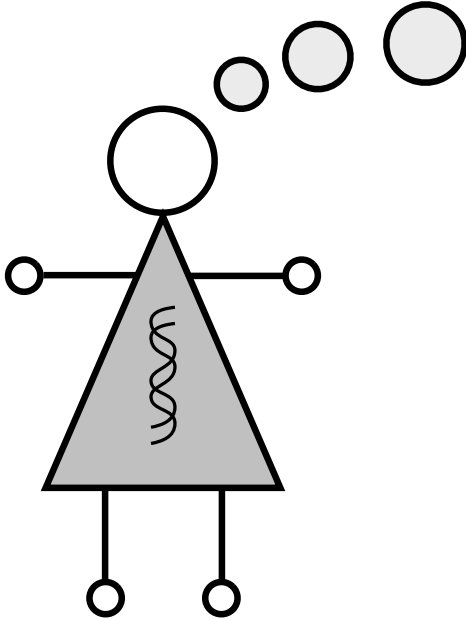
hmmmm... ahHA!



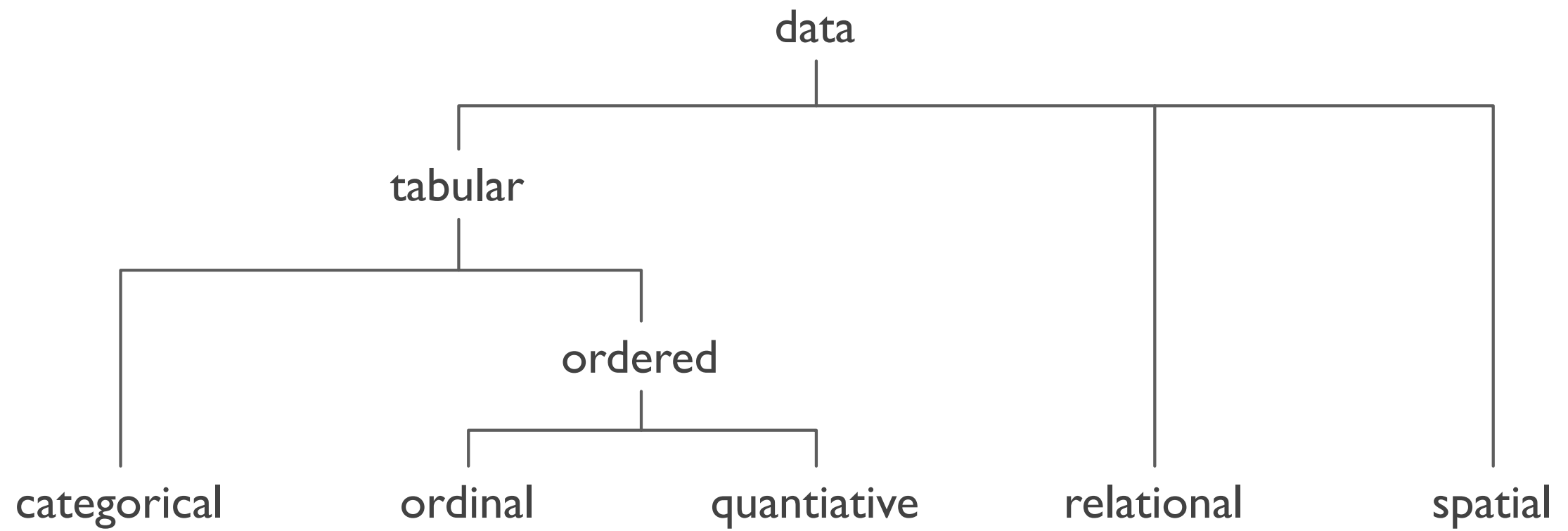
So, you want to characterize the differences in, and find similar groups of, time series that correspond to nodes in a binary tree.

*blah blah blah blah blah blah blah
blah blah blah blah blah blah blah
blah blah blah blah blah*

1010011



this Thursday



4 AN ANALYTIC TASK TAXONOMY

The ten tasks from the affinity diagramming analysis are:

- Retrieve Value
- Filter
- Compute Derived Value
- Find Extremum
- Sort
- Determine Range
- Characterize Distribution
- Find Anomalies
- Cluster
- Correlate

Each of the tasks is presented in the following sections, along with a *pro forma* abstract [9] and example questions that serve as general models and examples of the tasks. These tasks are not meant to be a normative picture of user analytic activity, but rather to provide a vocabulary for discussion.

target



translate



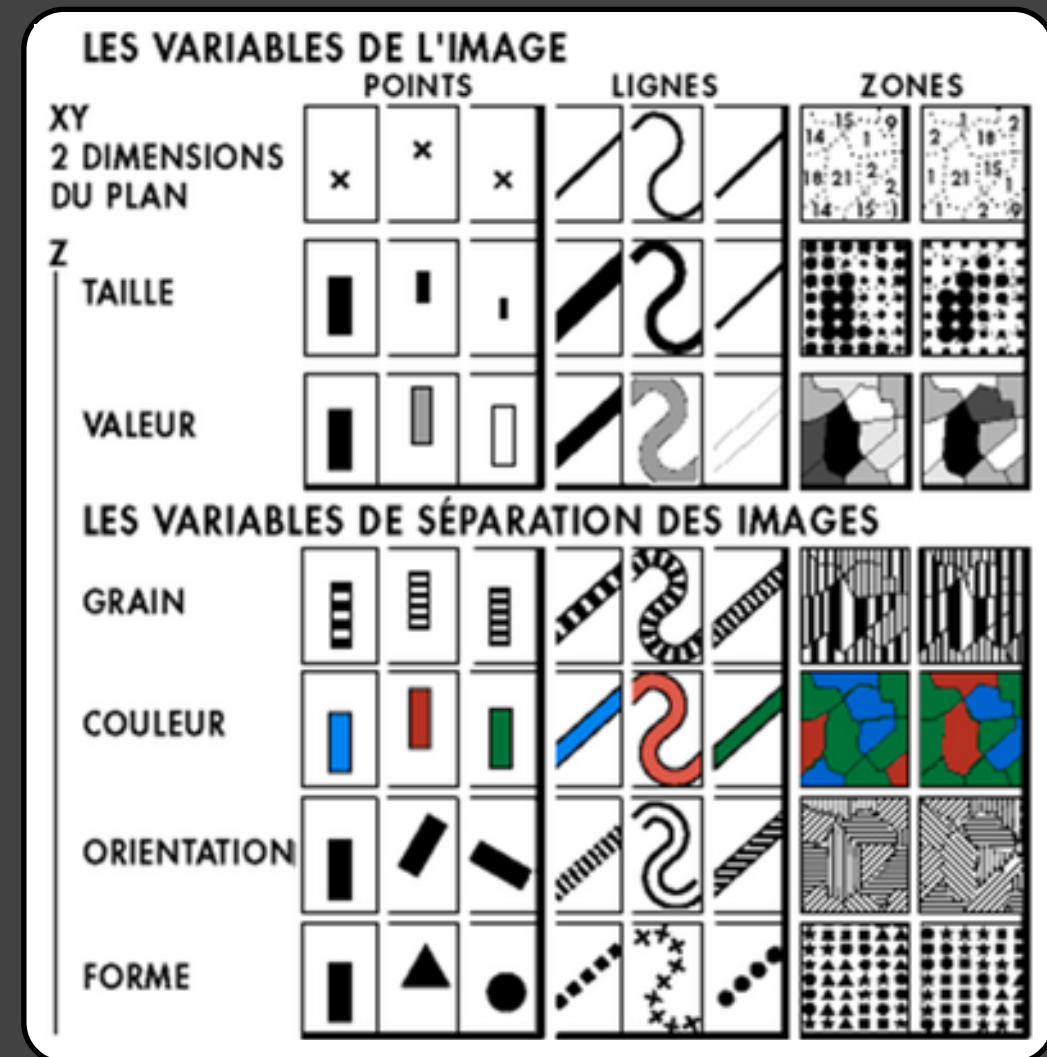
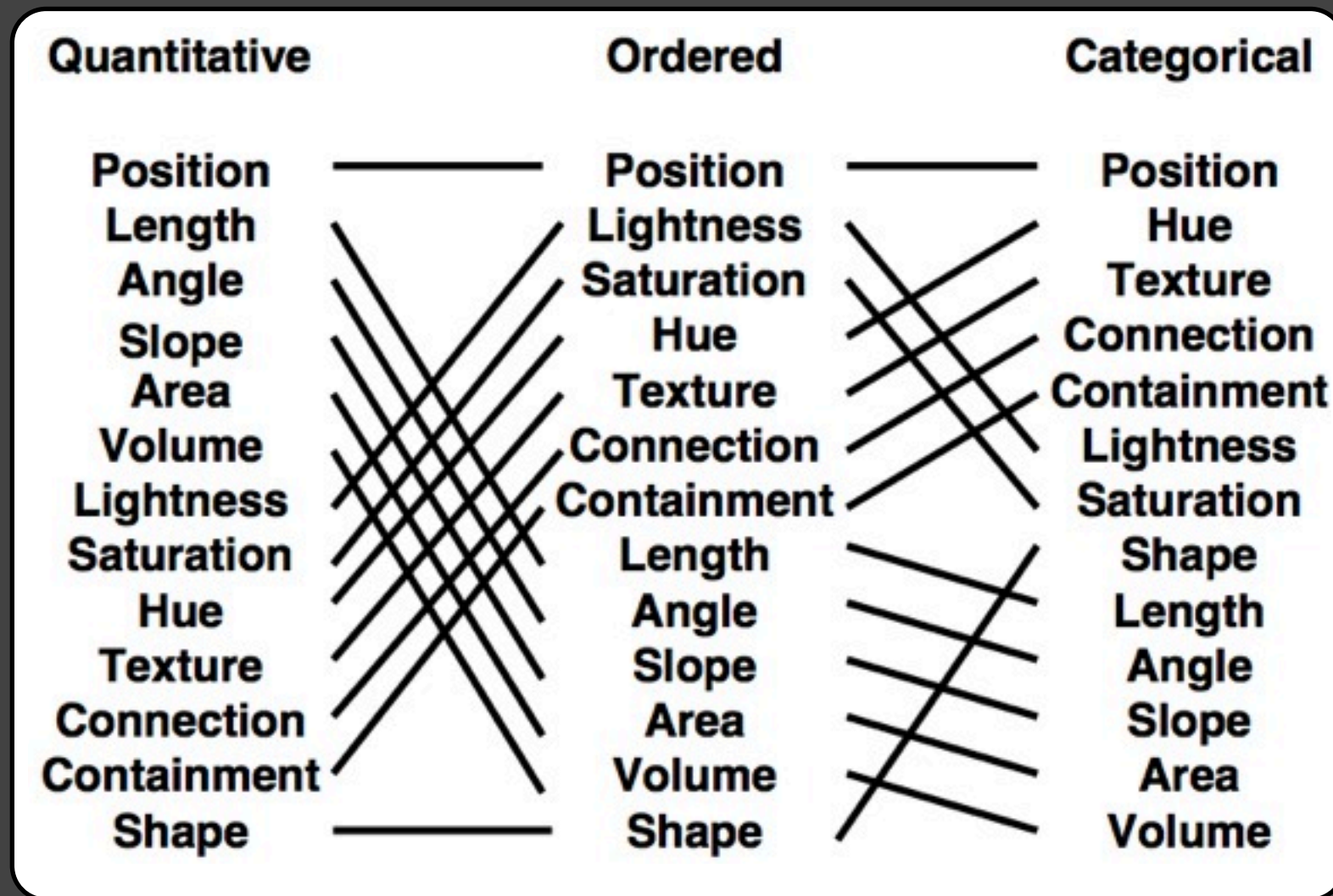
design

design visual encodings and interaction mechanisms to support data and task abstraction

transform data with appropriate computational methods

try many ideas!

next Tuesday

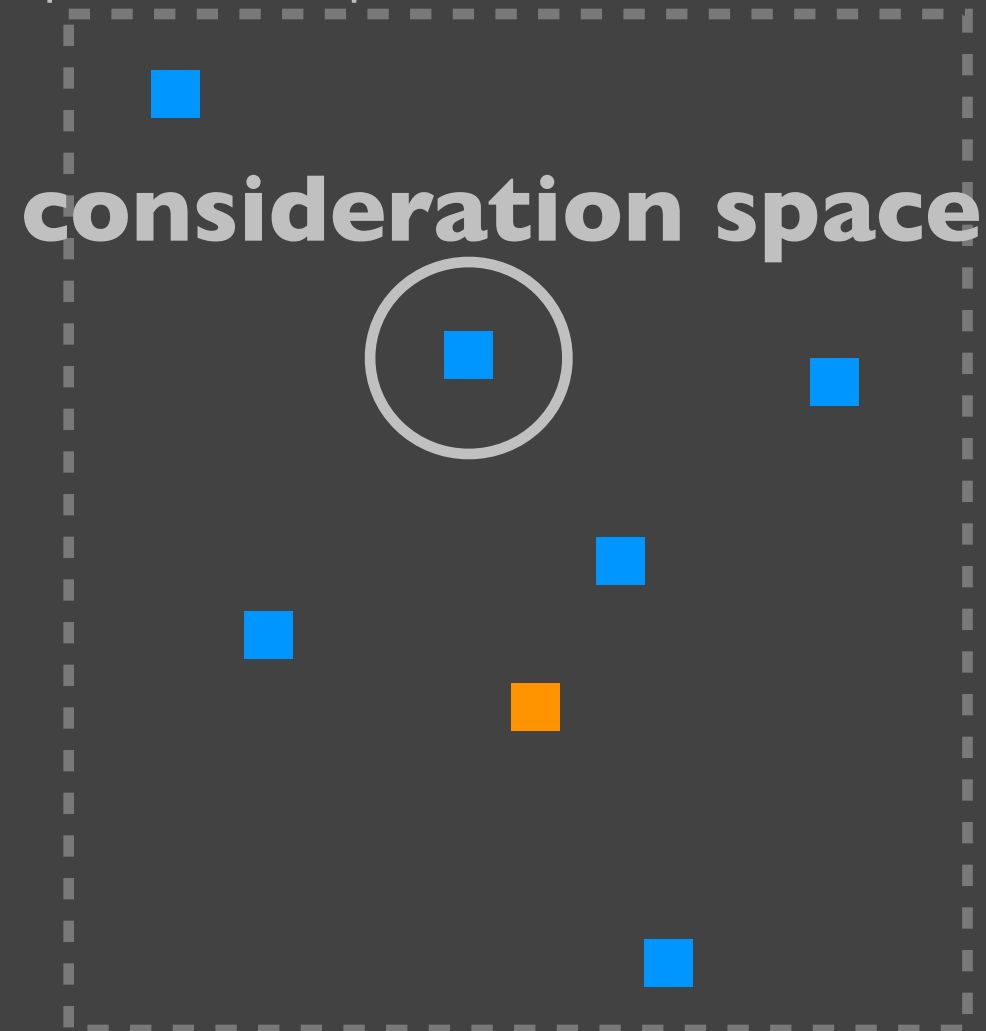


Automating the Design of Graphical Presentations of Relational Information
MacKinlay, 1986

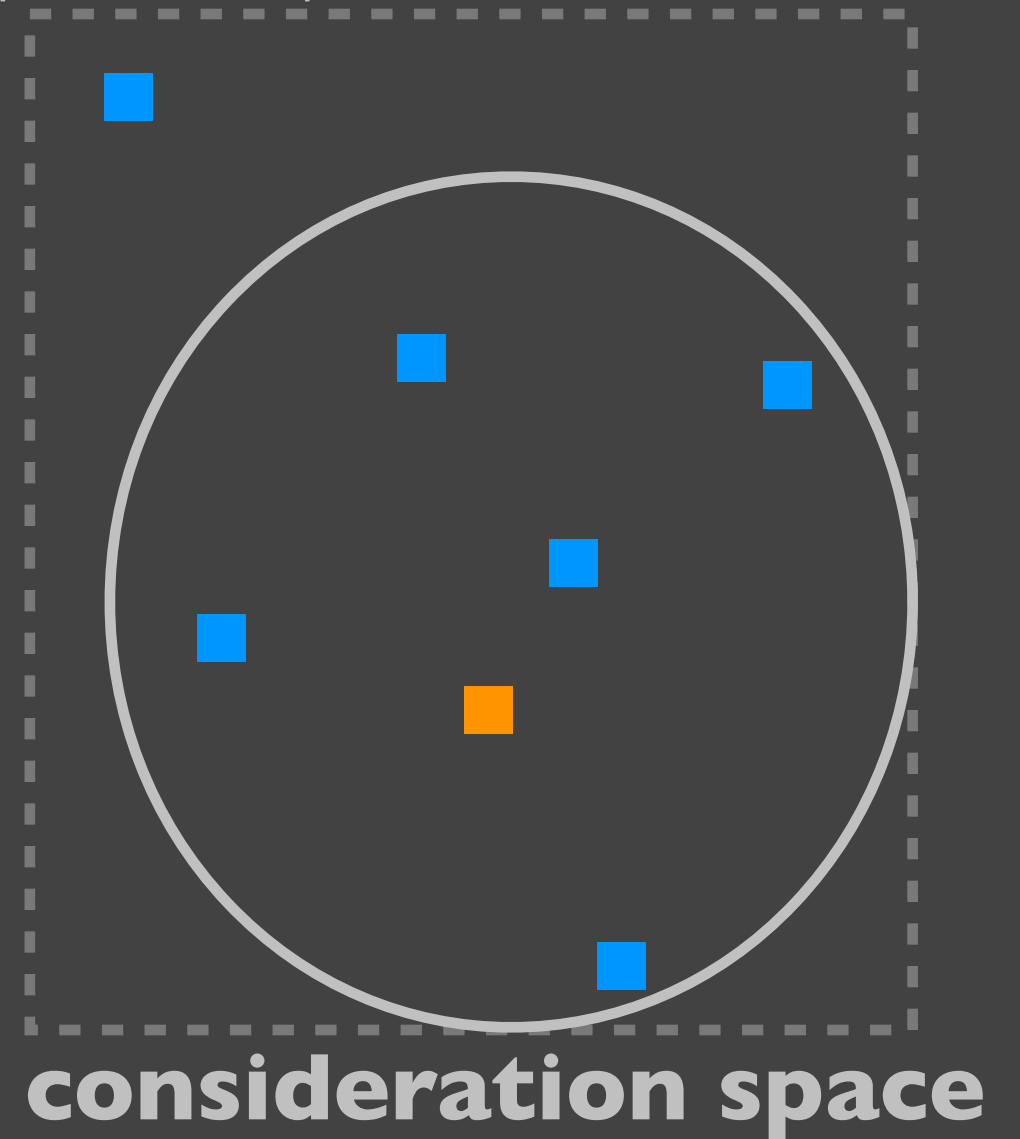
Semiology of Graphics
Bertin, 1967

try many ideas

space of possible solutions



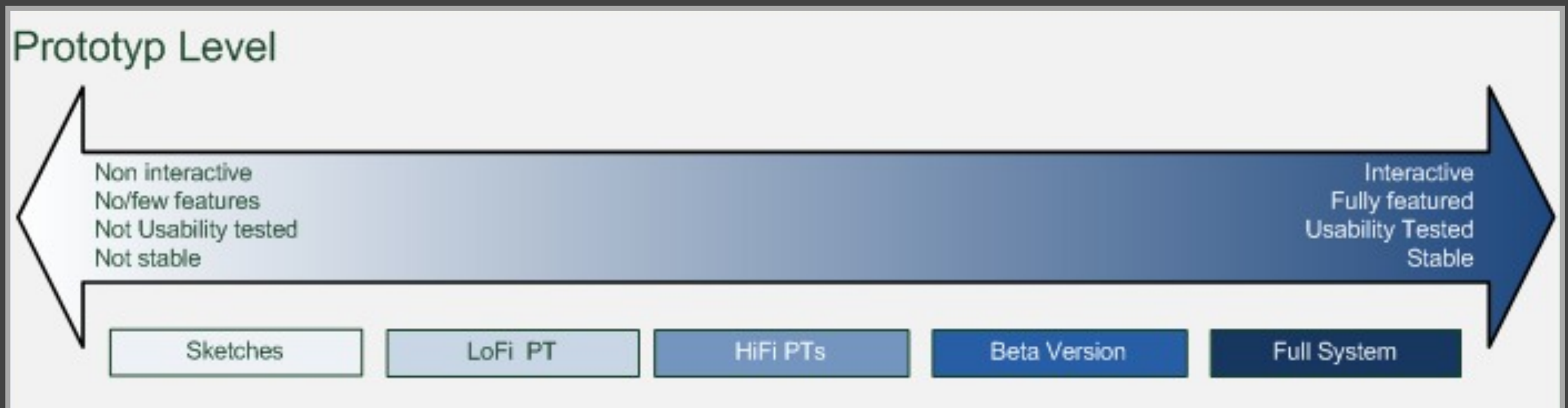
space of possible solutions



■ good solution

■ best solution

RAPID PROTOTYPING



ideation exercise ...

target



translate




design



implement

bring your design to life

optimize algorithms




Processing


Cover \ Exhibition \ Reference \ Learning \ Download \ Shop \ About

»Feed »Forum »Wiki »Code

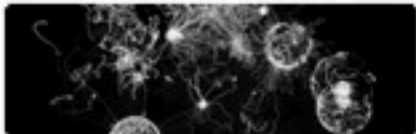
» Exhibition




[Composition No. 1](#)
by Visual Editions



[Max Planck Research Networks](#)
by Moritz Stefaner and Christopher Warnow



[The Creators](#)
by Constanza Casas, Mark C Mitchell and Pieter Steyaert

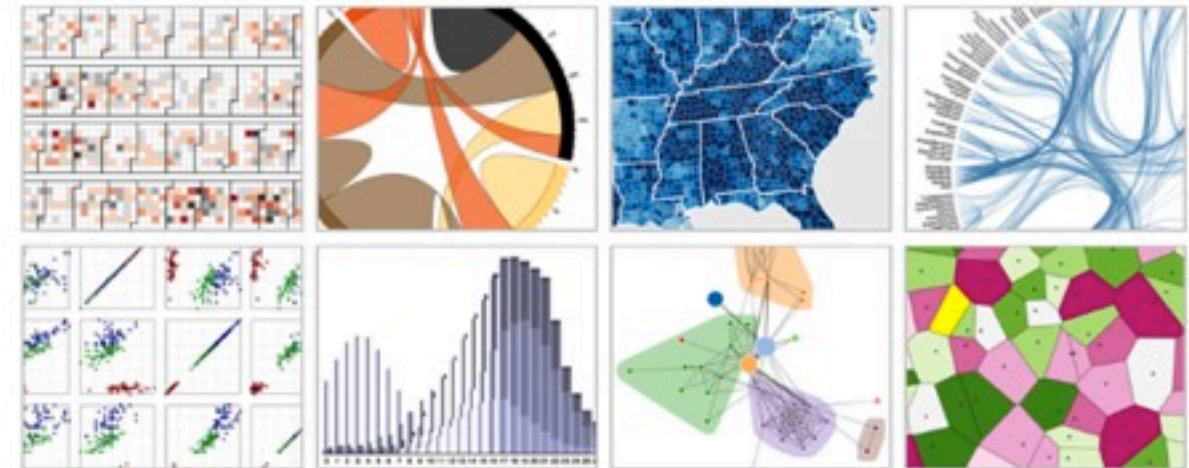


[The Digital Rube Goldberg Processor](#)
by The Product

Processing was initiated by Ben Fry and Casey Reas. It is developed by a small team of volunteers.
© Info \ Site hosted by Media Temple

D3: Data-Driven Documents

Michael Bostock, Vadim Ogievetsky, Jeffrey Heer



Interactive visualizations built with D3. From left to right: calendar view, chord diagram, choropleth map, hierarchical edge bundling, scatterplot matrix, grouped & stacked bars, force-directed graph clusters, Voronoi tessellation.


ABSTRACT

Data-Driven Documents (D3) is a novel representation-transparent approach to visualization for the web. Rather than hide the underlying scenegraph within a toolkit-specific abstraction, D3 enables direct inspection and manipulation of a native representation: the standard document object model (DOM). With D3, designers selectively bind input data to arbitrary document elements, applying dynamic transforms to both generate and modify content. We show how representational transparency improves expressiveness and better integrates with developer tools than prior approaches, while offering comparable notational efficiency and retaining powerful declarative components. Immediate evaluation of operators further simplifies debugging and allows iterative development. Additionally, we demonstrate how D3 transforms naturally enable animation and interaction with dramatic performance improvements over intermediate representations.

MATERIALS AND LINKS

PDF (2.2 MB) | Software | Video | BibTeX Citation

CITATION

 **D3: Data-Driven Documents**
Michael Bostock, Vadim Ogievetsky, Jeffrey Heer
IEEE Trans. Visualization & Comp. Graphics (Proc. InfoVis), 2011
PDF (2.2 MB) | Software | Video

target



translate

problem characterization
and abstraction

80%



design

visualization design

20%



implement



target



translate



design



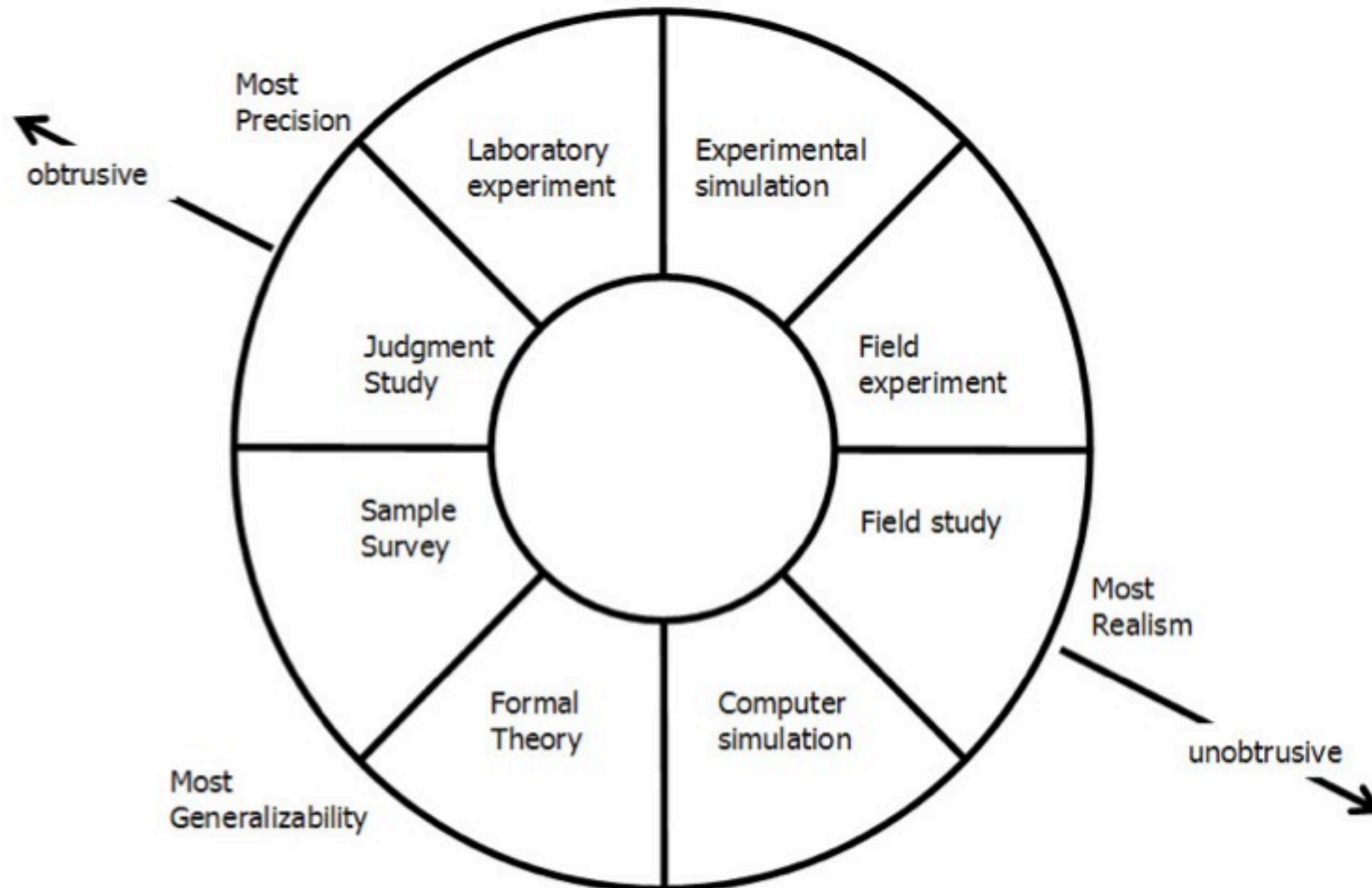
implement



validate

ensure visualization system
supports target users' goals

in late March



**no amount of brilliant design
can overcome designing for
the wrong thing**

target

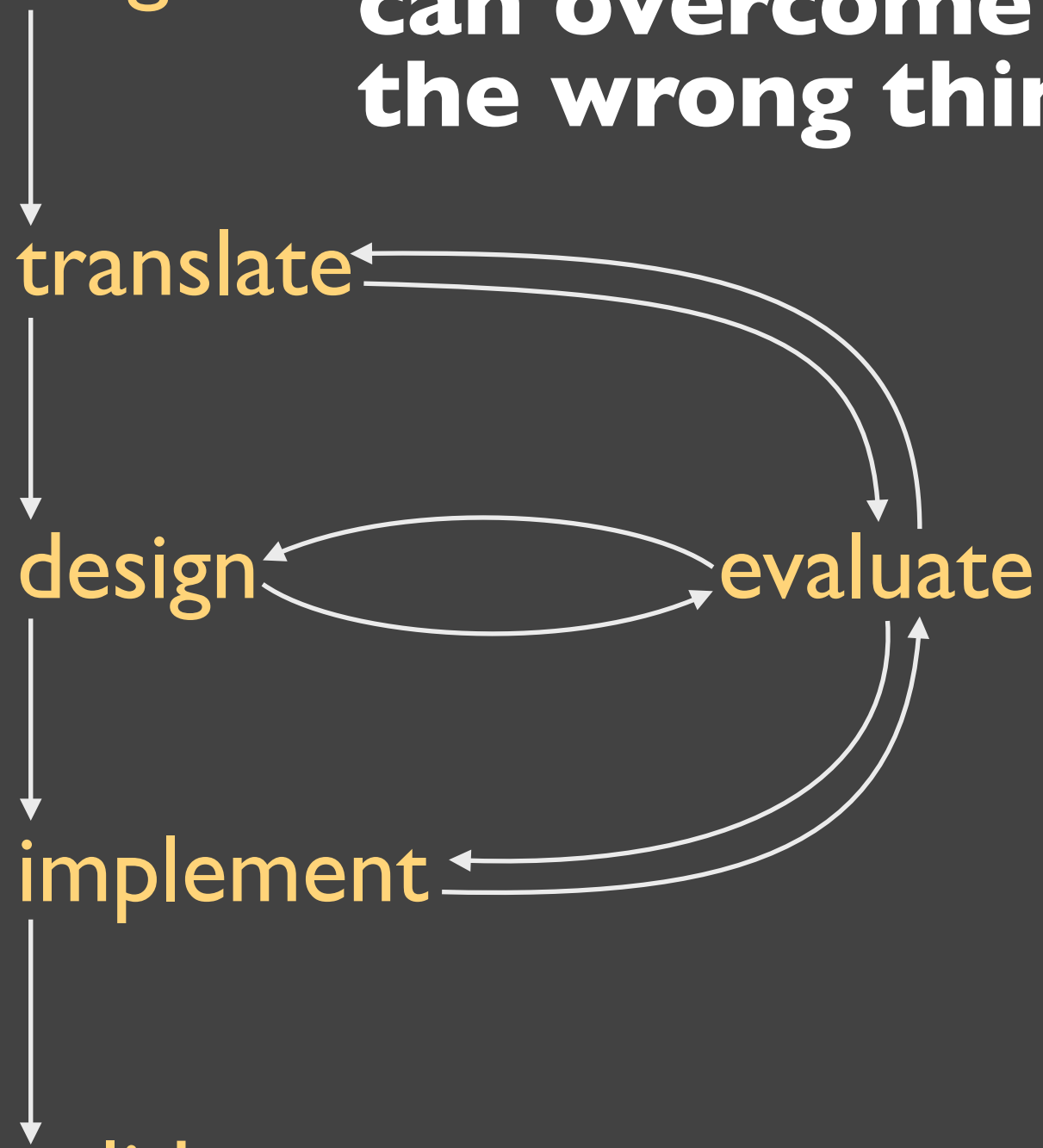
translate

design

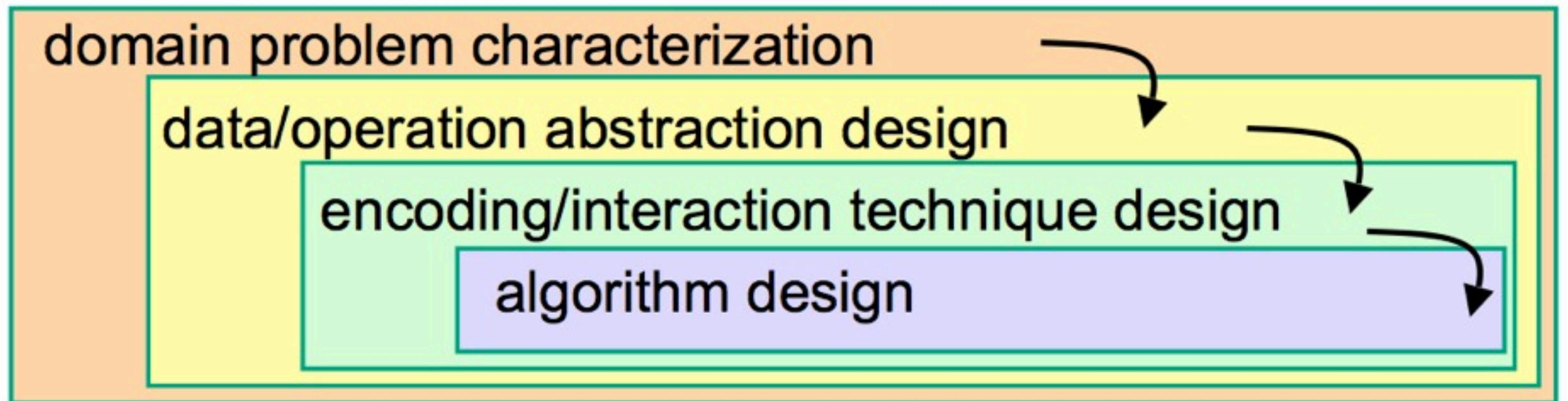
evaluate

implement

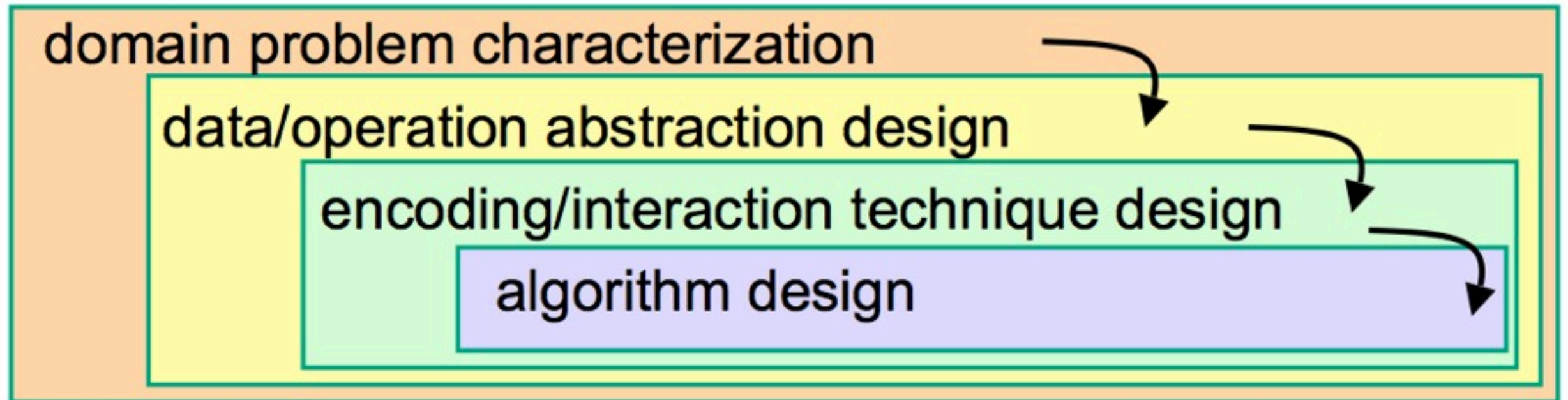
validate



NESTED MODEL



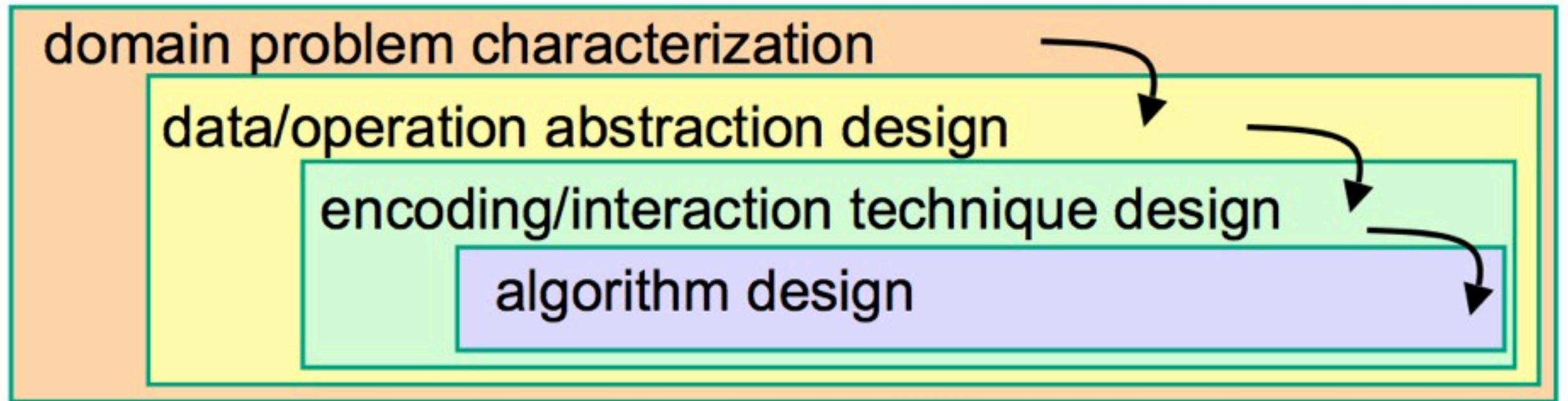
what can go wrong, and how do we validate?



problem

threat: they don't do that

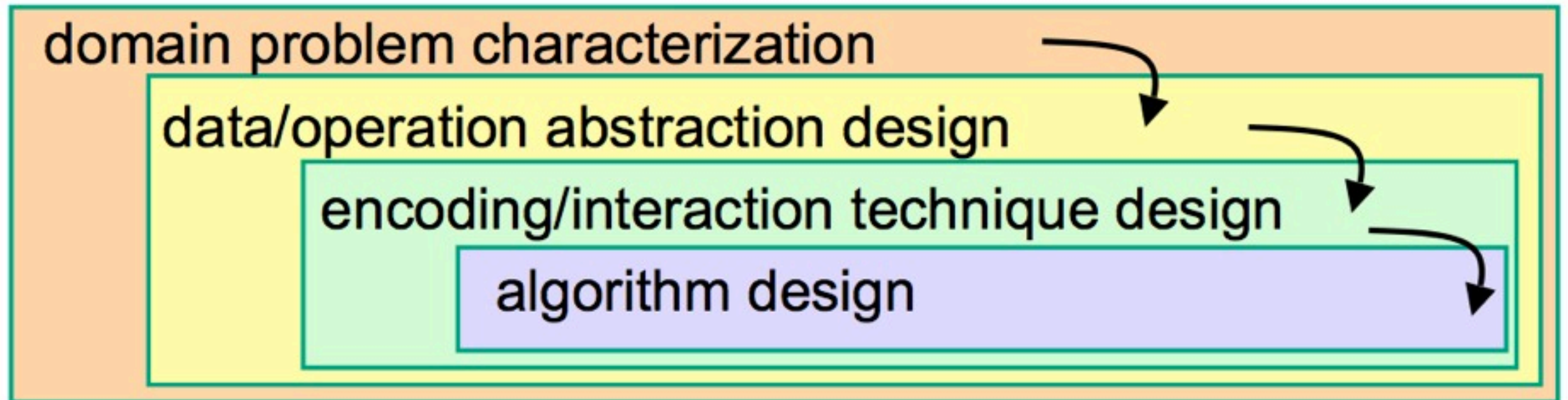
validation: *immediate*, observe and interview users
downstream, notice adoption rates



abstraction

threat: you're showing them the wrong thing

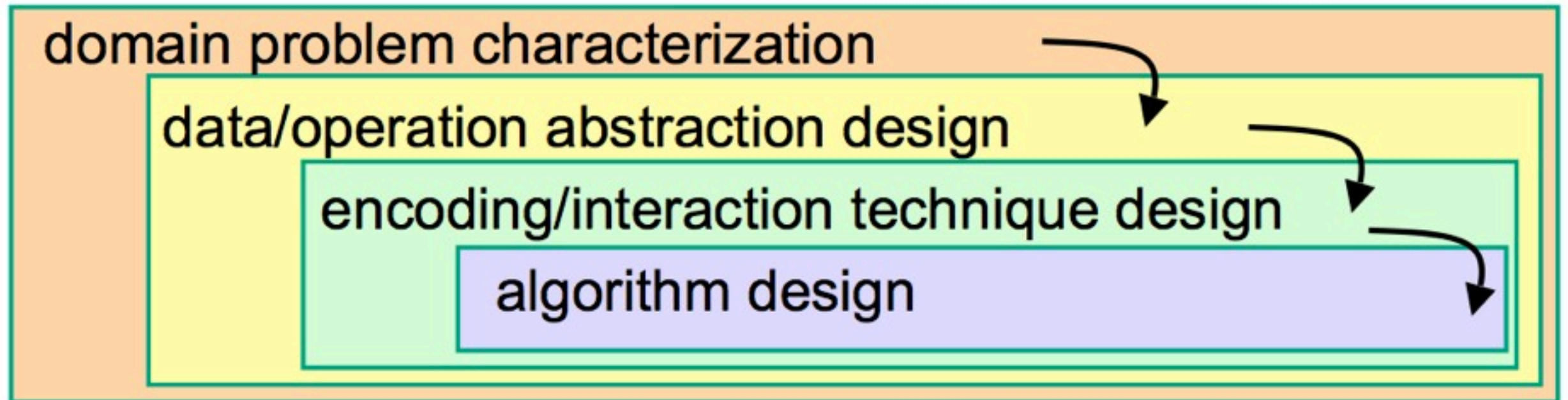
validation: *downstream*, deploy and observe usage



design

threat: the way you show it doesn't work

validation: *immediate*, justification with known principles
downstream, qualitative or quantitative
analysis of results; lab study measuring time
and error

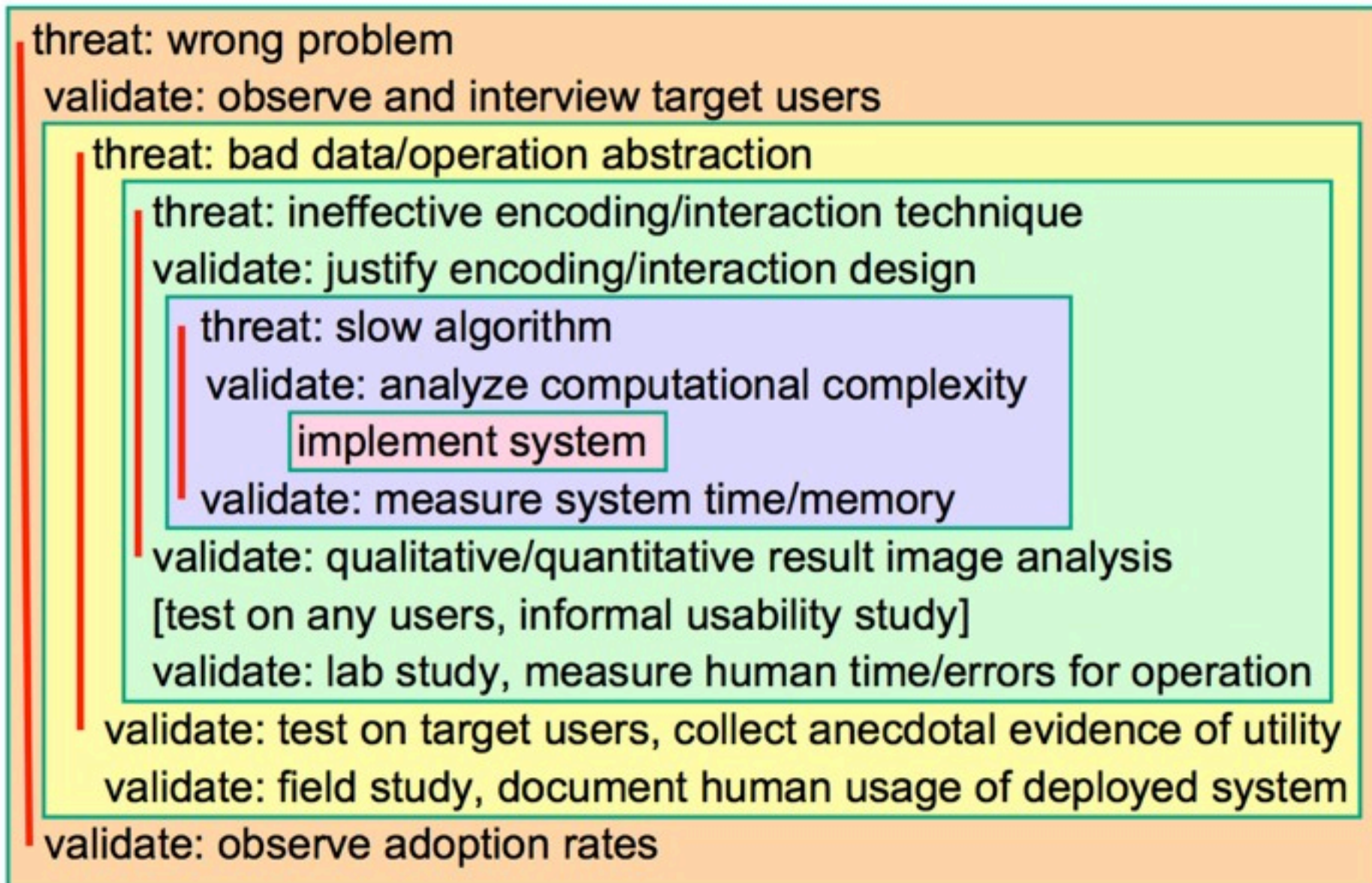


algorithm

threat: you're code is too slow

validation: *immediate*, complexity analysis

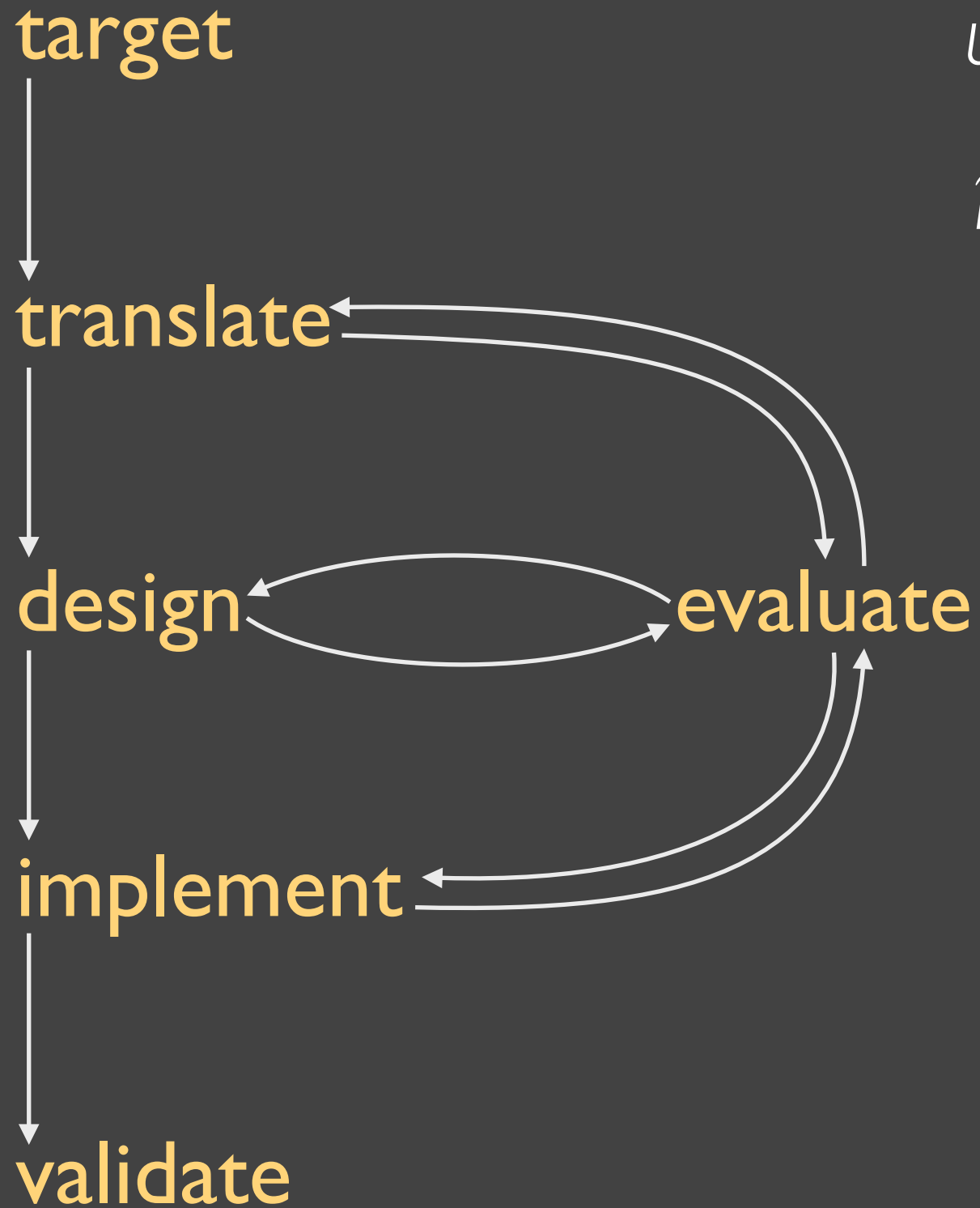
downstream, benchmarks for system time
and memory

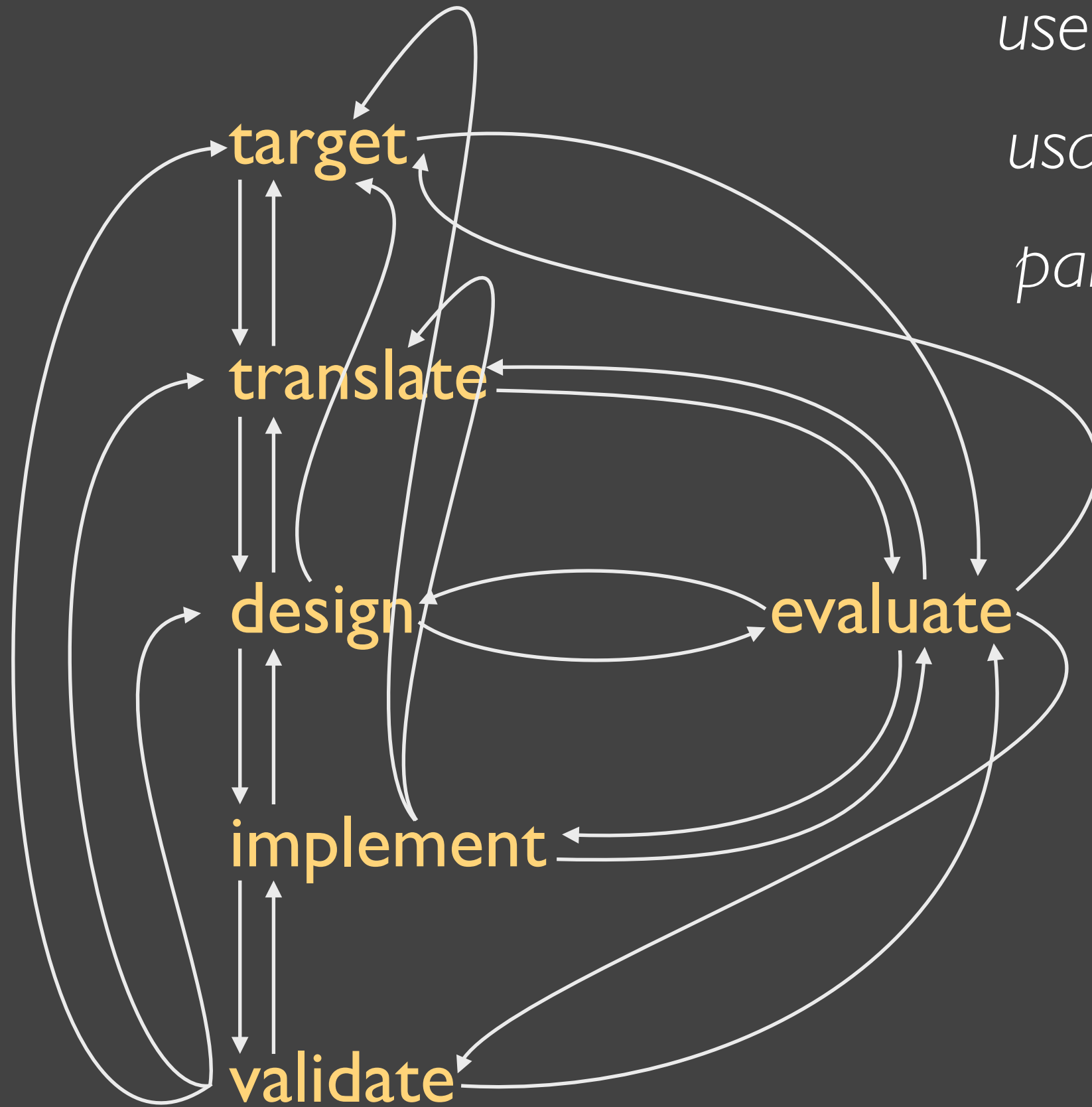


avoid validation mismatch

- cannot validate encoding with system timings
- cannot validate abstraction with lab studies

user-centered design
usability engineering
participatory design
design thinking





user-centered design
usability engineering
participatory design
design thinking

target

translate

design

implement

validate

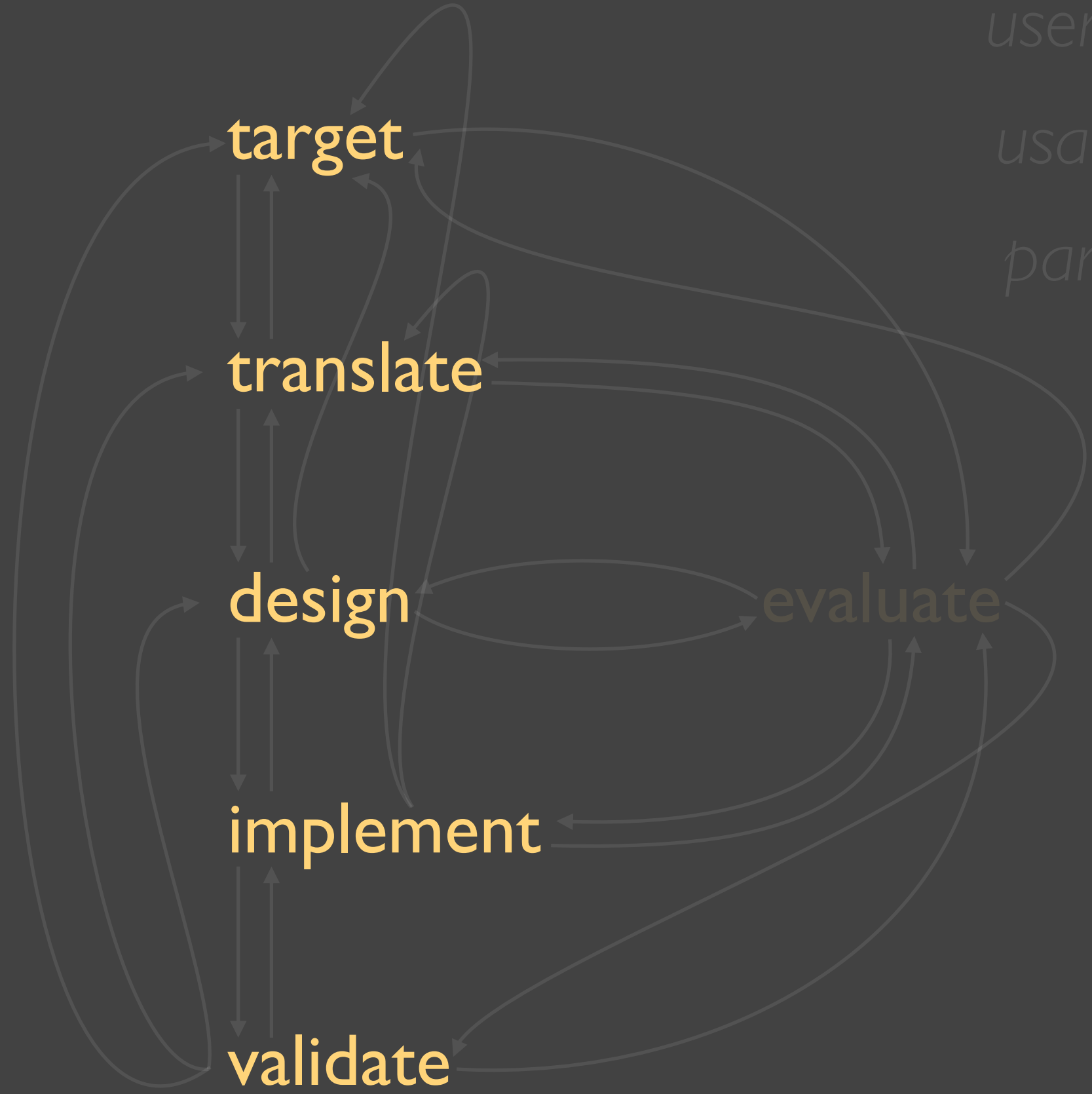
evaluate

user-centered design

usability engineering

participatory design

design thinking

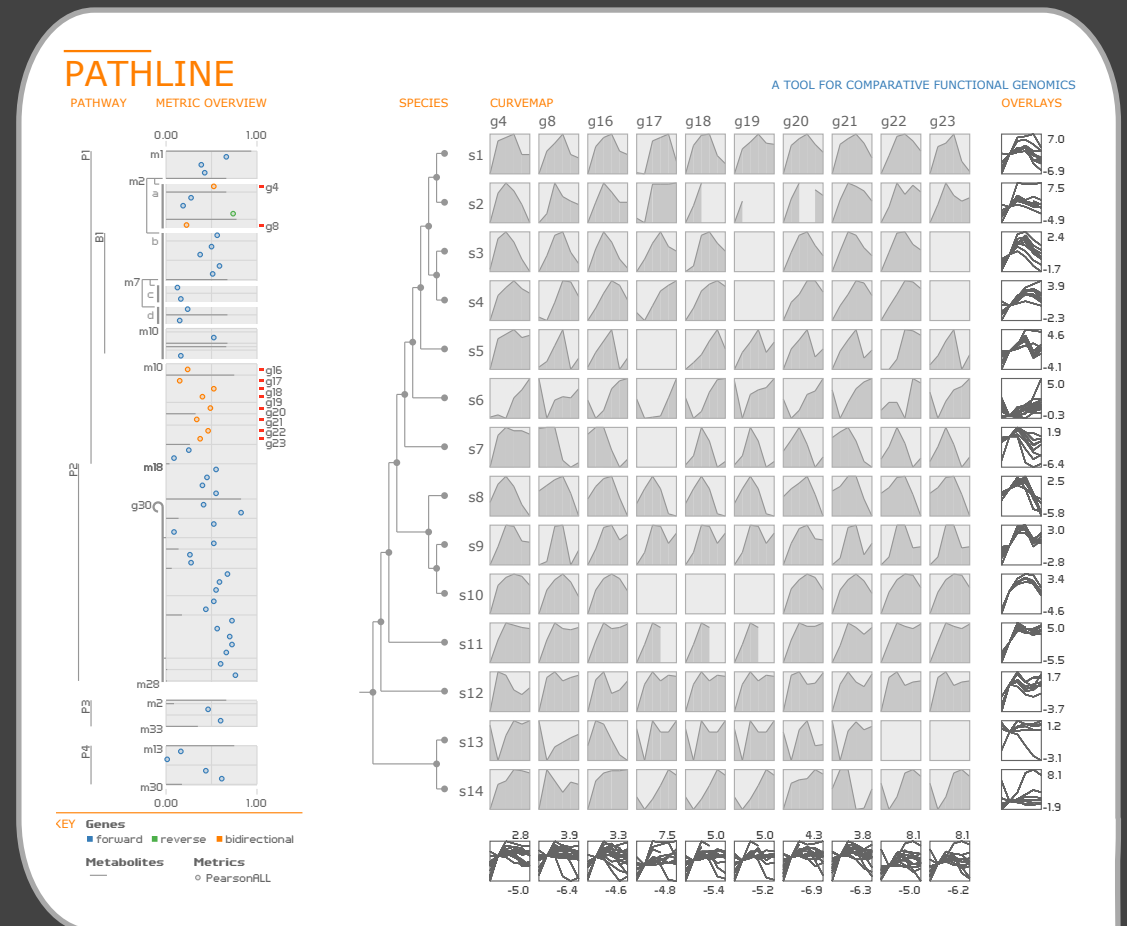


Pathline

A Tool for Comparative Functional Genomics Data

joint work with:

Bang Wong, Mark Styczynski, Tamara Munzner, Hanspeter Pfister



Pathline: A Tool for Comparative Functional Genomics
M. Meyer et al., IEEE/Eurographics EuroVis 2010.

target

translate

design

implement

validate

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 ...

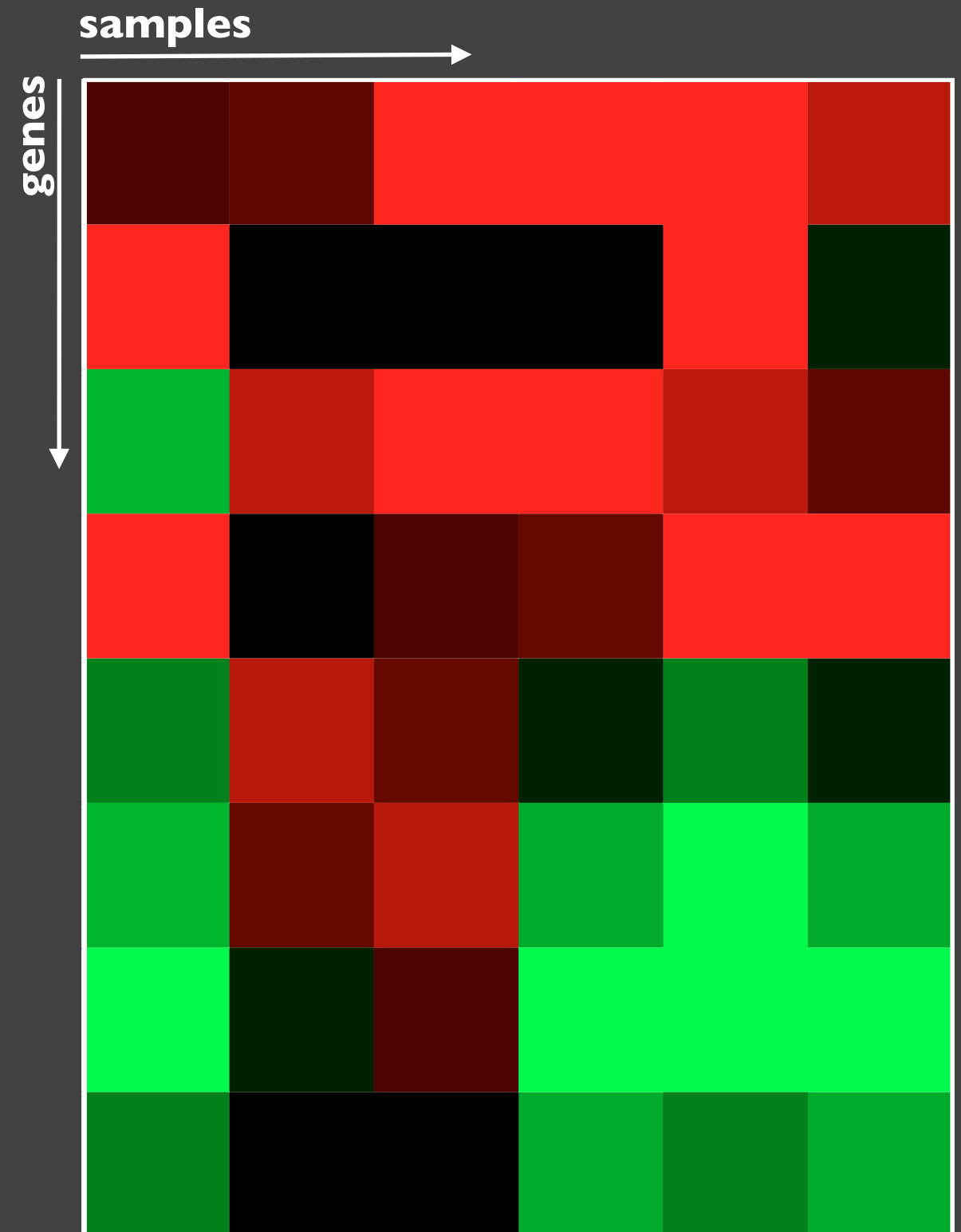
... for many genes

... 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

... a single quantitative value

biologists measure it ...

... for many genes

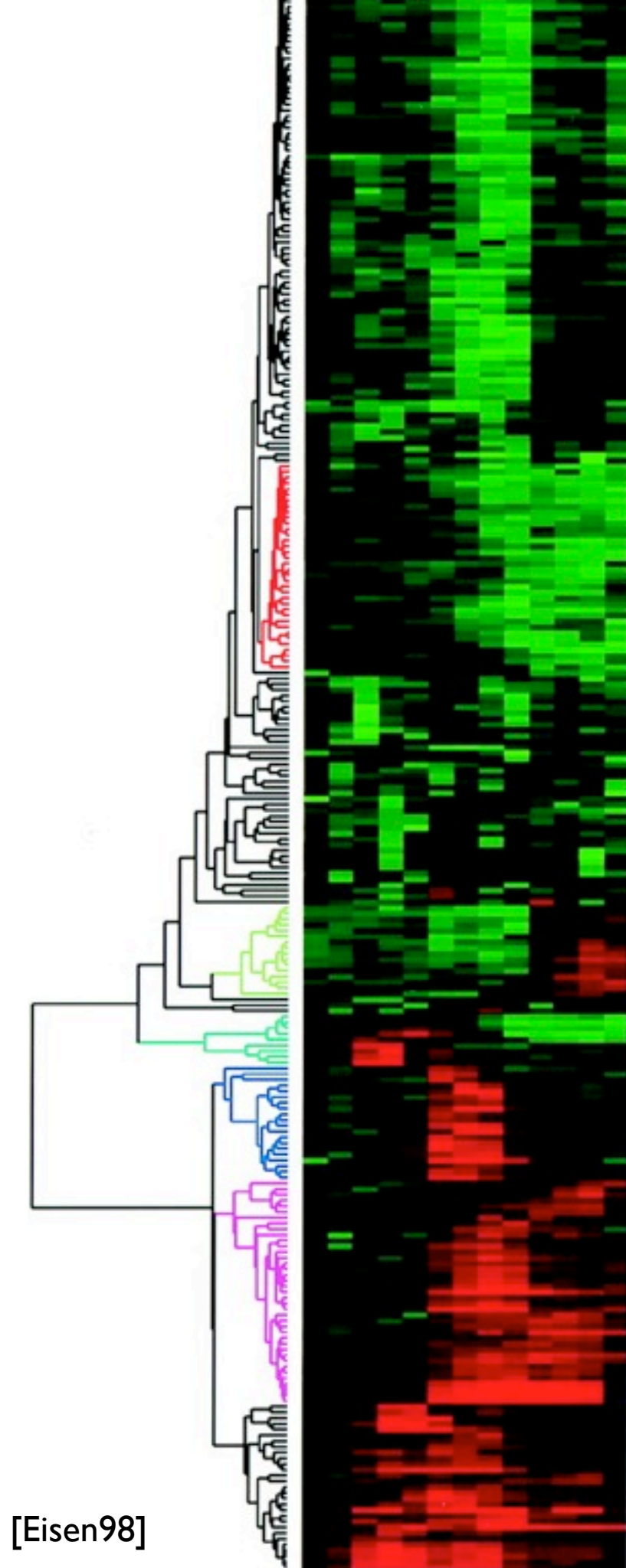
... in many samples (time points, tissue types, species)

visualized with heatmaps

[Wilkinson09] [Saldanha04] [Seo02] [Eisen98]
[Gehlenborg10] [Weinstein08]

encode value with color

augmented with clustering



[Eisen98]

functional genomics data

gene expression

molecular pathways

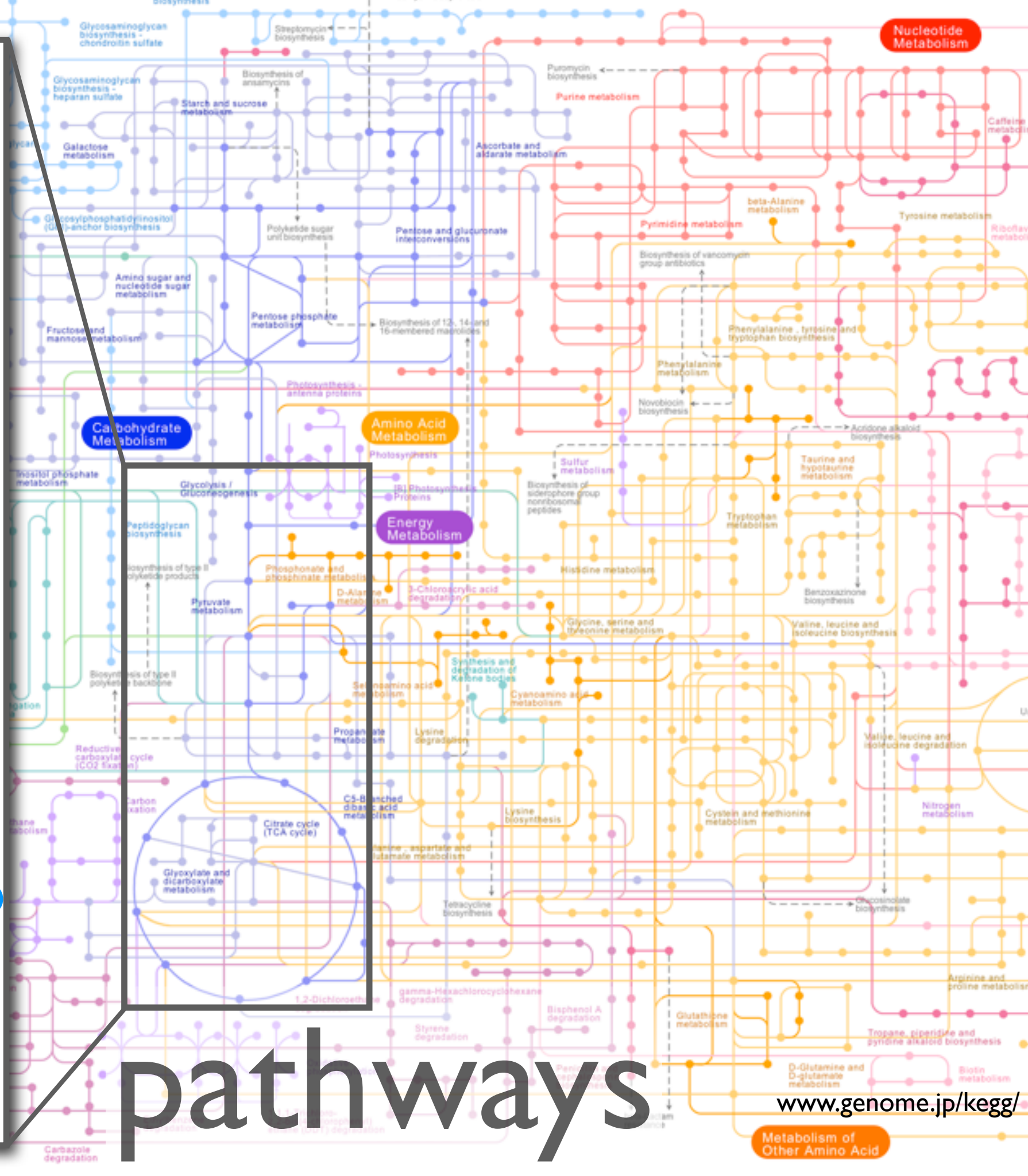
the functioning of a cell is controlled by many interrelated chemical reactions performed by genes



glycolysis

tca cycle

pathways



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?*

target

translate

design

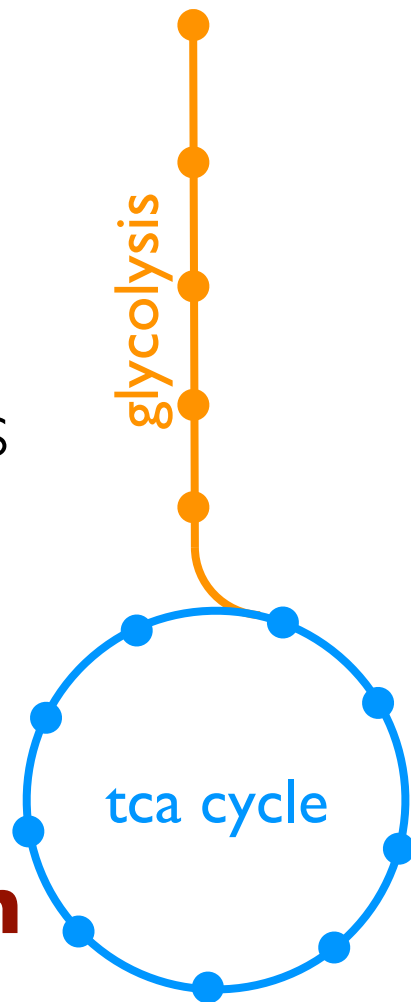
implement

validate

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**

A 3D table visualization showing gene expression data. The table is structured with species (s1 to s6) on the vertical axis and time points (t1 to t6) on the horizontal axis. The data is presented as a series of overlapping planes, each representing a different species. The values range from -1.0 to 1.0.

| Species | t1 | t2 | t3 | t4 | t5 | t6 |
|---------|------|-----|-----|------|------|------|
| g1 | 0.2 | 0.4 | 1.0 | 1.0 | 1.0 | 1.0 |
| m1 | 1.0 | 0.0 | 0.0 | 0.0 | 1.0 | 0.8 |
| g2 | -0.7 | 0.8 | 1.0 | 1.0 | 0.8 | 0.2 |
| m2 | 1.0 | 0.0 | 0.2 | 0.5 | 1.0 | 0.2 |
| g3 | -0.5 | 0.8 | 0.5 | -0.3 | -0.5 | -0.5 |
| m3 | -0.7 | 0.5 | 0.8 | -0.7 | -1.0 | 0.5 |

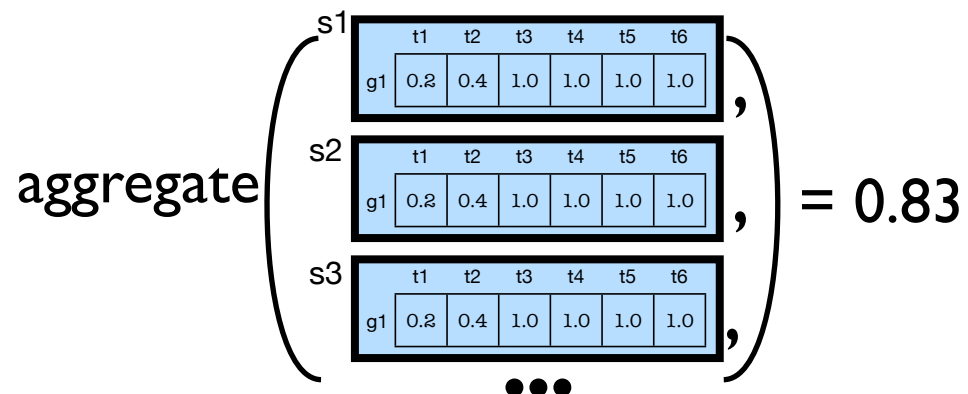
similarity scores

- aggregate time series for a gene/metabolite over species

- similarity of expression across species

- aggregate: Pearson, Spearman, others

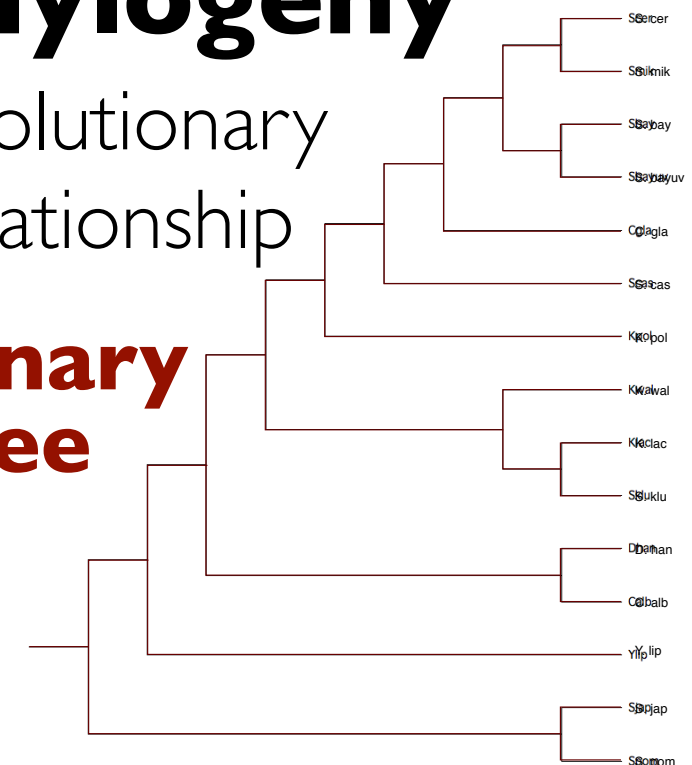
• **quantitative value**



phylogeny

- evolutionary relationship

• **binary tree**

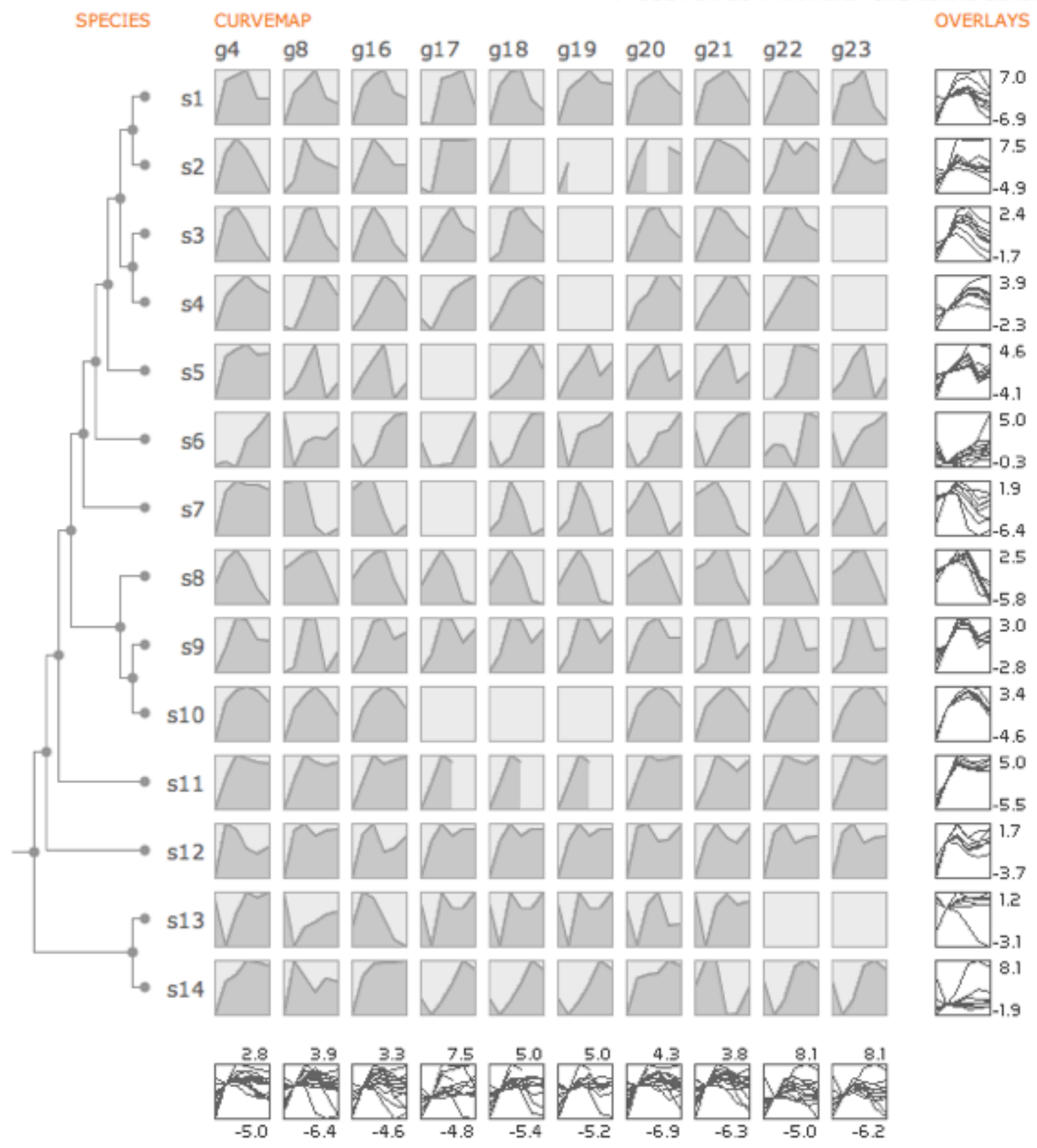
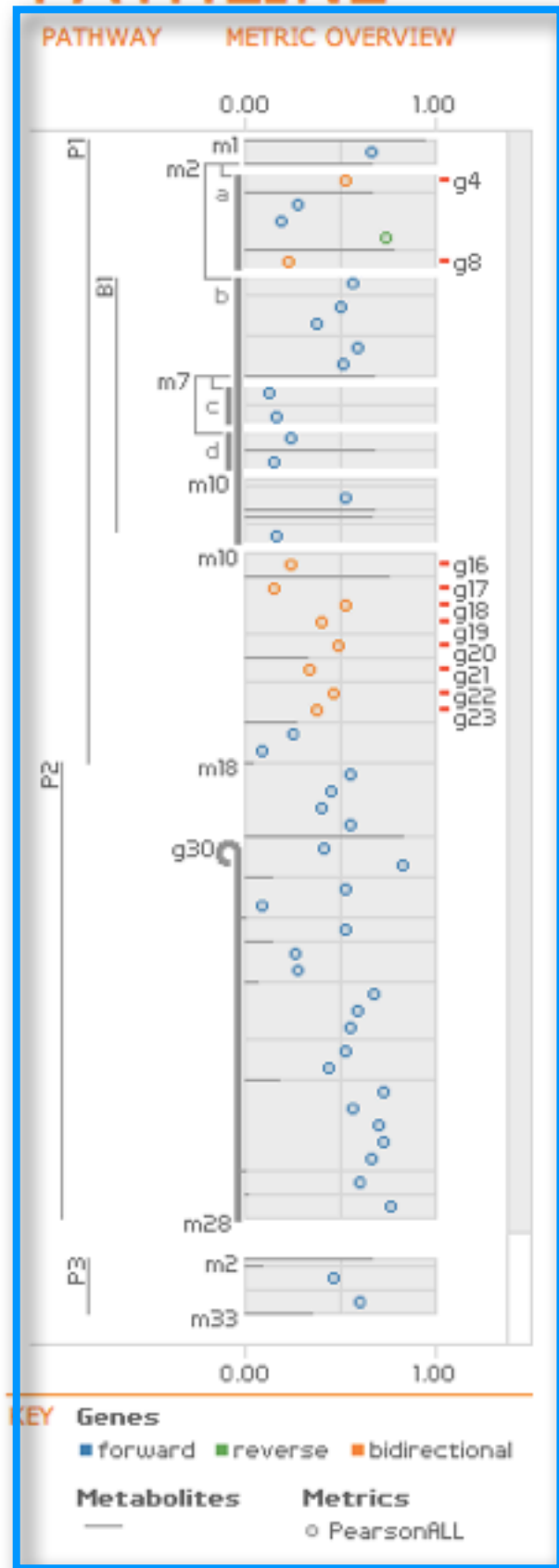


tasks

- 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

PATHLINE

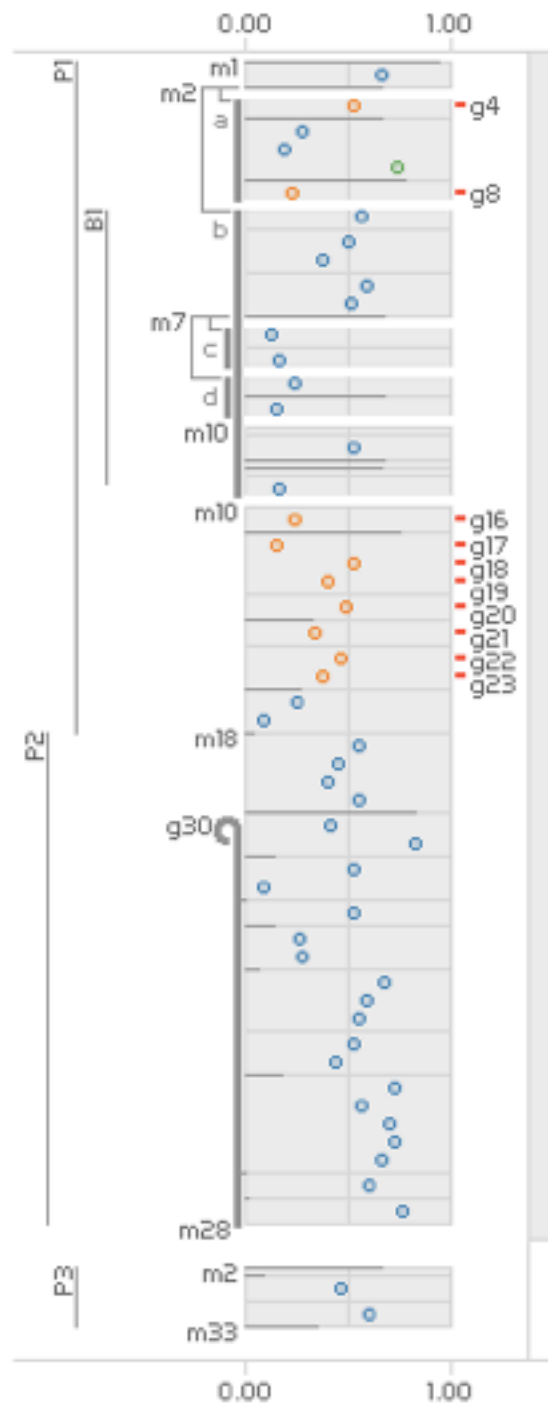
A TOOL FOR COMPARATIVE FUNCTIONAL GENOMICS



PATHLINE

A TOOL FOR COMPARATIVE FUNCTIONAL GENOMICS

PATHWAY METRIC OVERVIEW



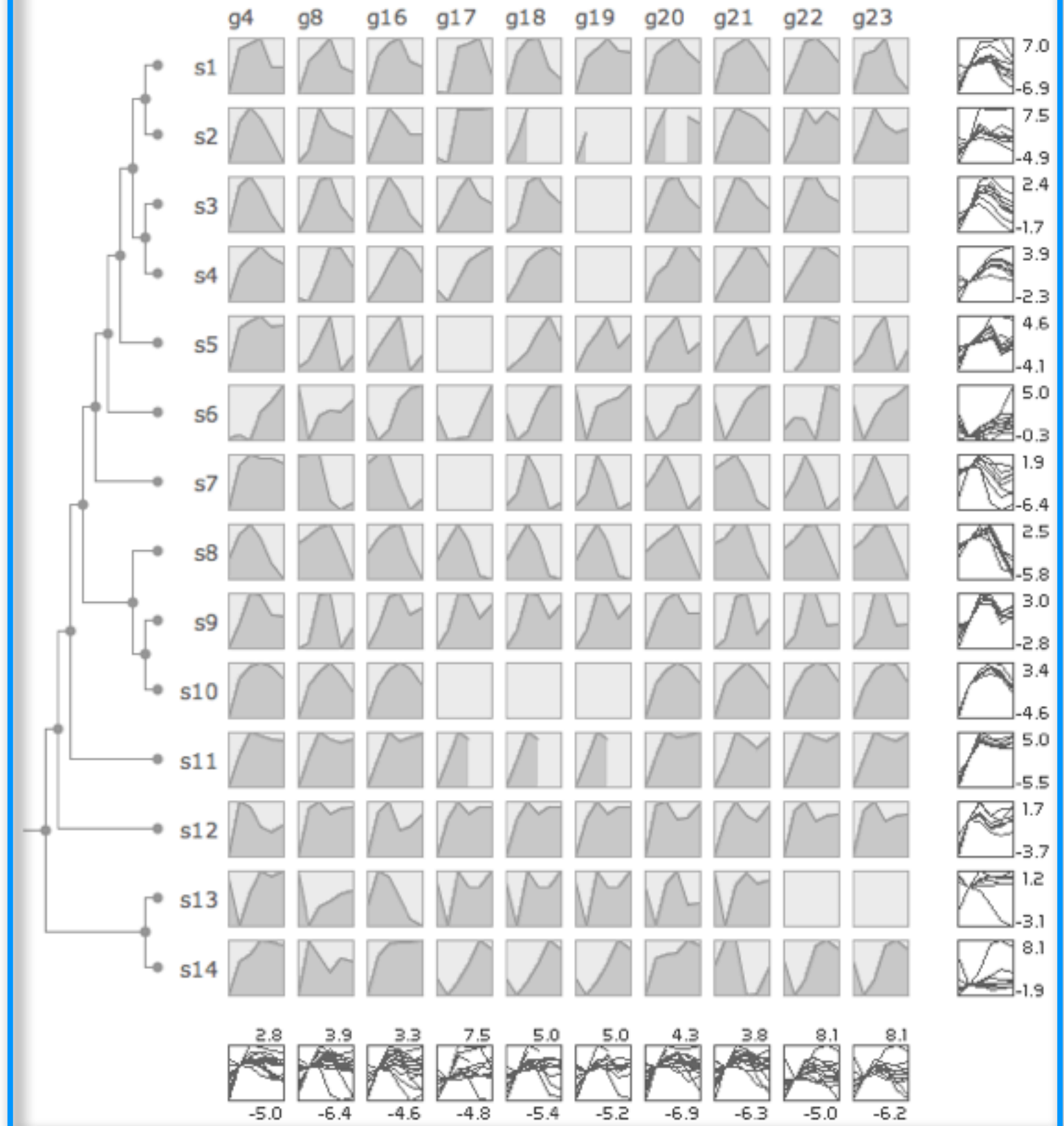
KEY Genes
 ■ forward ■ reverse ■ bidirectional

Metabolites —
Metrics ○ PearsonALL

SPECIES

CURVEMAP

OVERLAYS



PATHLINE

A TOOL FOR COMPARATIVE FUNCTIONAL GENOMICS

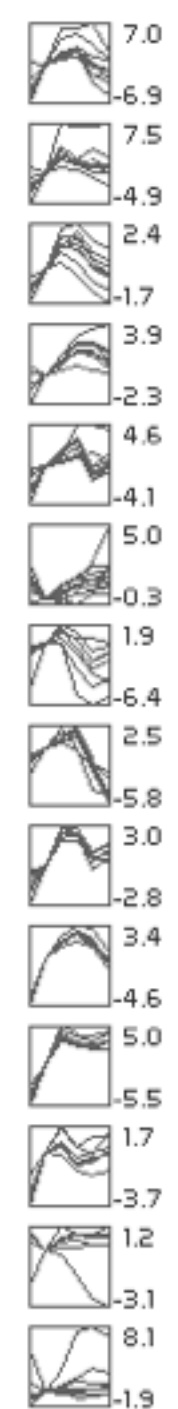
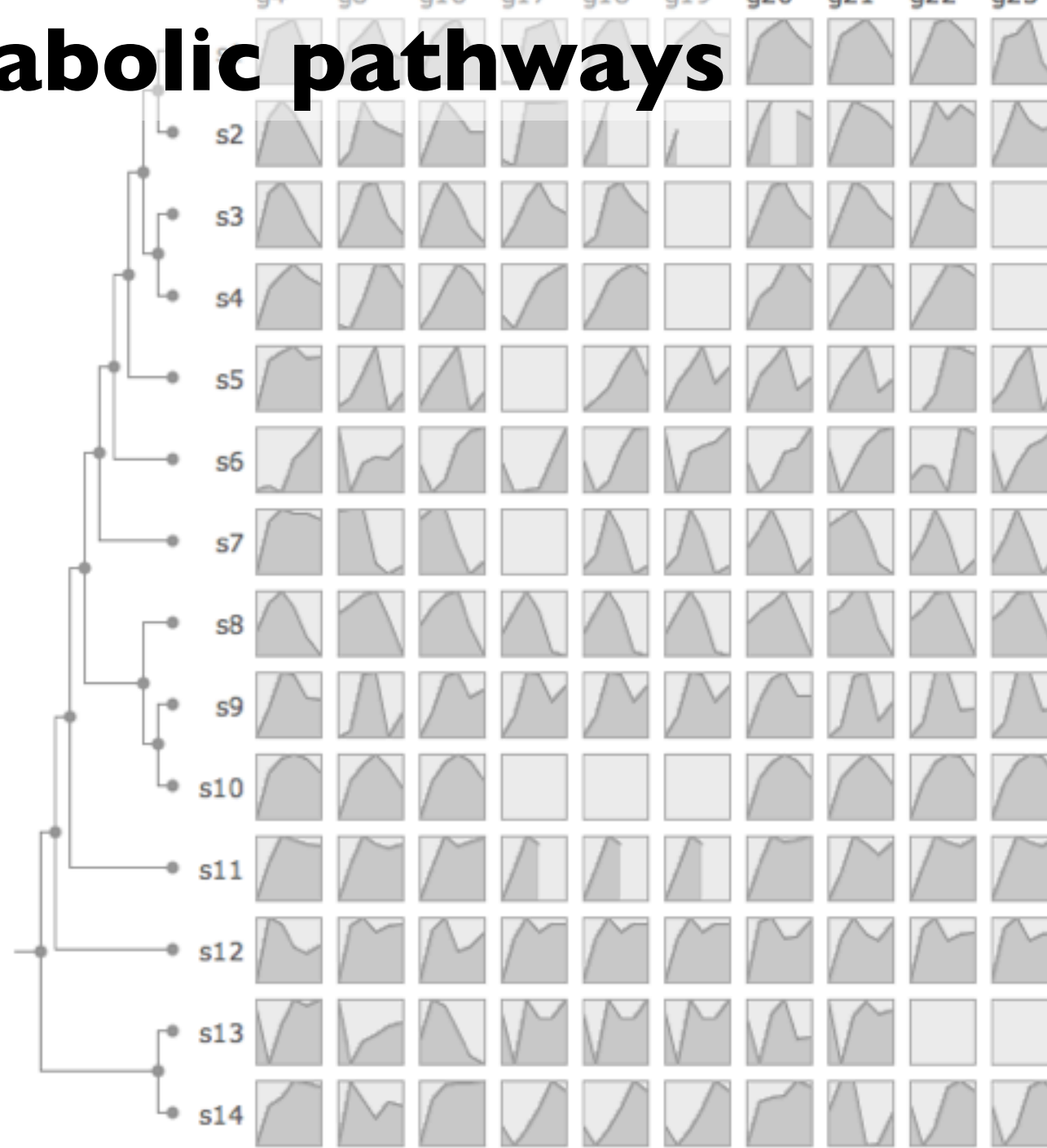
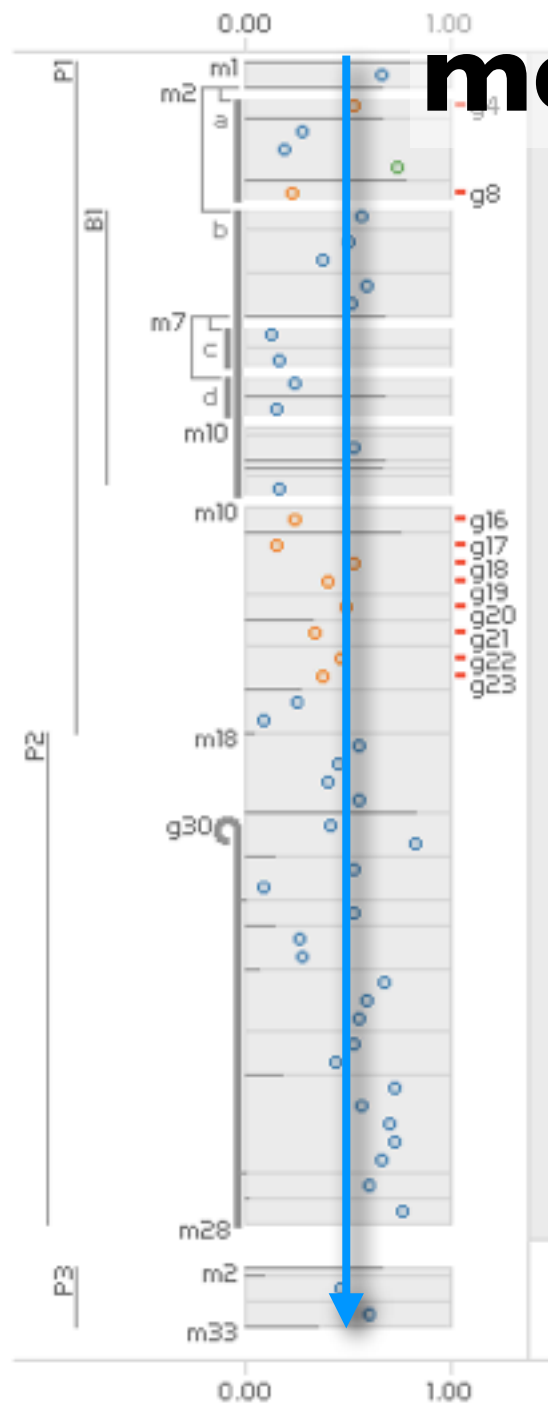
PATHWAY METRIC OVERVIEW

SPECIES

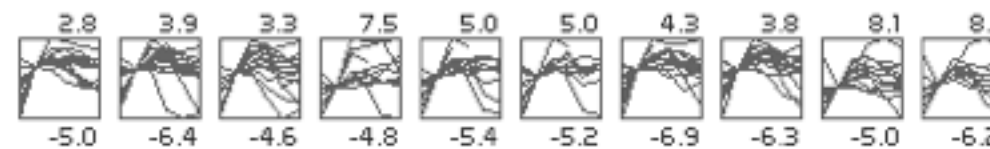
CURVEMAP

OVERLAYS

metabolic pathways



KEY Genes
■ forward ■ reverse ■ bidirectional
Metabolites Metrics
○ PearsonALL



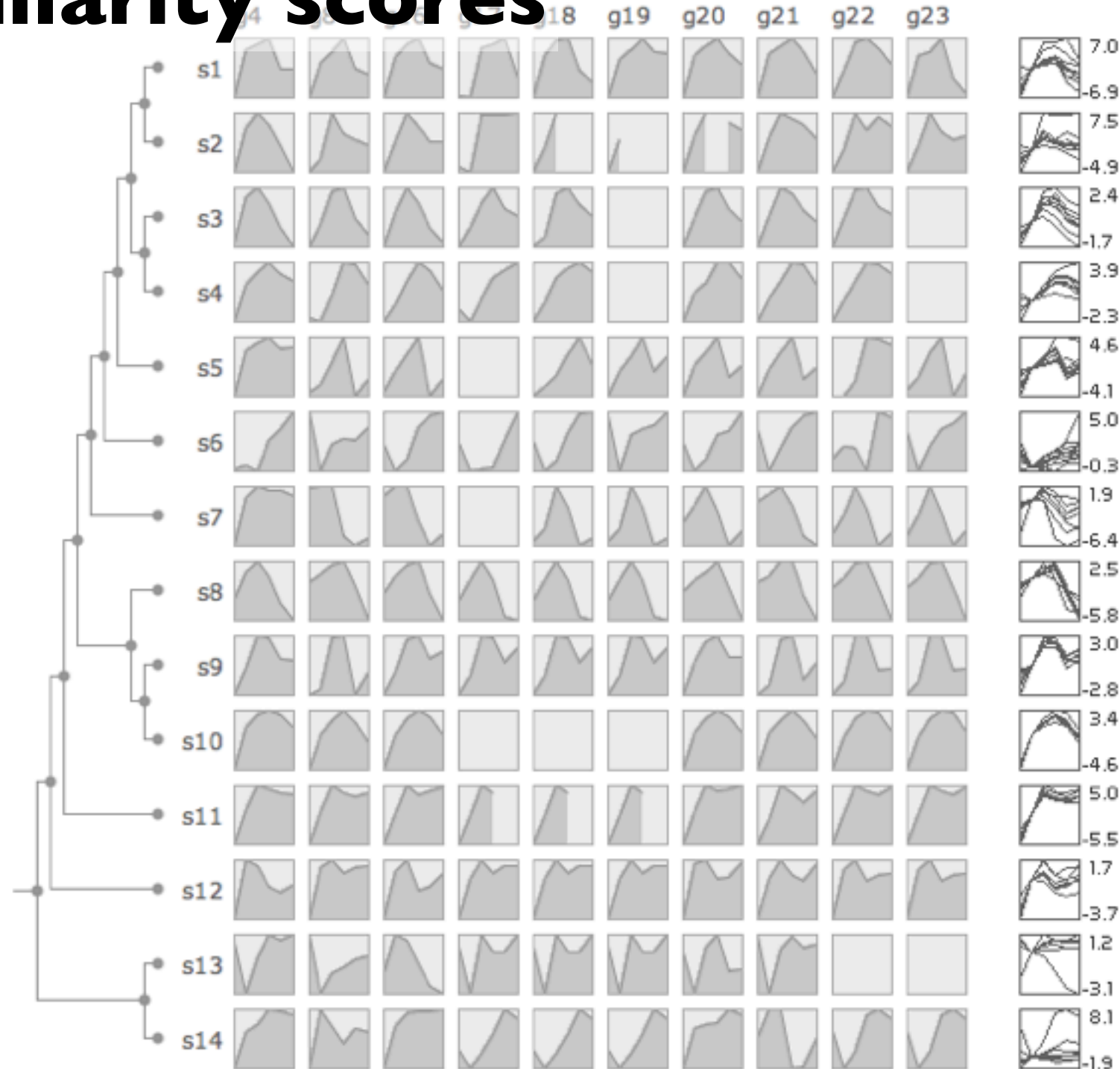
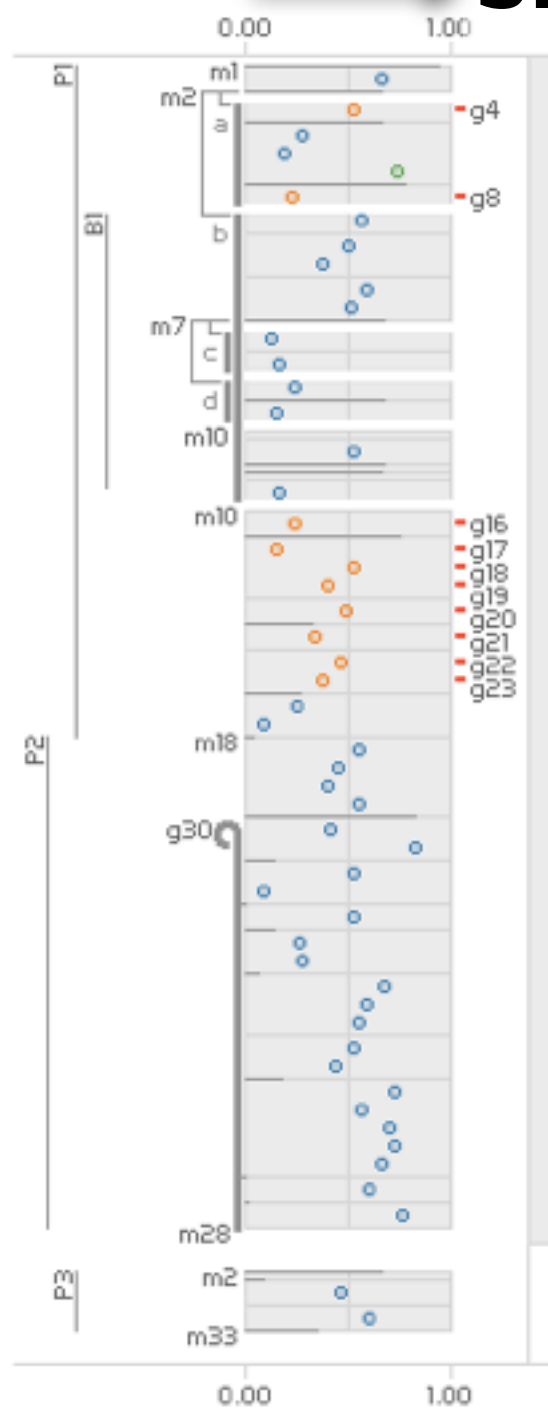
PATHLINE

A TOOL FOR COMPARATIVE FUNCTIONAL GENOMICS

PATHWAY METRIC OVERVIEW

similarity scores

OVERLAYS



KEY Genes
 ■ forward ■ reverse ■ bidirectional

Metabolites
 —

Metrics
 ○ PearsonALL



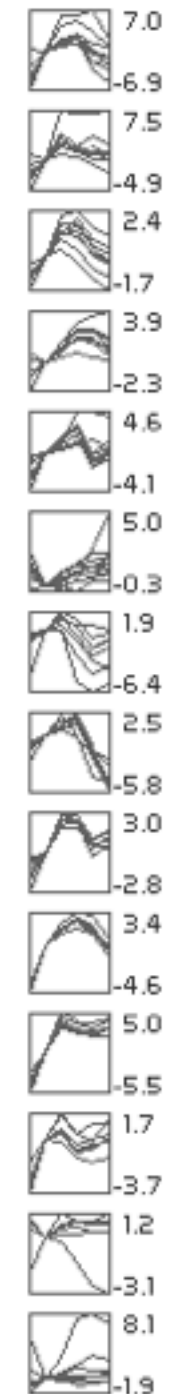
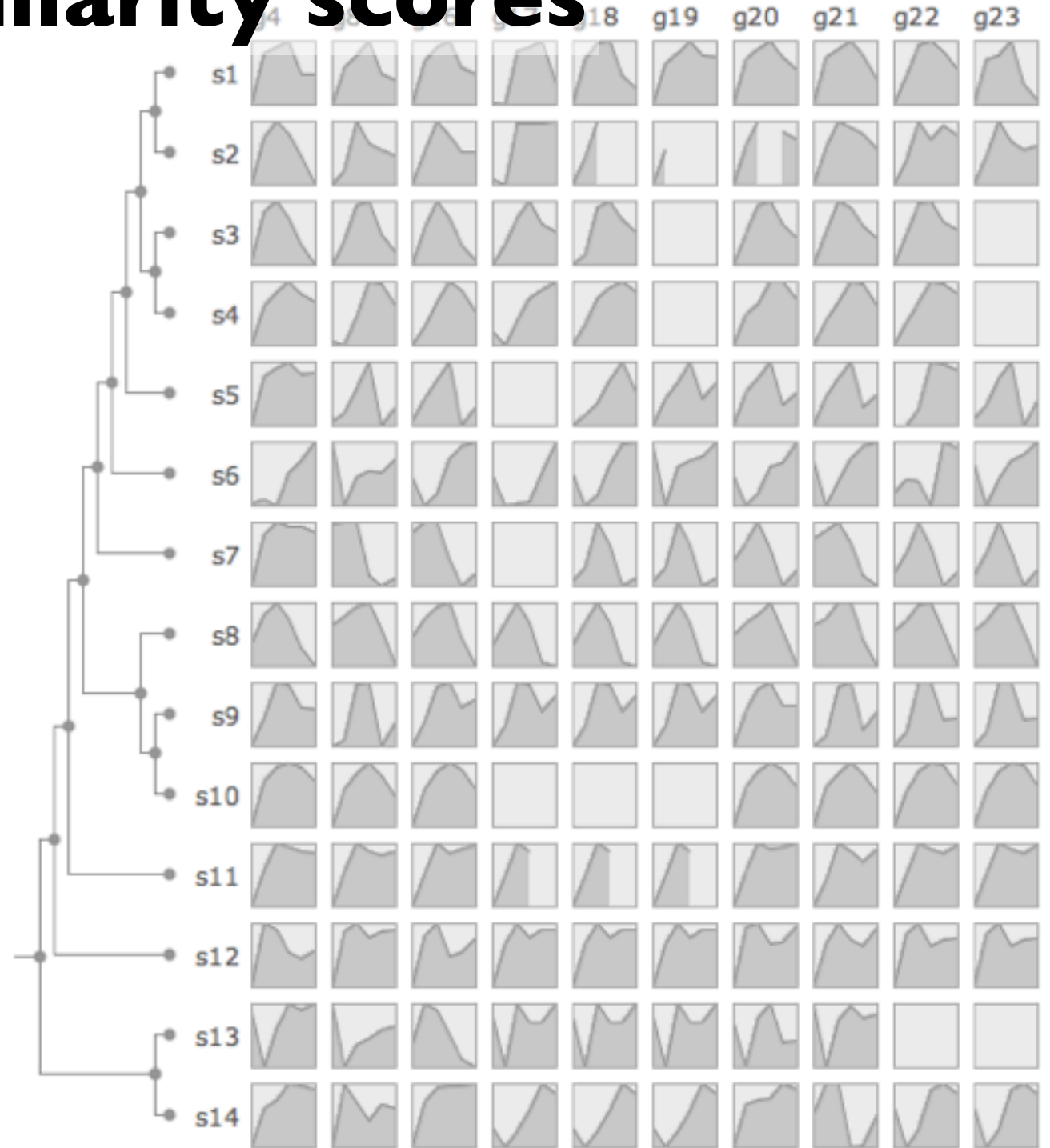
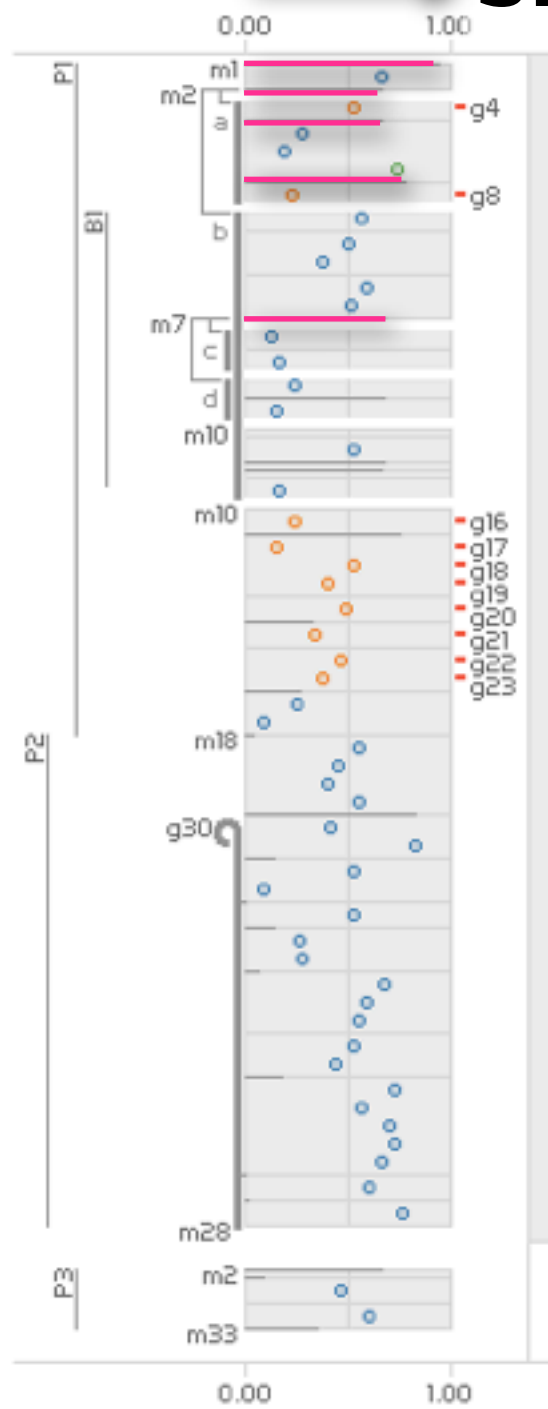
PATHLINE

A TOOL FOR COMPARATIVE FUNCTIONAL GENOMICS

PATHWAY METRIC OVERVIEW

similarity scores

OVERLAYS



KEY Genes
 ■ forward ■ reverse ■ bidirectional
 Metabolites Metrics
 ○ PearsonALL



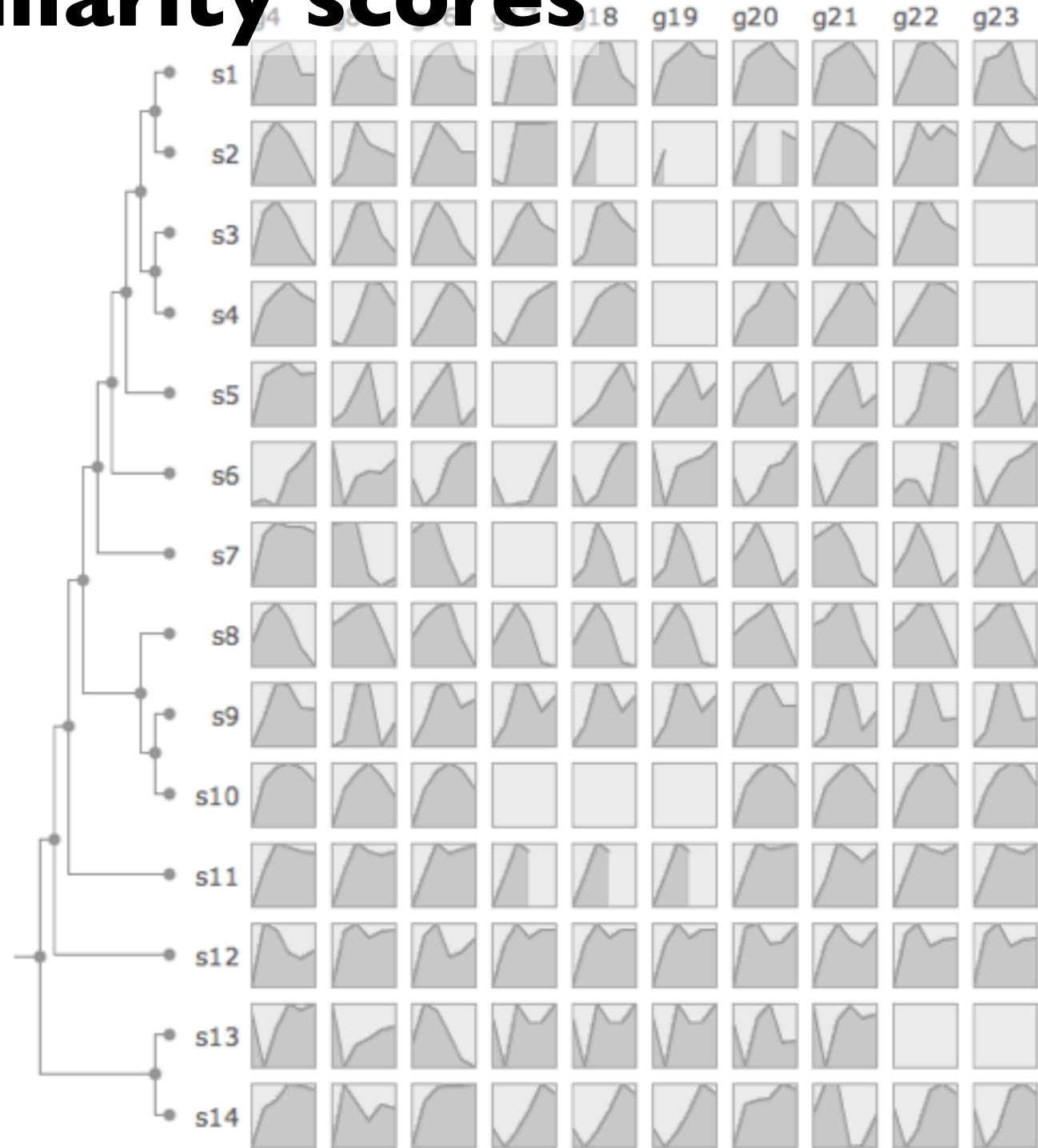
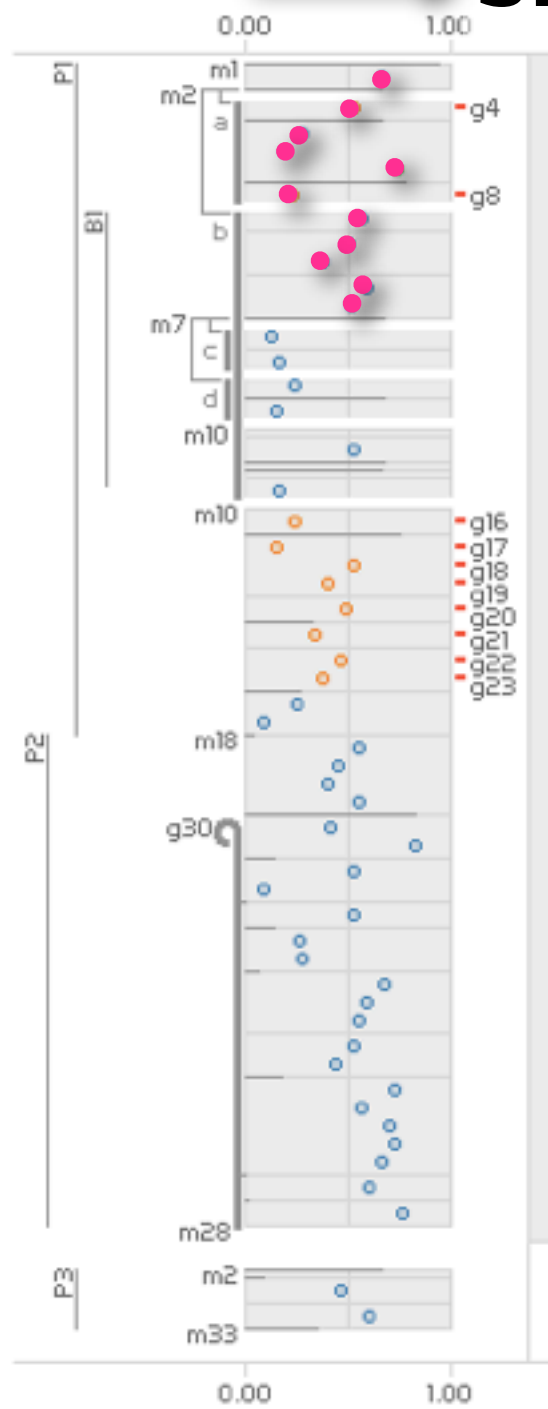
PATHLINE

A TOOL FOR COMPARATIVE FUNCTIONAL GENOMICS

PATHWAY METRIC OVERVIEW

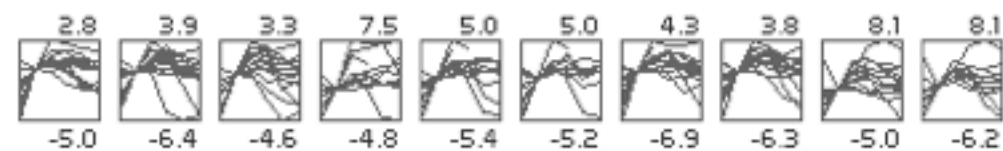
similarity scores

OVERLAYS

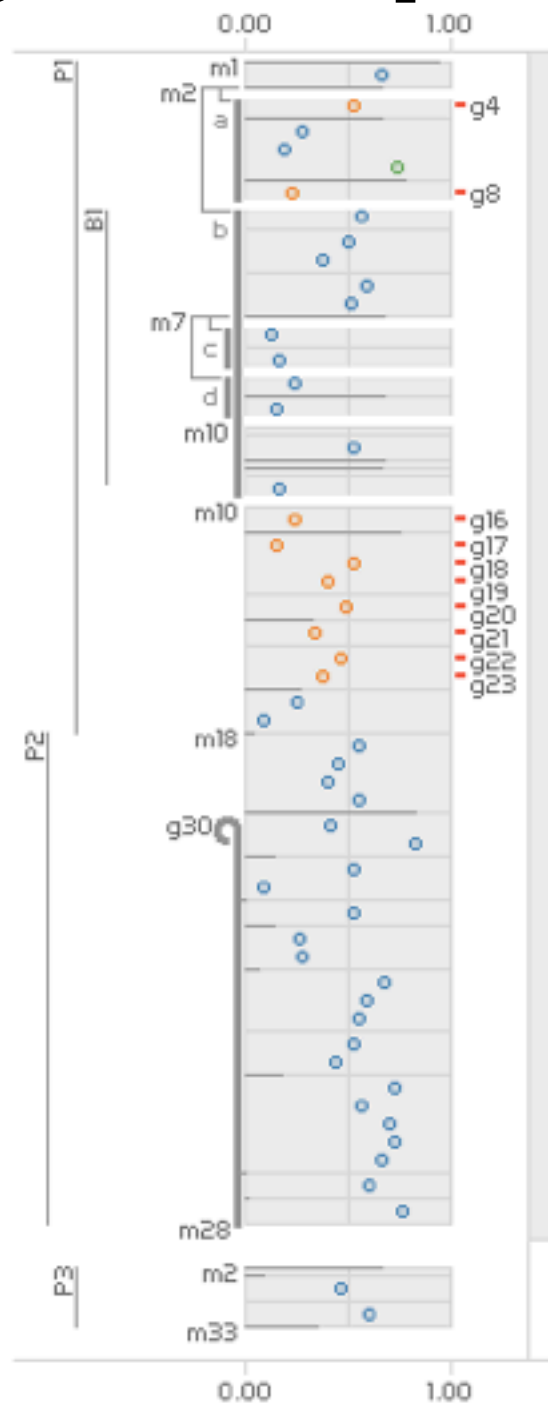


KEY Genes
 ■ forward ■ reverse ■ bidirectional

Metabolites —
Metrics ○ PearsonALL

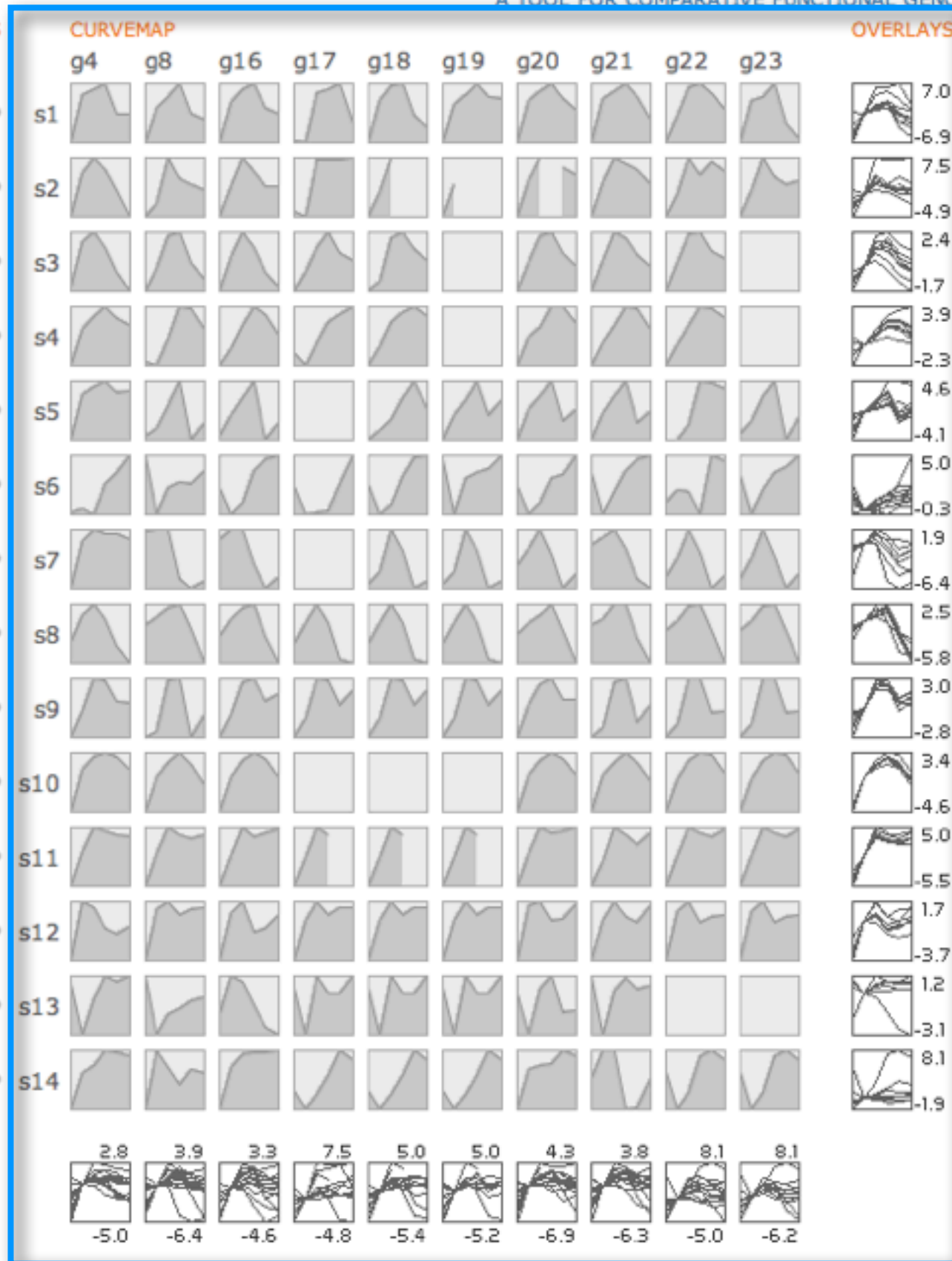


PATHLINE gene expression



KEY Genes
 ■ forward ■ reverse ■ bidirectional

Metabolites Metrics
 ○ PearsonALL

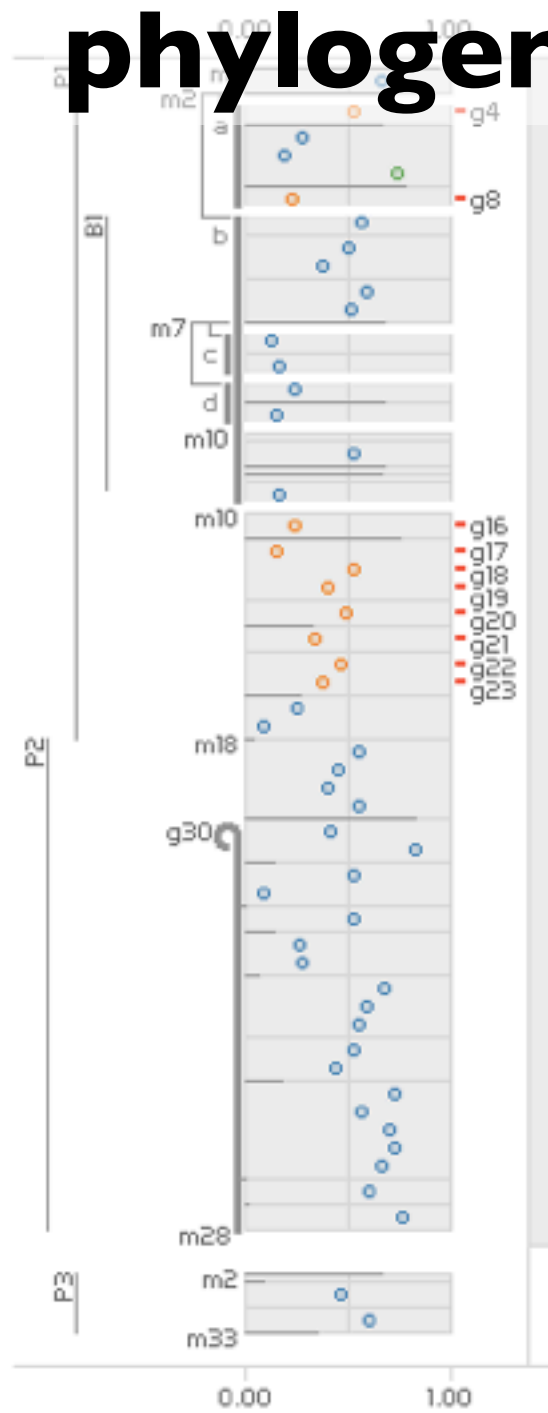


PATHLINE

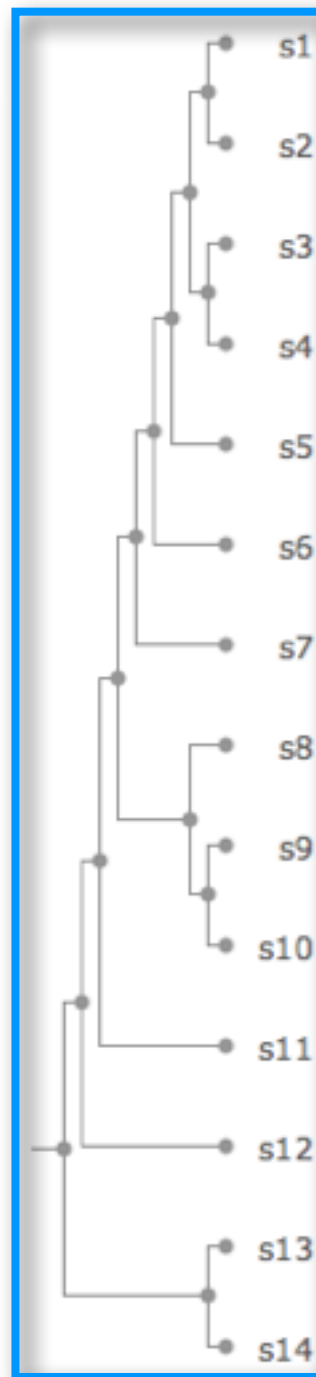
PATHWAY METRIC OVERVIEW

A TOOL FOR COMPARATIVE FUNCTIONAL GENOMICS

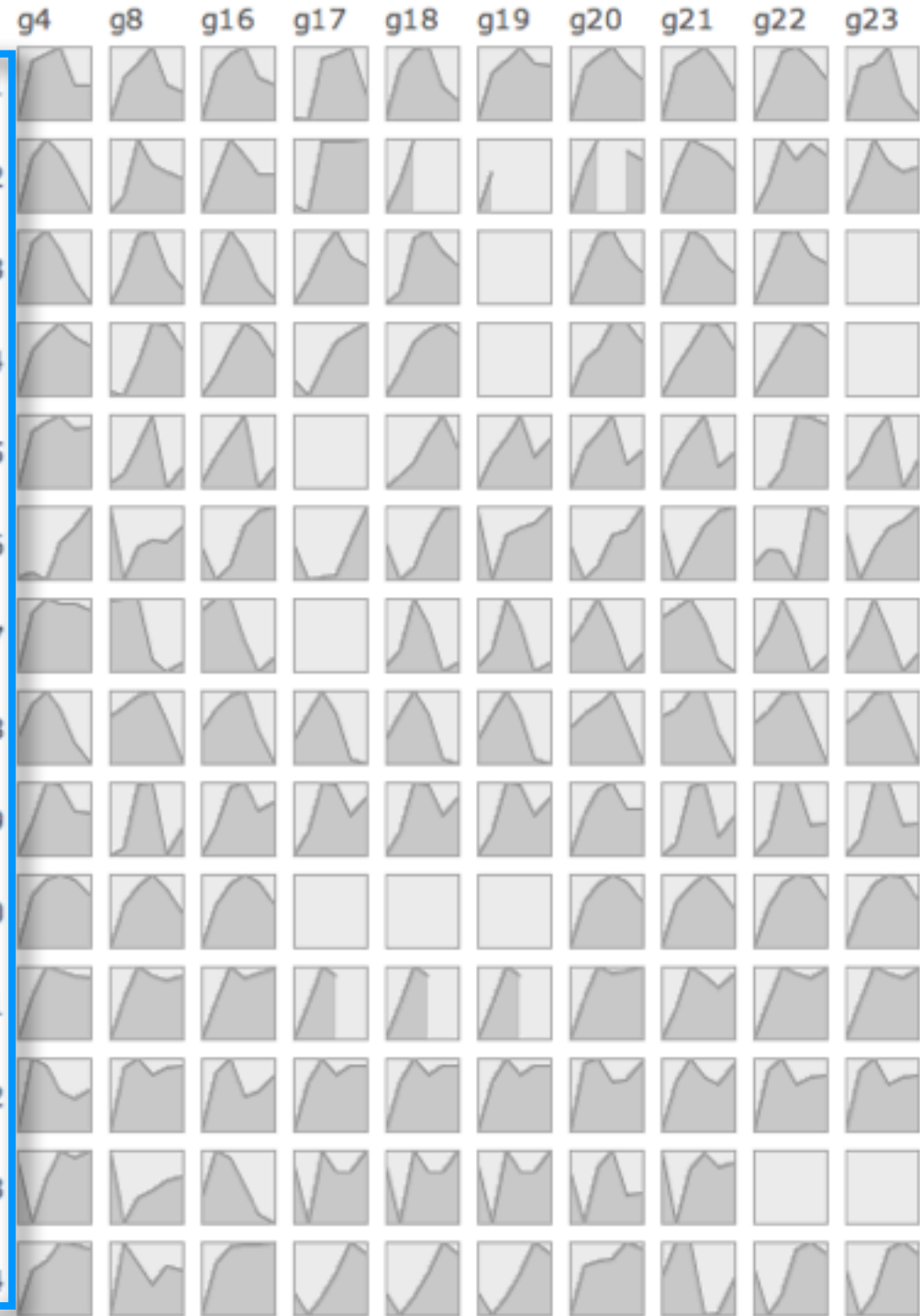
phylogeny



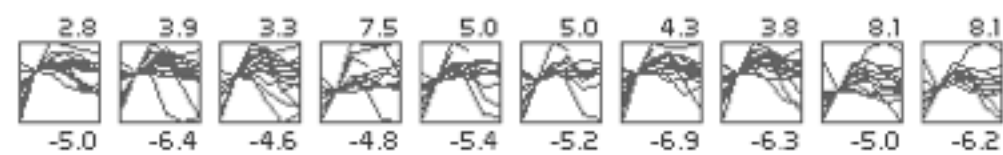
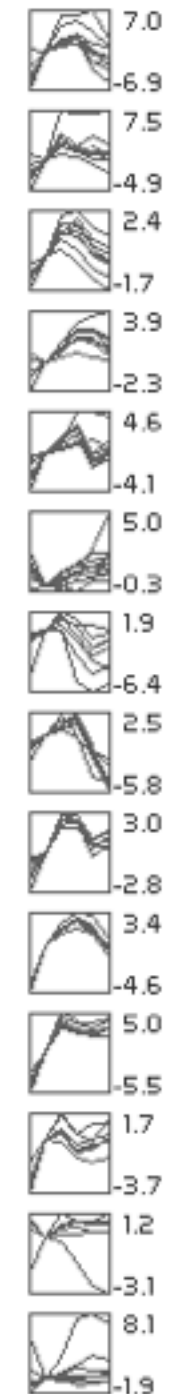
SPECIES



CURVEMAP



OVERLAYS



target

translate

design

implement

validate



color

A row of four colored circles: blue, green, yellow, and red.

volume

Two 3D cubes, one larger than the other.

area

Two circles, one larger than the other.

angle

Two right-angled corners, one larger than the other.

length

Two horizontal lines, one longer than the other.

position

A horizontal line with two points above it.

density

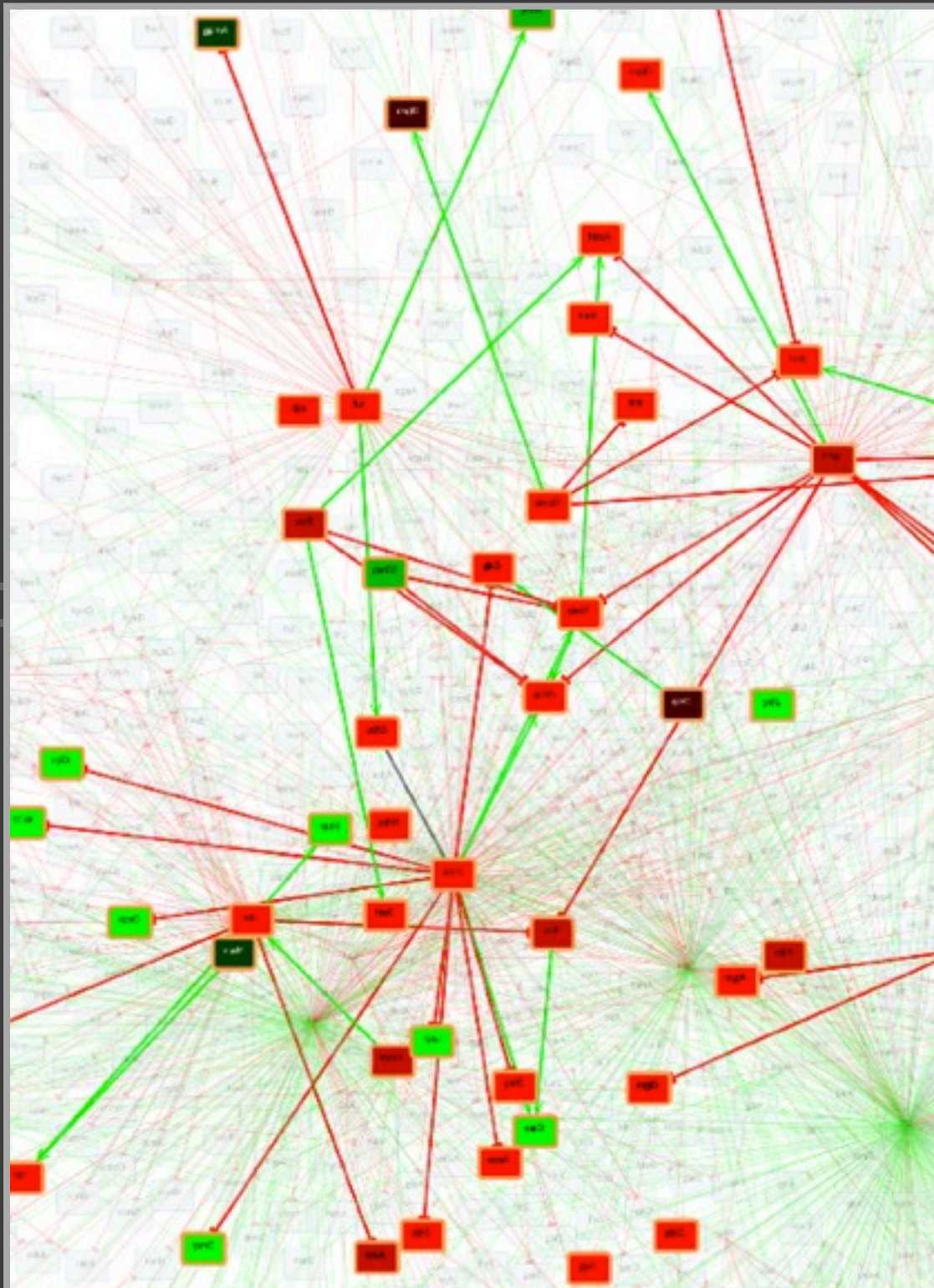
A row of four circles of increasing gray shade from left to right.

slope

Two lines with different slopes, one steeper than the other.

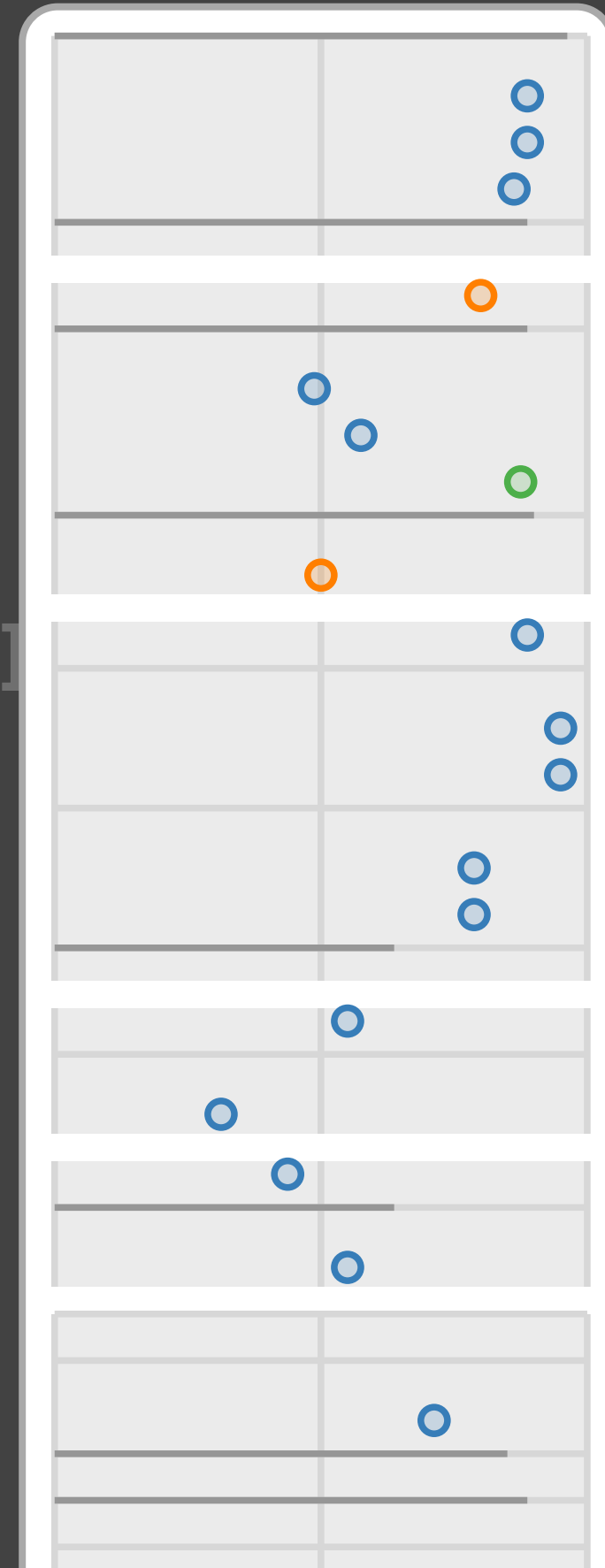
encode quantitative values with spatial position

topological layout

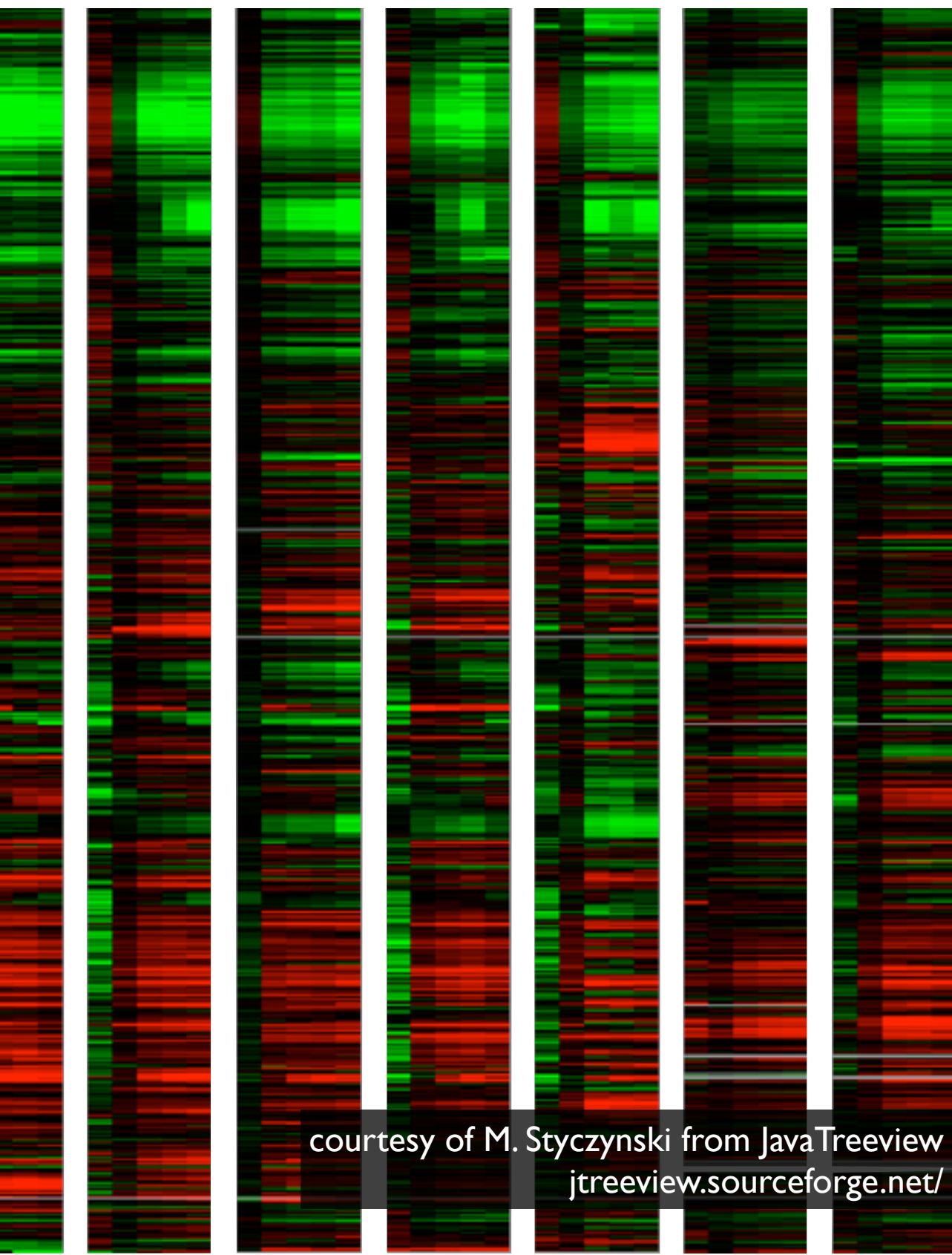


www.win.tue.nl/~mwestenb/genevis/

linearized pathway

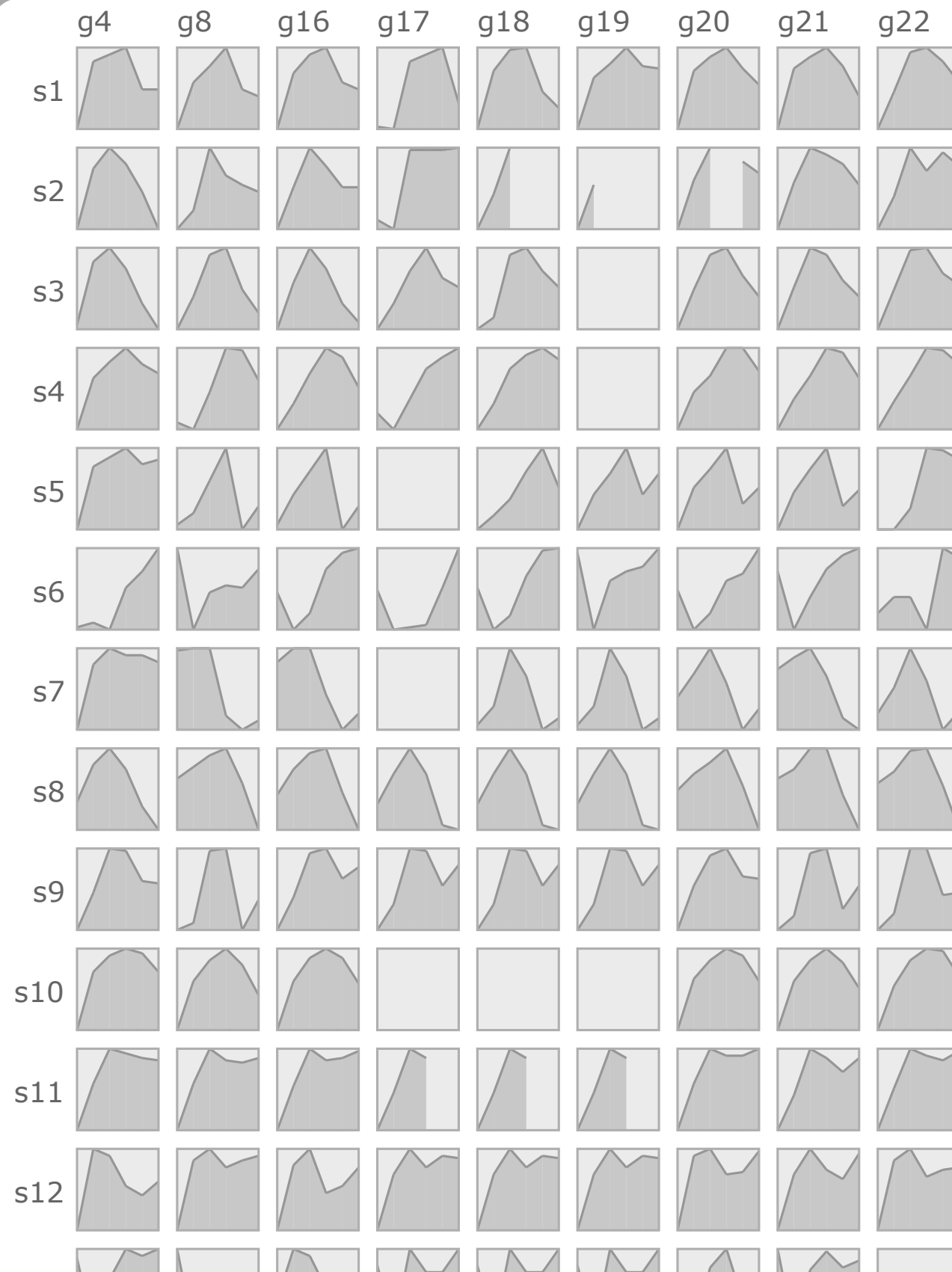


heatmap



courtesy of M. Styczynski from JavaTreeView
jtreeview.sourceforge.net/

curvemap



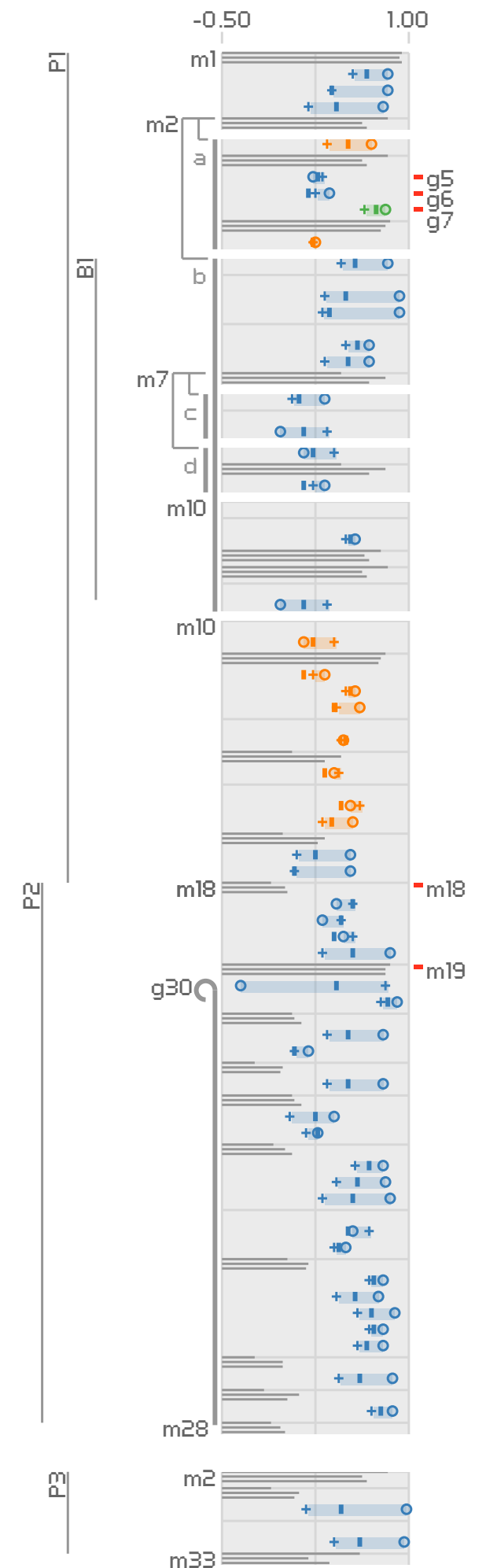
Pathline

linearized pathway representation

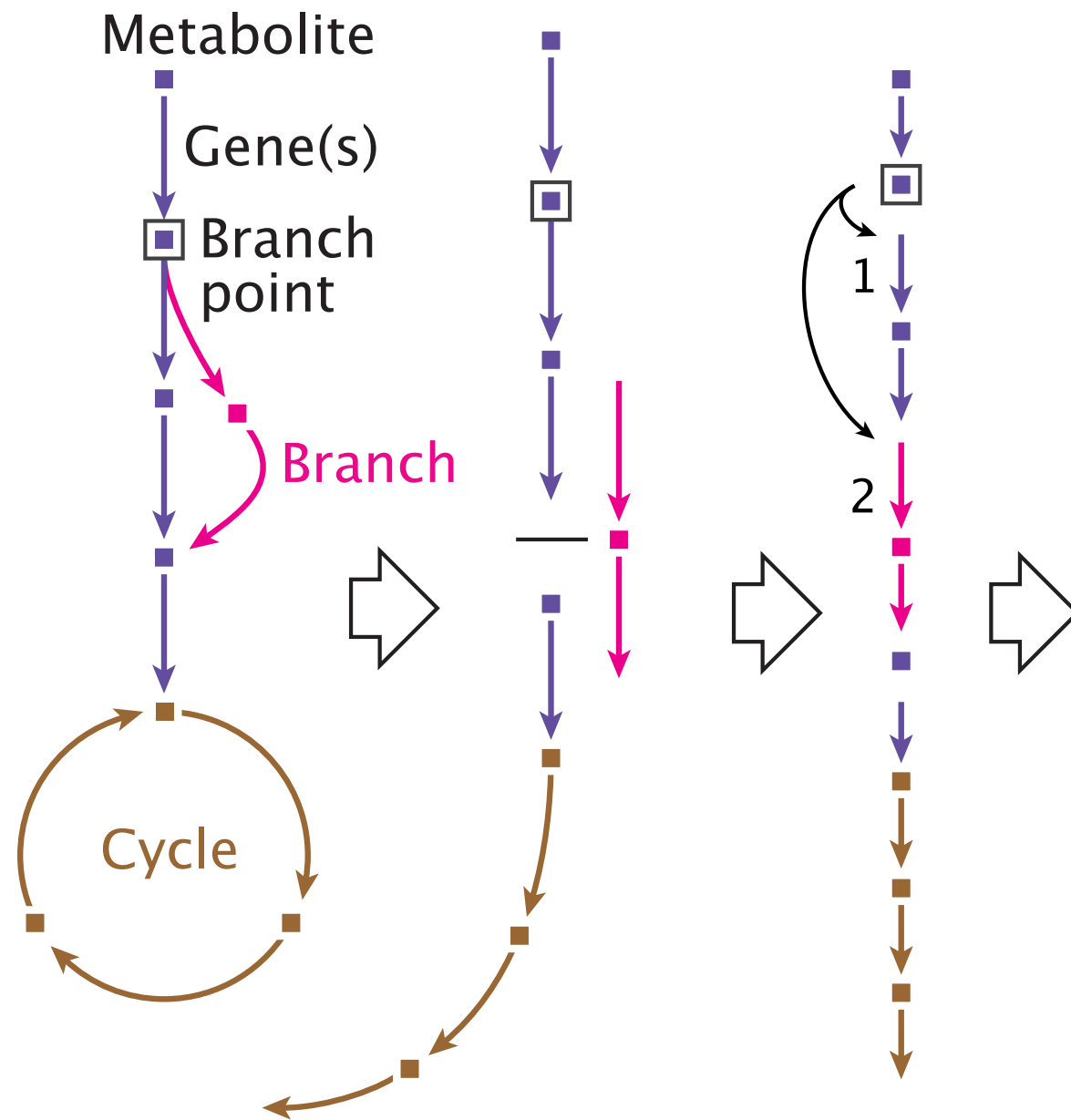
linearized pathway representation

common axes to compare similarity scores

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

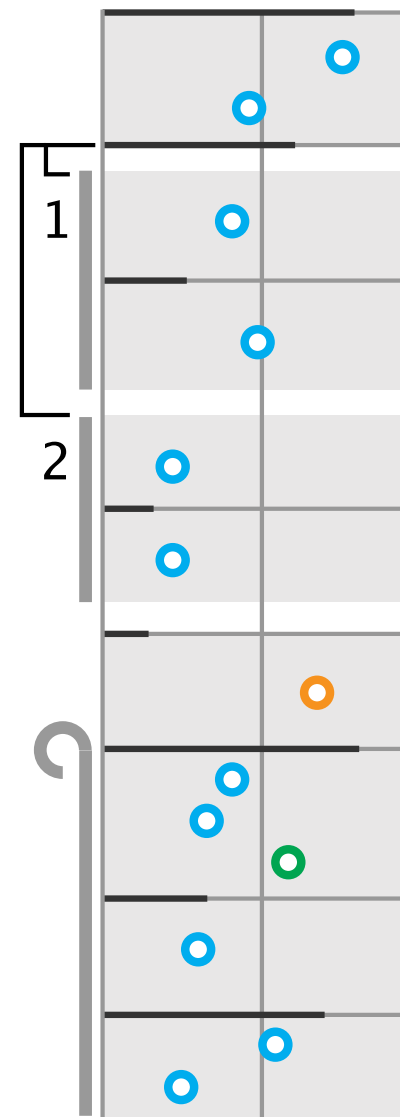


pathway to ordered list of nodes



unroll and cut

reinsert

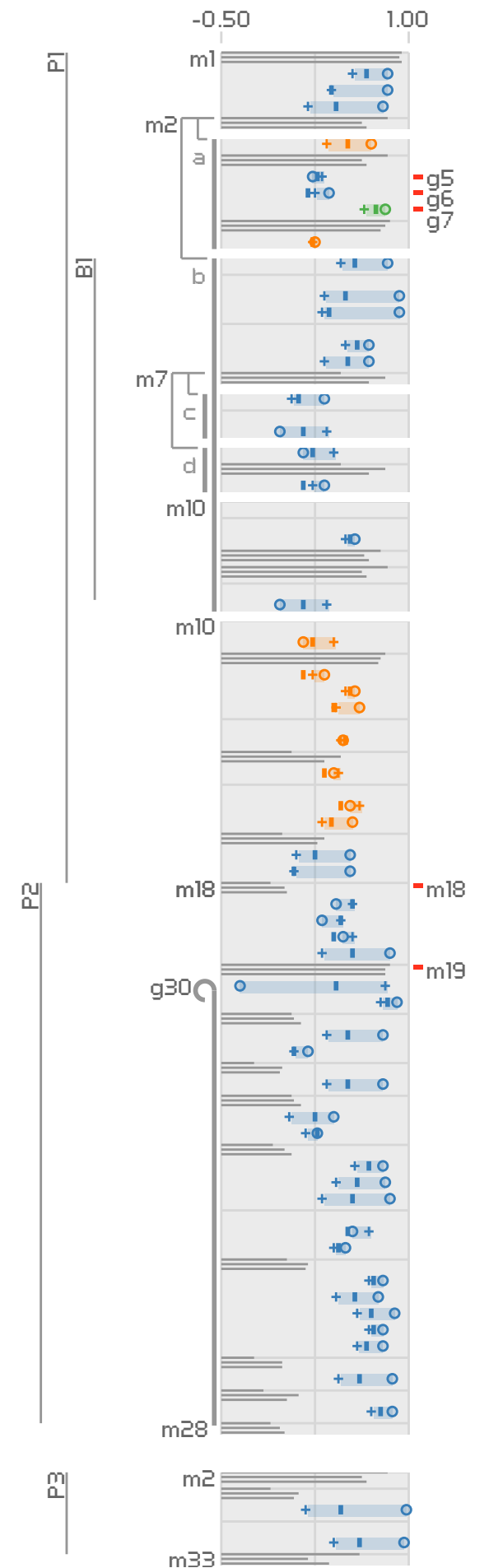


**shared coordinate frame
and stylized marks**

linearized pathway representation

putting it together . . .

- use spatial position for similarity scores
- topology is secondary

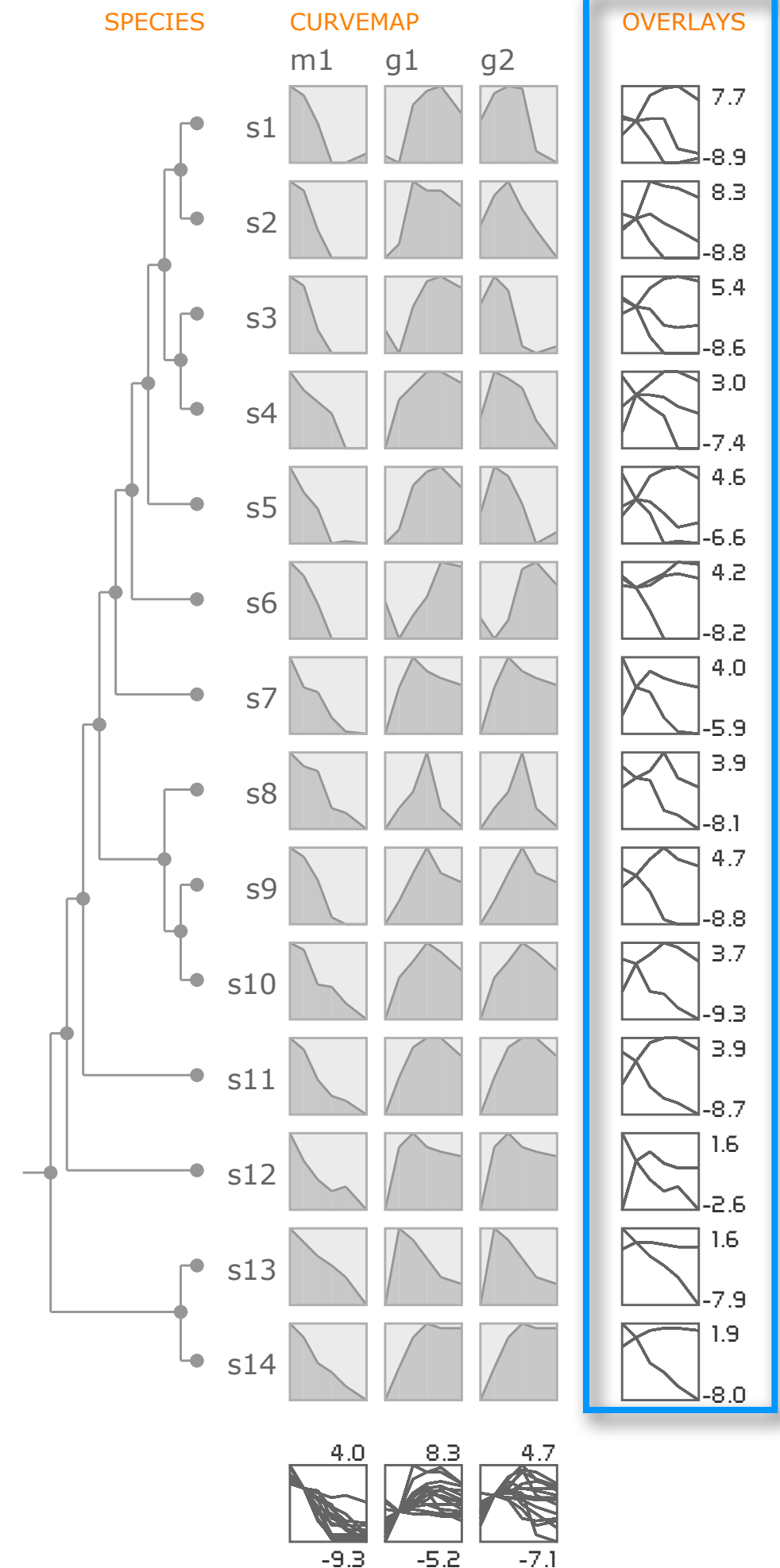


Pathline
curvemap

curvemap

inspired by heatmaps

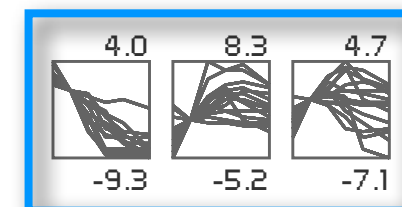
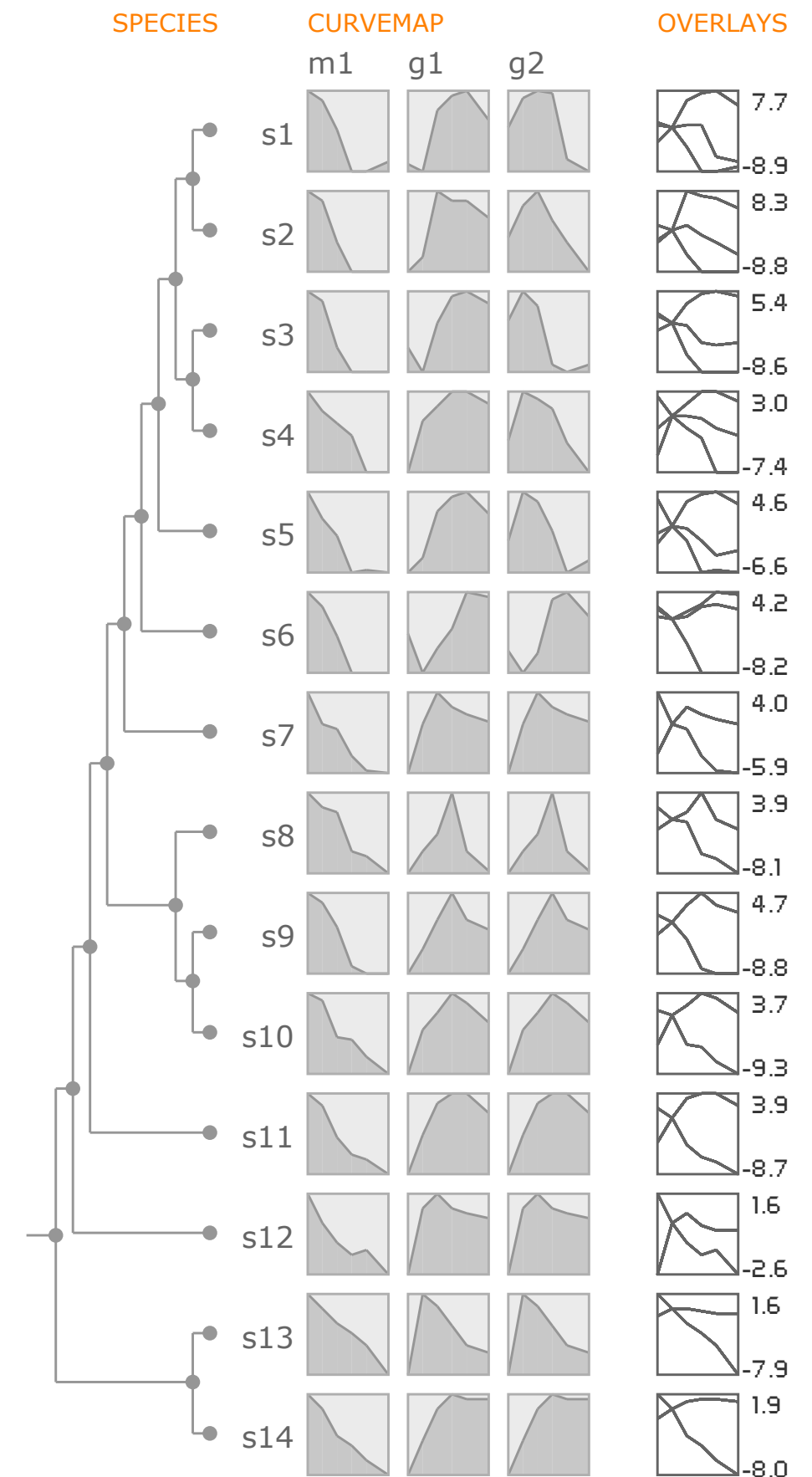
- base visual unit is a curve
- filled, framed line charts to enhance shape perception
- rows are species
- columns are genes/metabolites
- overlays to enhance trends



curvemap

inspired by 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



target

translate

design

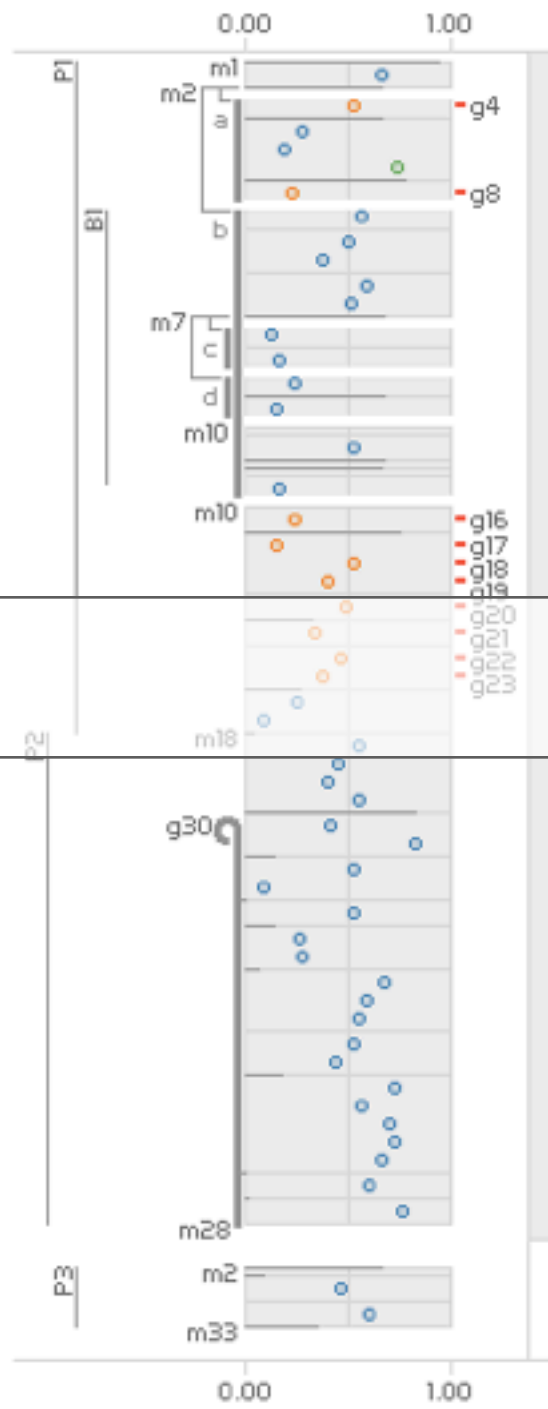
implement

validate

PATHLINE

A TOOL FOR COMPARATIVE FUNCTIONAL GENOMICS

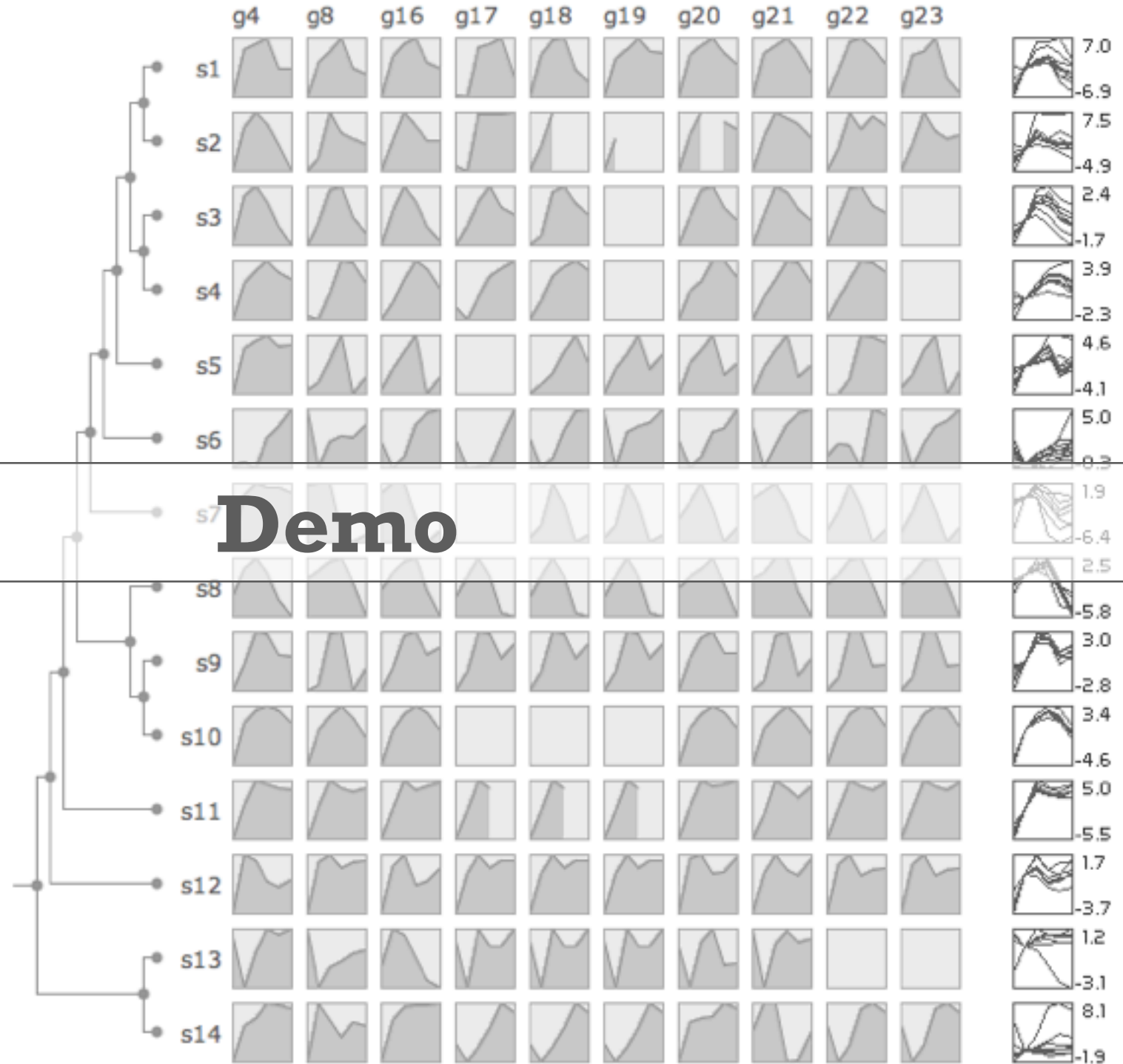
PATHWAY METRIC OVERVIEW



SPECIES

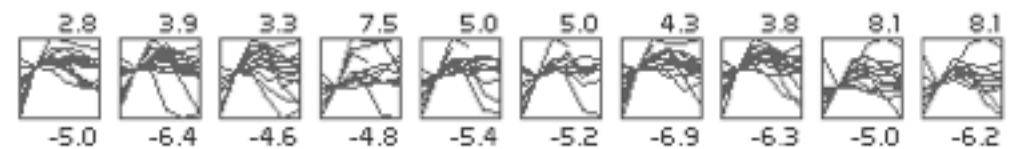
CURVEMAP

OVERLAYS



Demo

KEY Genes
 ■ forward ■ reverse ■ bidirectional
 Metabolites Metrics
 ○ PearsonALL



target

translate

design

implement

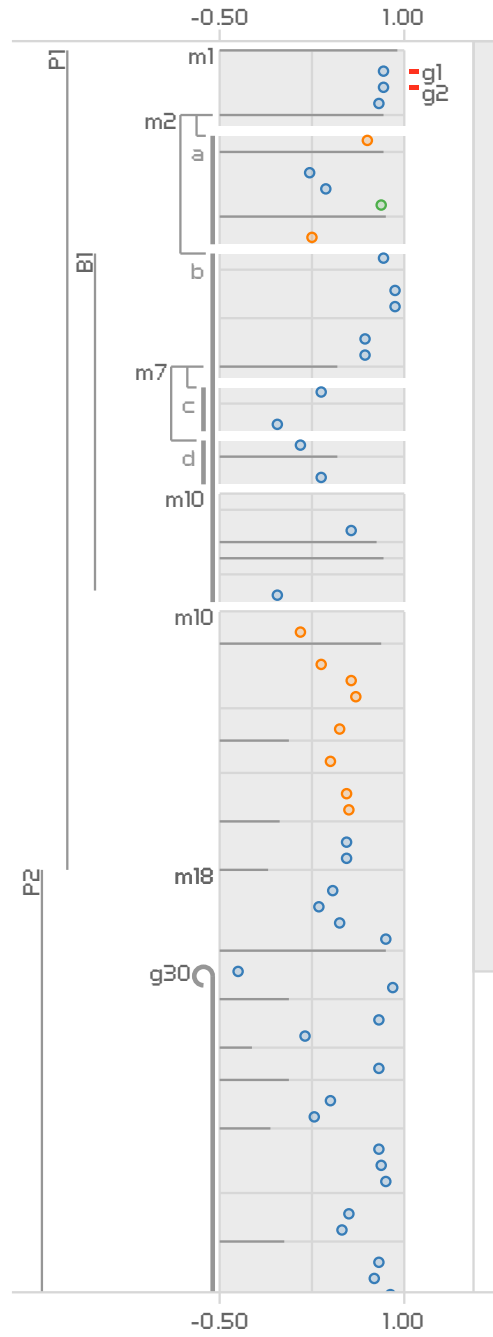
validate

case study

- qualitative research method
- in-depth study of individual or group
- real-world setting
- description and interpretation

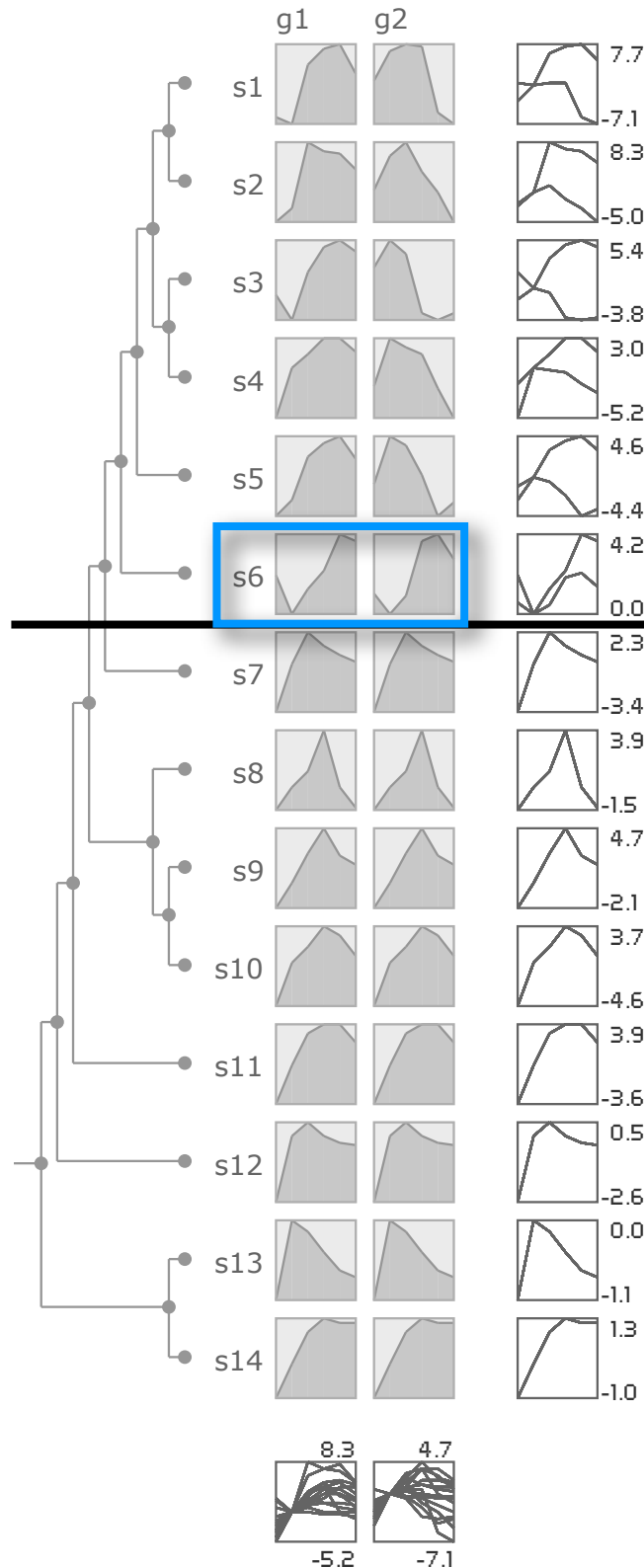
PATHLINE

PATHWAY METRIC OVERVIEW



A TOOL FOR COMPARATIVE FUNCTIONAL GENOMICS

SPECIES CURVEMAP OVERLAYS



both genes
one gene

KEY Genes
 ■ forward ■ reverse ■ bidirectional

Metabolites
 —

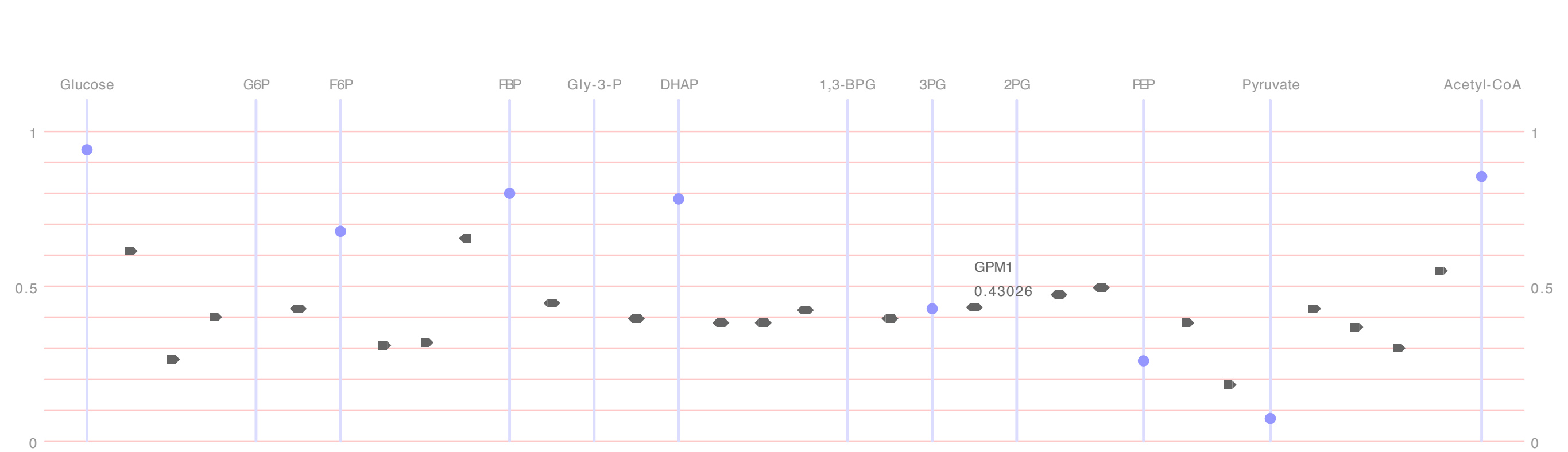
Metrics
 ○ PearsonSubgroup1
 + PearsonSubgroup2
 † PearsonALL

whole genome duplication

www.pathline.org

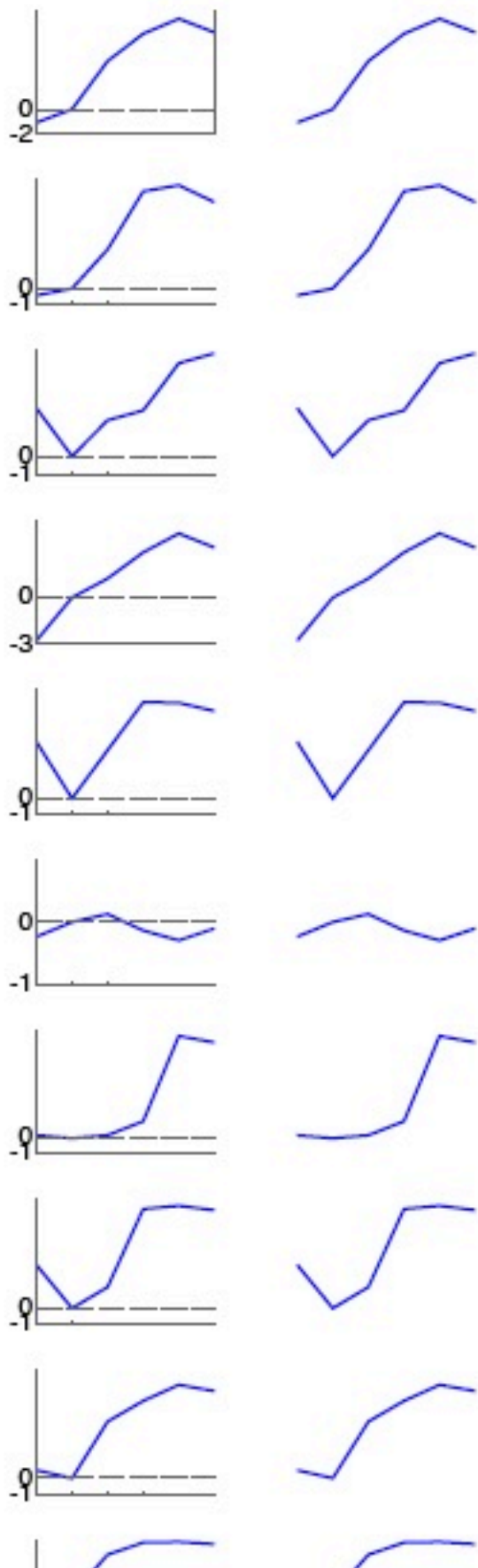
LESSONS LEARNED

- **process supports efficient development**
- **collaborators' time commitment is front loaded**
- **rapid prototyping is essential**

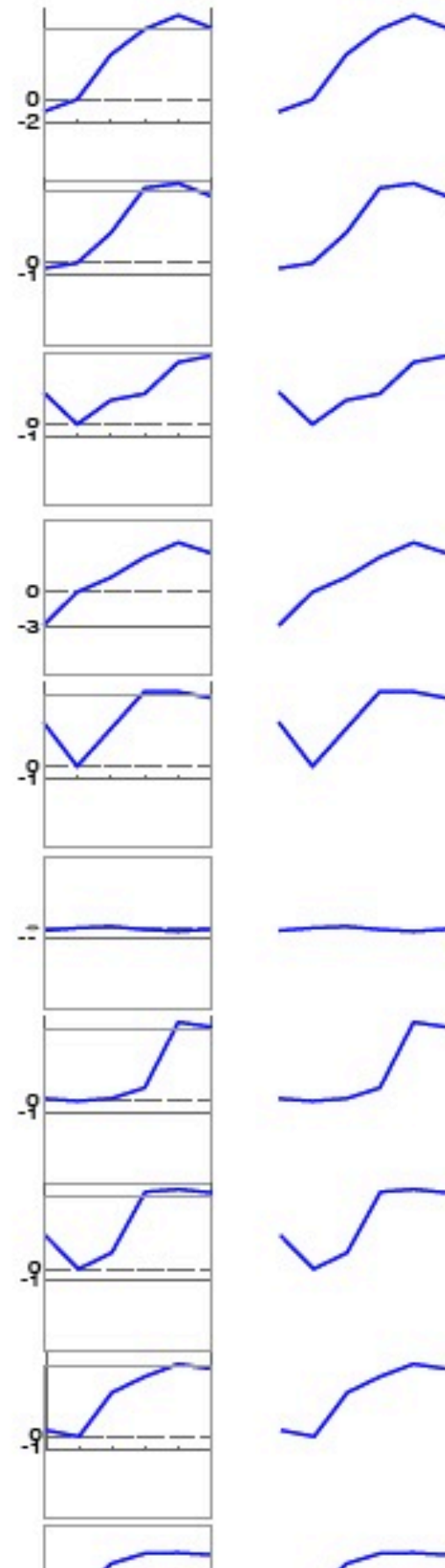




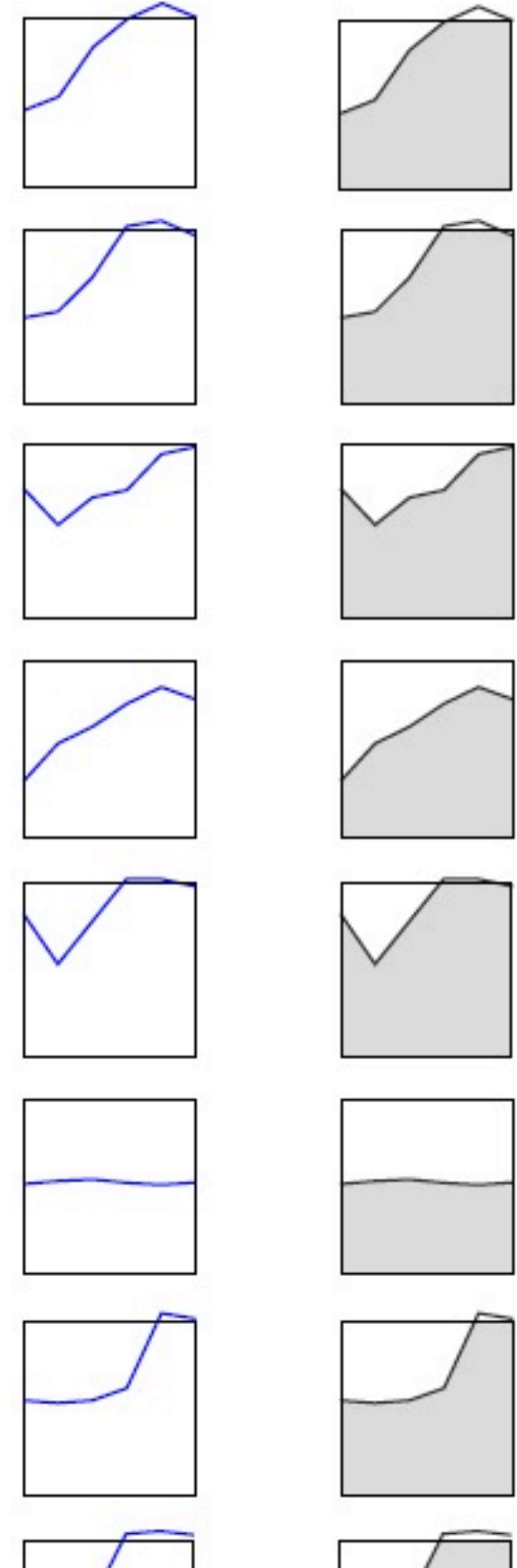
Relative scale

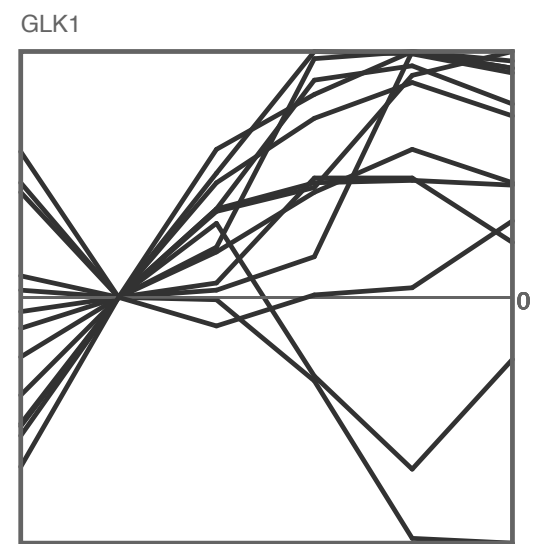
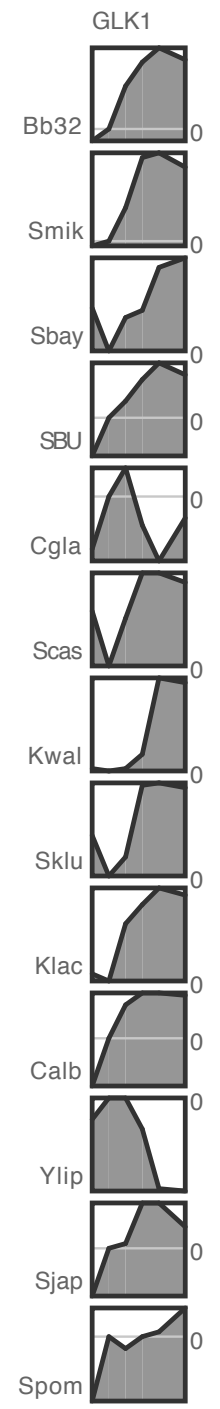
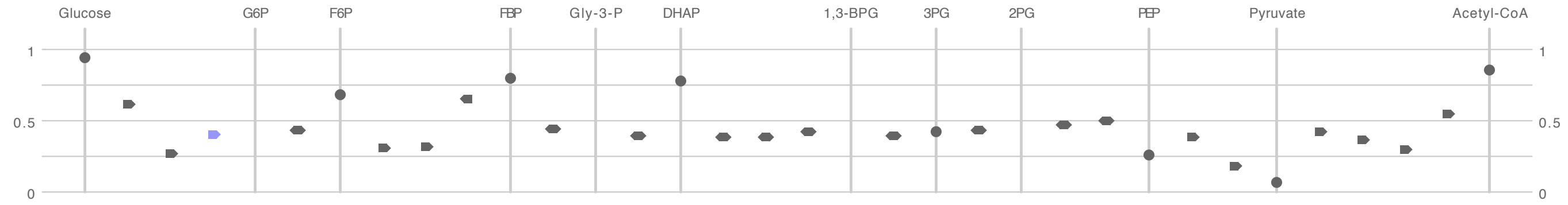


Absolute scale



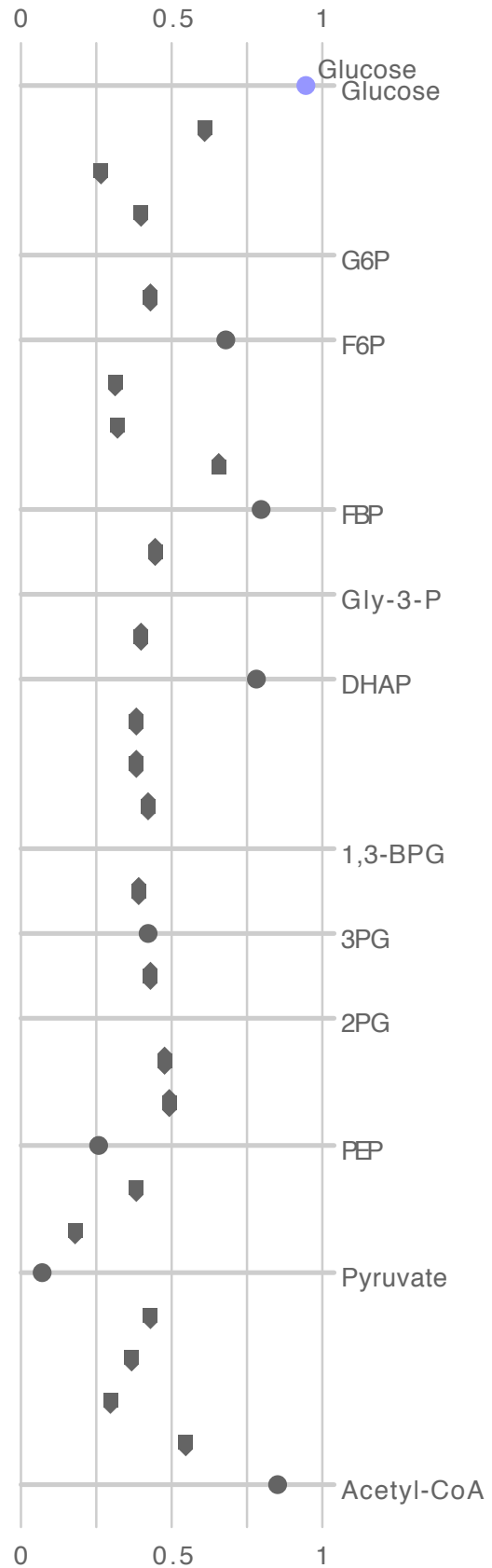
Absolute scale, highlight pattern



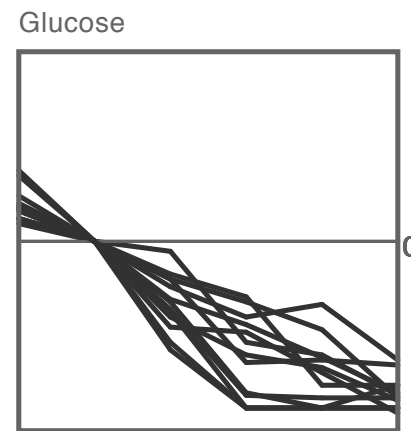
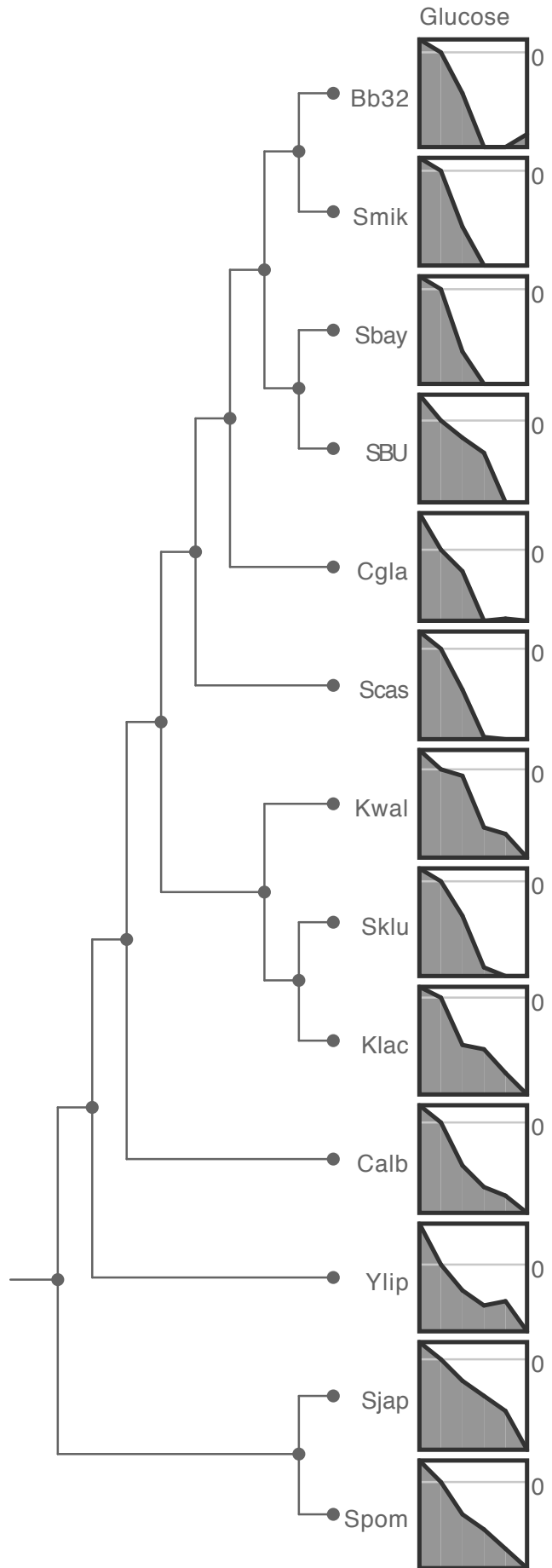


normalized ■
absolute

GLYCOLYSIS

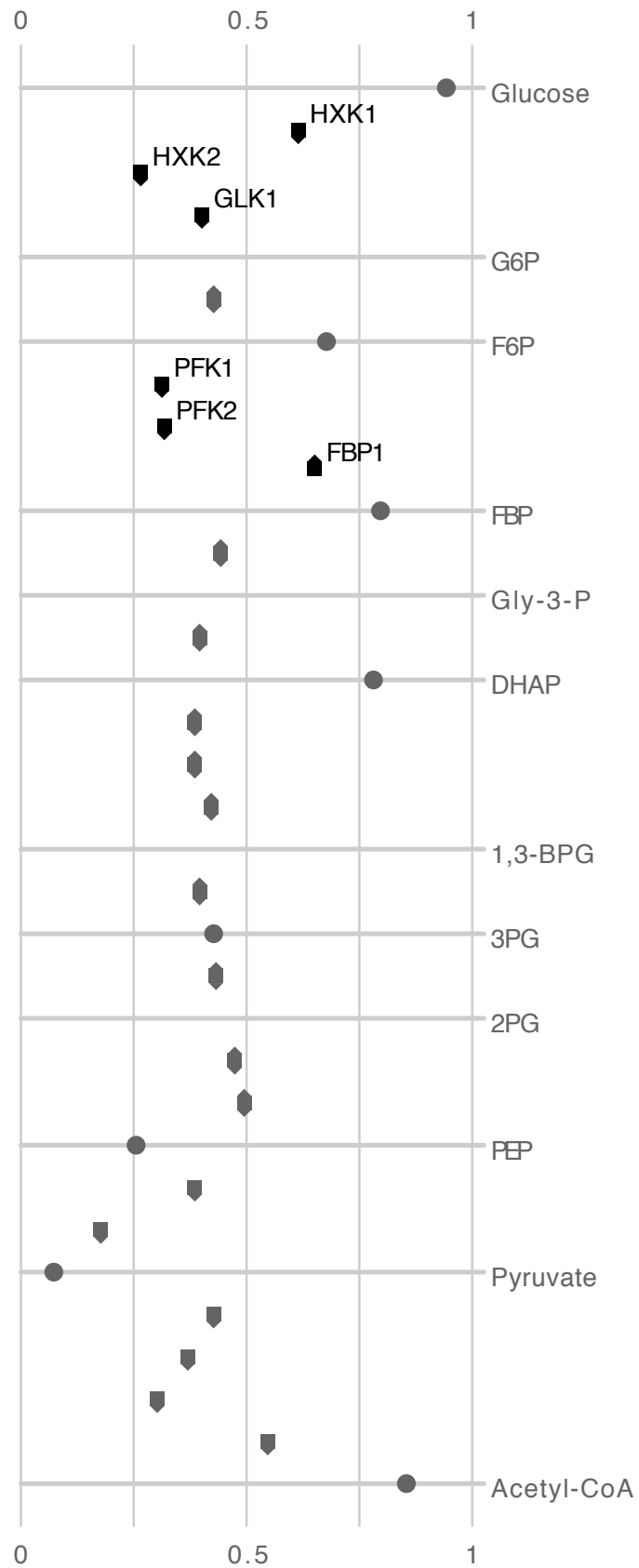


- ▾ forward enzyme
- ▲ reverse enzyme
- ◆ bidirectional enzyme
- metabolite

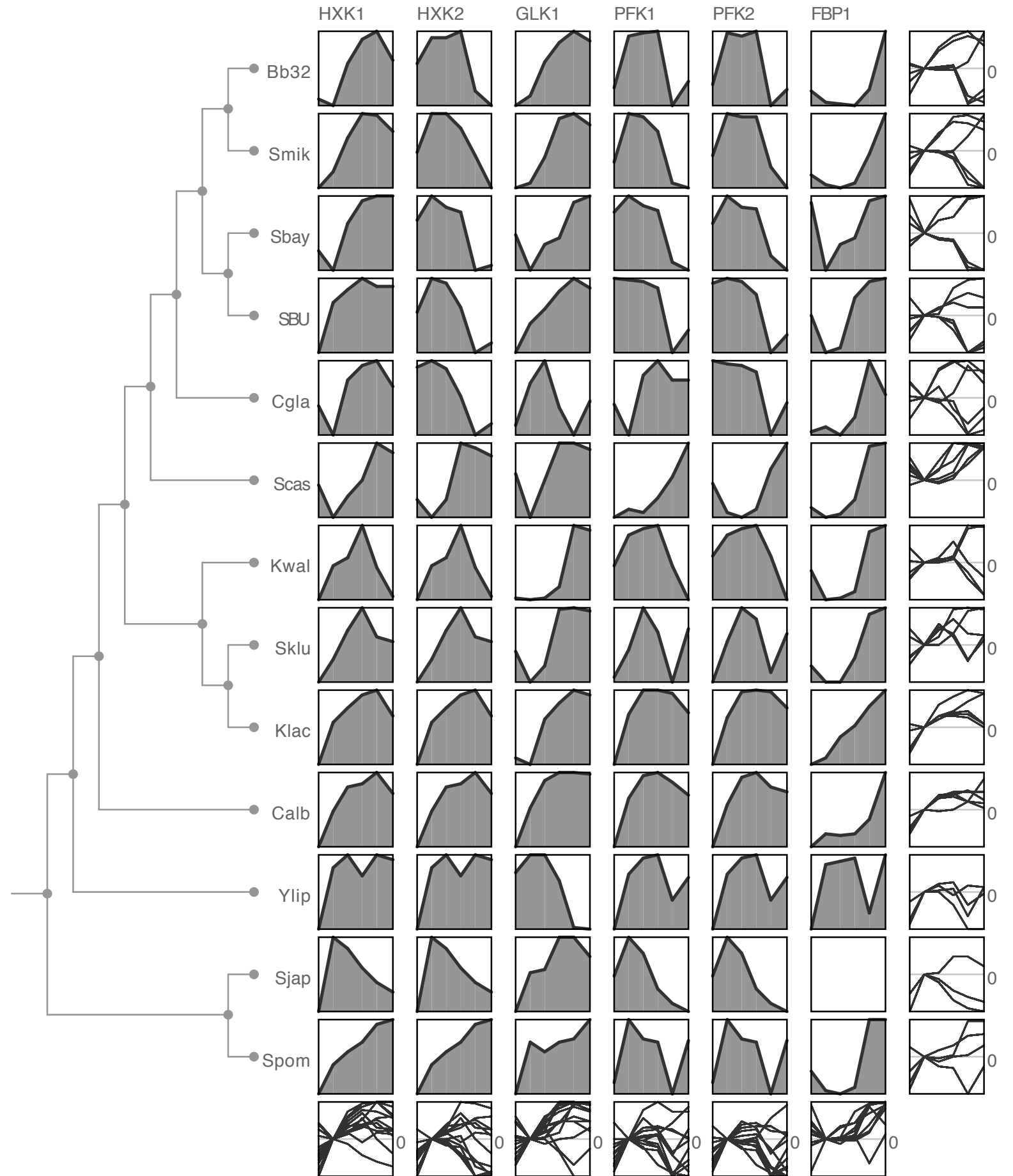


- normalized ■
- absolute

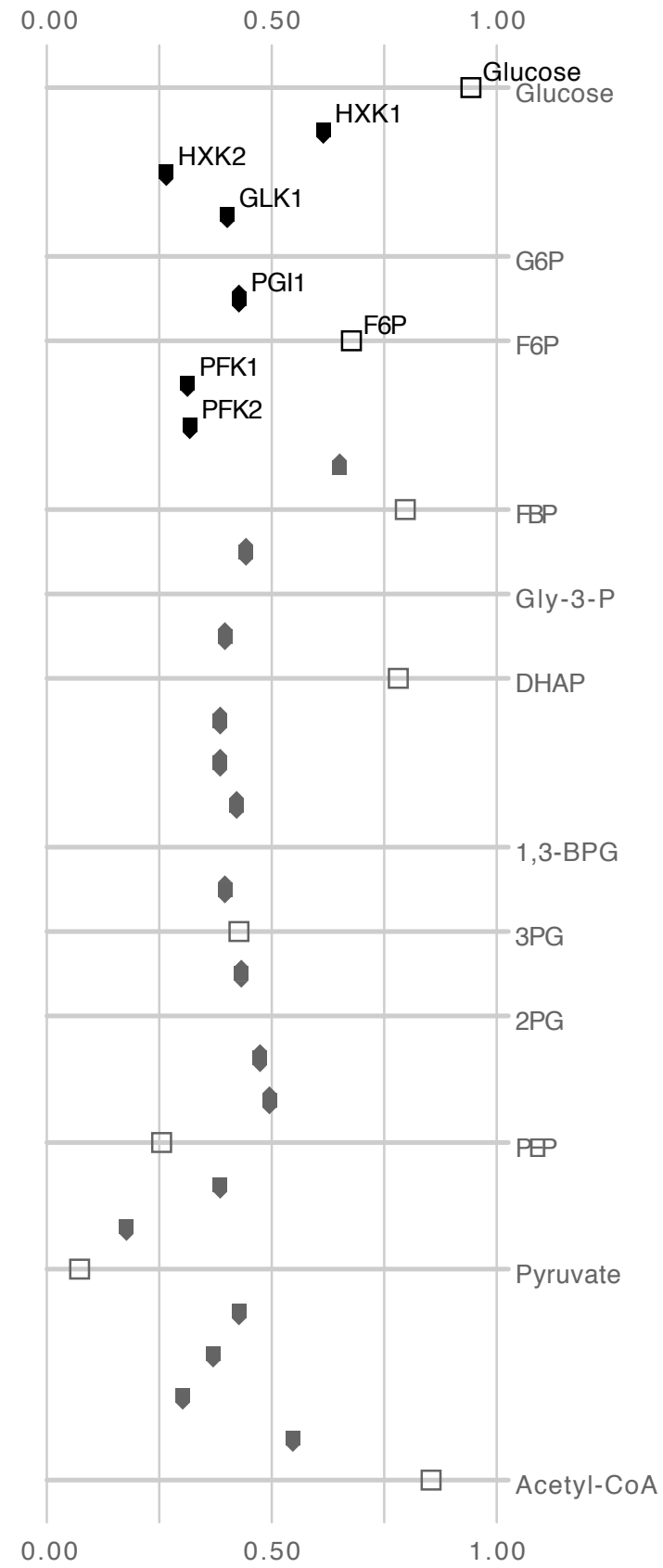
GLYCOLYSIS



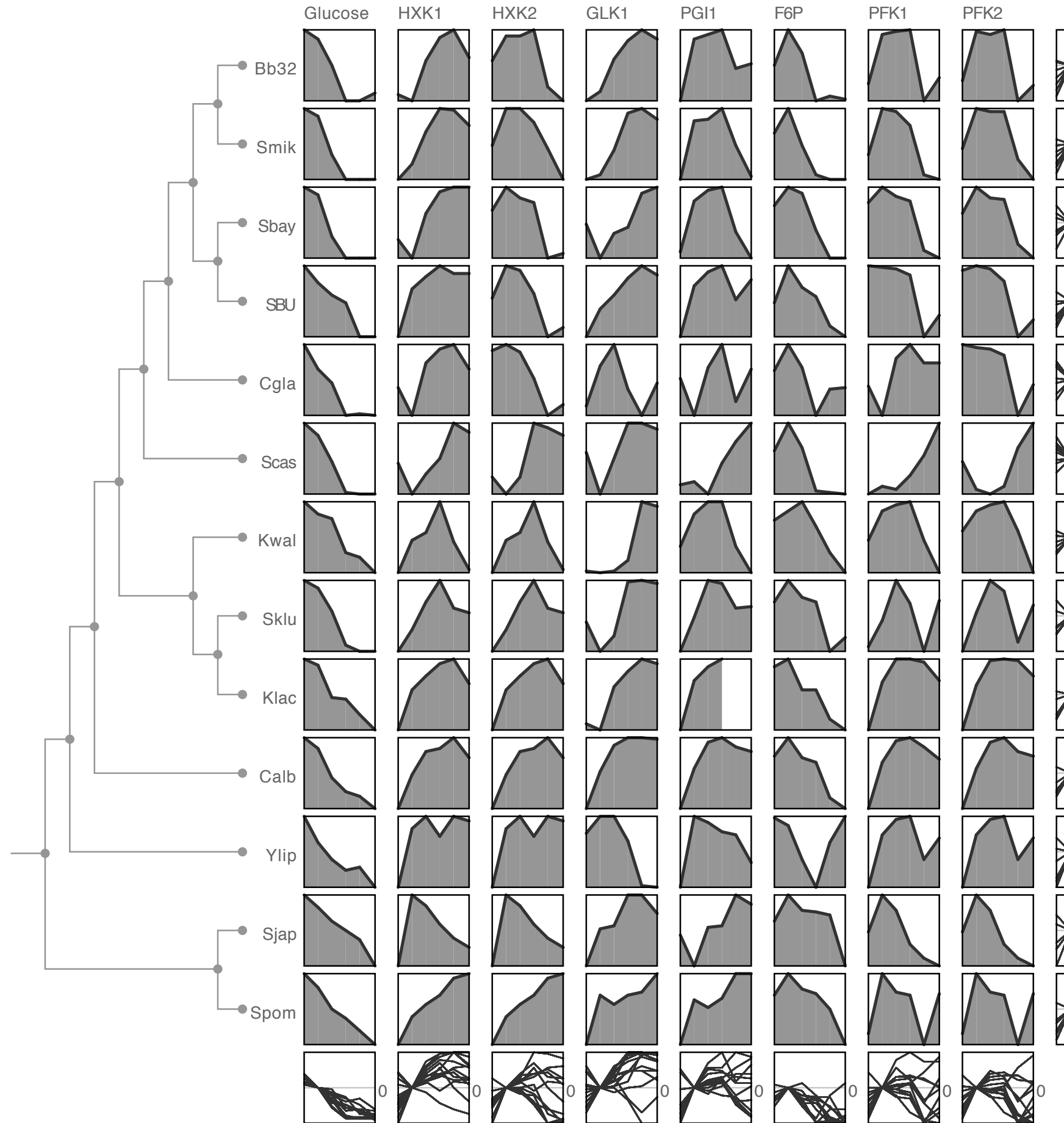
- ▾ forward enzyme
- ▴ reverse enzyme
- ◈ bidirectional enzyme
- metabolite



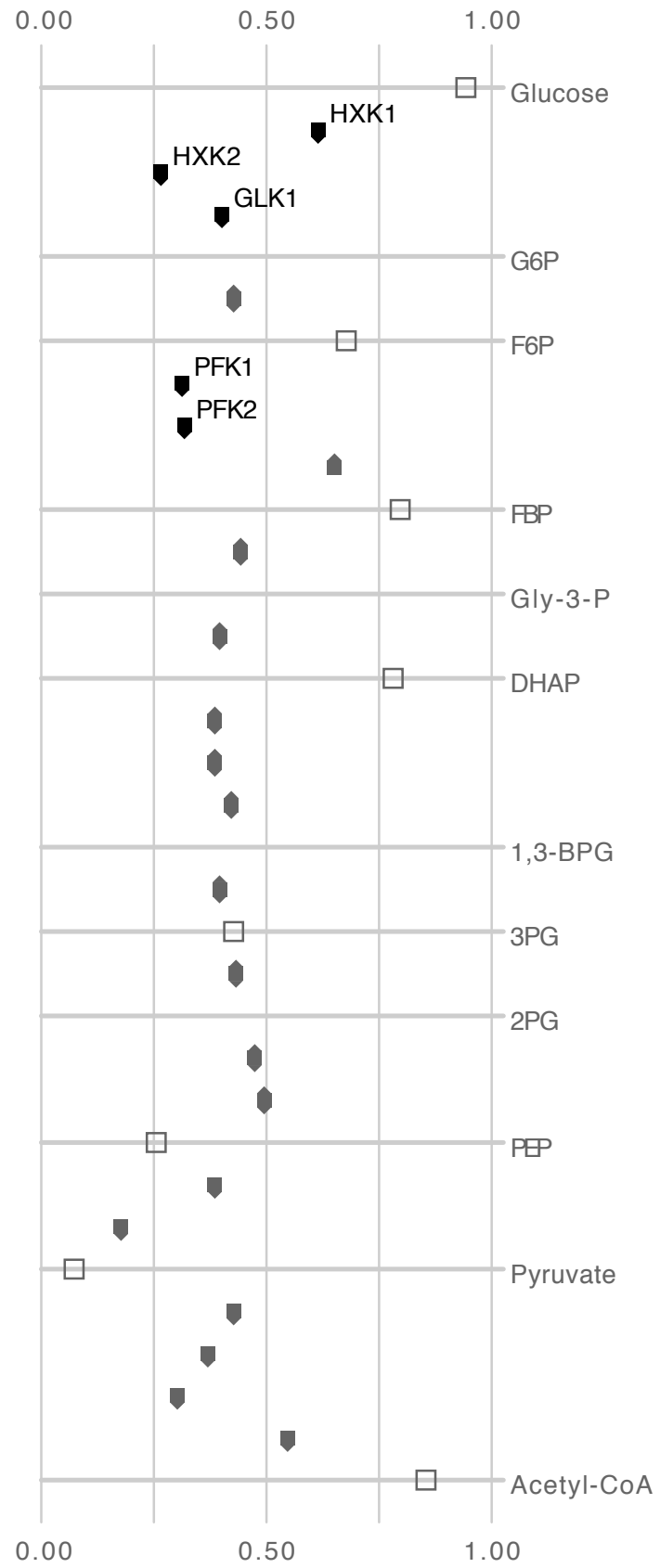
GLYCOLYSIS



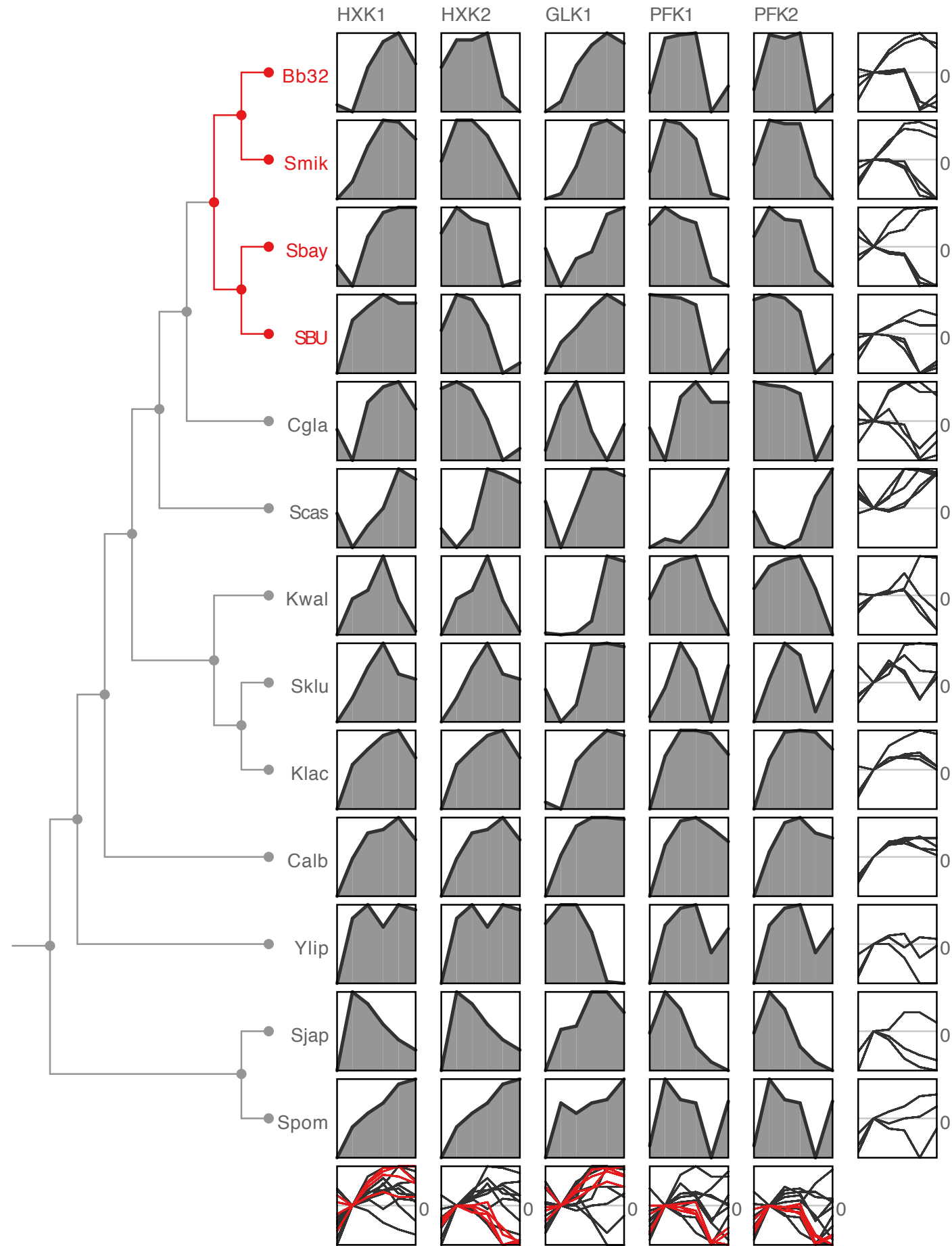
- ▾ forward enzyme
- ▴ reverse enzyme
- ◆ bidirectional enzyme
- metabolite



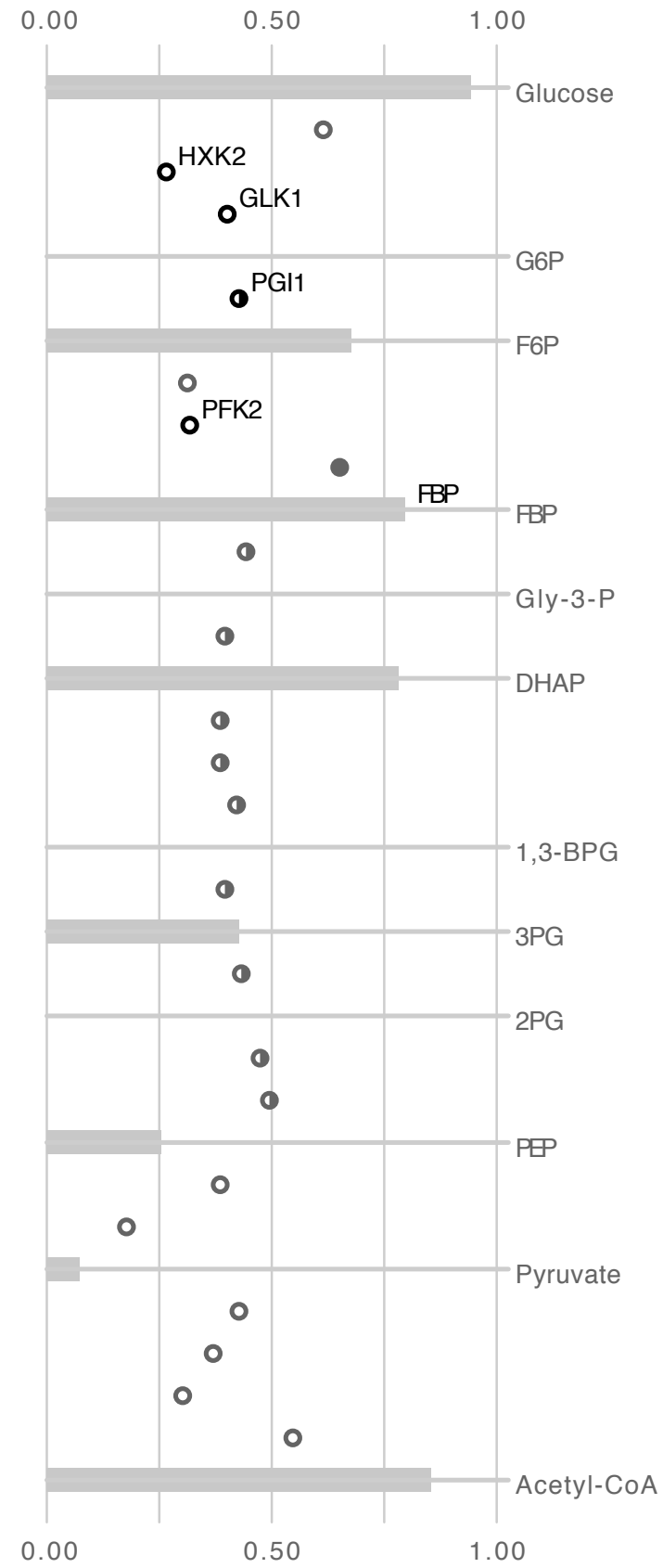
GLYCOLYSIS



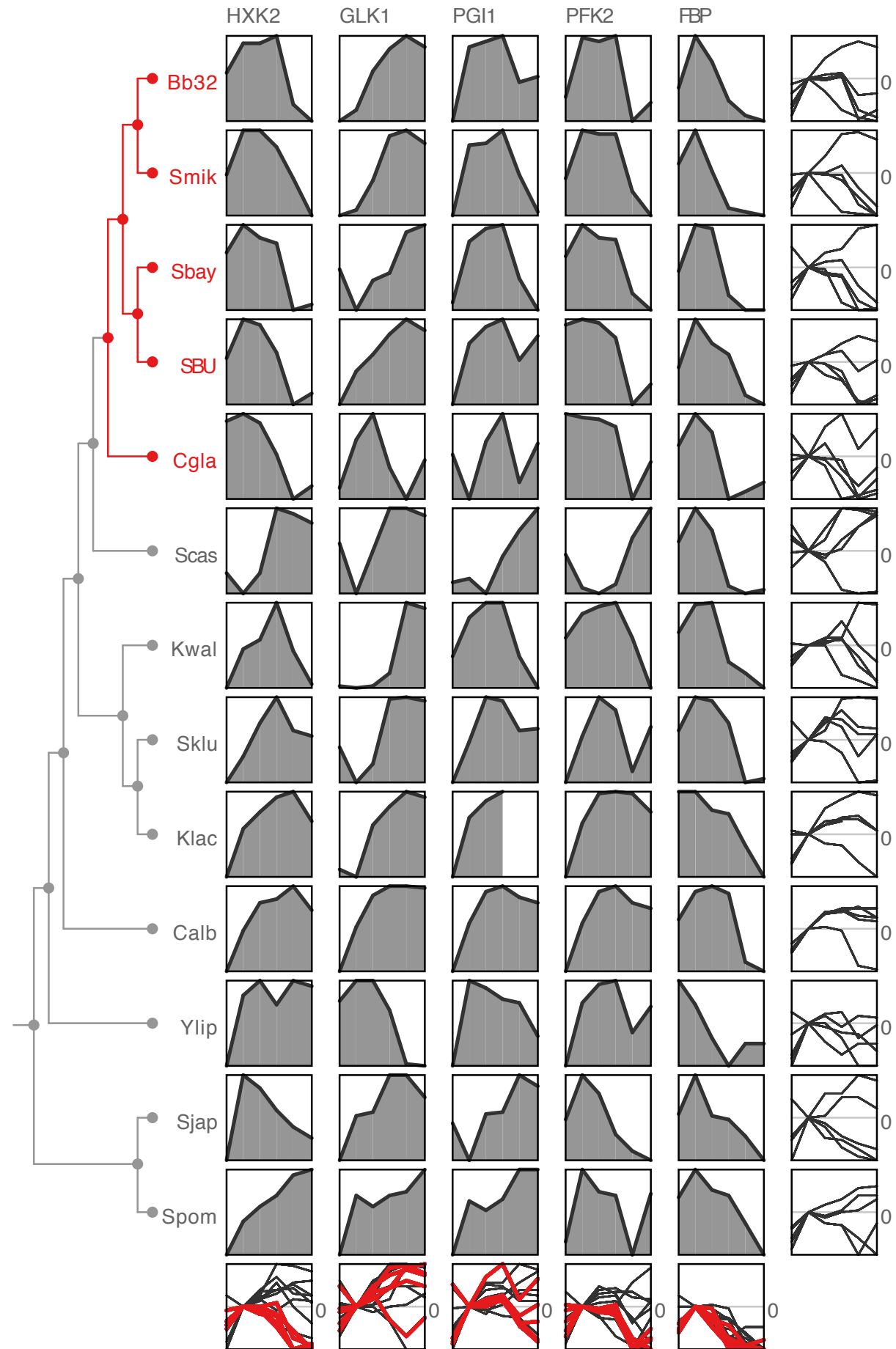
- ▾ forward enzyme
- ▴ reverse enzyme
- ◆ bidirectional enzyme
- metabolite



GLYCOLYSIS



- forward enzyme
- reverse enzyme
- ◐ bidirectional enzyme



LESSONS LEARNED

- **process supports efficient development**
- **collaborators' time commitment is front loaded**
- **rapid prototyping is essential**
- **put off coding as long as possible**

contributions

- **Pathline**

- multiple genes, time points, species, and pathways

- **linearized pathway representation**

- **curvemap**

- **tool deployment**

- open source

- used daily by several collaborators

-visualization design process

-types of research contributions

5.- PAPER TYPES

A Visweek paper typically falls into one of five categories: technique, system, design study, evaluation, or model. We briefly discuss these categories below. Although your main paper type has to be specified during the paper submission process, papers can include elements of more than one of these categories. Please see "Process and Pitfalls in Writing Information Visualization Research Papers" by Tamara Munzner for more detailed discussion on how to write a successful Visweek paper.

Technique papers introduce novel techniques or algorithms that have not previously appeared in the literature, or that significantly extend known techniques or algorithms, for example by scaling to datasets of much larger size than before or by generalizing a technique to a larger class of uses. The technique or algorithm description provided in the paper should be complete enough that a competent graduate student in visualization could implement the work, and the authors should create a prototype implementation of the methods. Relevant previous work must be referenced, and the advantage of the new methods over it should be clearly demonstrated. There should be a discussion of the tasks and datasets for which this new method is appropriate, and its limitations. Evaluation through informal or formal user studies, or other methods, will often serve to strengthen the paper, but are not mandatory.

System papers present a blend of algorithms, technical requirements, user requirements, and design that solves a major problem. The system that is described is both novel and important, and has been implemented. The rationale for significant design decisions is provided, and the system is compared to documented, best-of-breed systems already in use. The comparison includes specific discussion of how the described system differs from and is, in some significant respects, superior to those systems. For example, the described system may offer substantial advancements in the performance or usability of visualization systems, or novel capabilities. Every effort should be made to eliminate external factors (such as advances in processor performance, memory sizes or operating system features) that would affect this comparison. For further suggestions, please review "How (and How Not) to Write a Good Systems Paper" by Roy Levin and David Redell, and "Empirical Methods in CS and AI" by Toby

L4: Data

REQUIRED READING

Data Principles

Many aspects of a visualization design are driven by the kind of data that we have at our disposal: what kind of data do we need to look at? What information can we figure out from the data itself, versus the meanings that we must be told explicitly? What high-level concepts will allow us to split datasets apart into general and useful pieces? What kind of attributes does the data have to begin with, and what kinds of derived data might we compute in order to draw a more effective picture?

This chapter approaches these questions with a taxonomy of visualization data types and semantics that meshes well with the principles in Part II and methods in Part III. Figure 2.1 shows the big picture, which will be expanded on in the rest of the chapter.

The chapter begins with a discussion of dataset types, then makes a distinction between semantics and types, and continues with attribute types and attribute semantics. It returns to datasets with semantics. The chapter then covers derived attributes and spaces. It concludes by relating this taxonomy of data principles to the idea of the data abstraction level of the nested design model.

2.1 Dataset Types

Polaris: A System for Query, Analysis, and Visualization of Multidimensional Relational Databases

Chris Stolte, Diane Tang, and Pat Hanrahan

Abstract—In the last several years, large multidimensional databases have become common in a variety of applications such as data warehousing and scientific computing. Analysis and exploration tasks place significant demands on the interfaces to these databases. Because of the size of the data sets, dense graphical representations are more effective for exploration than spreadsheets and charts. Furthermore, because of the exploratory nature of the analysis, it must be possible for the analysts to change visualizations rapidly as they pursue a cycle involving first hypothesis and then experimentation. In this paper, we present Polaris, an interface for exploring large multidimensional databases that extends the well-known Pivot Table interface. The novel features of Polaris include an interface for constructing visual specifications of table-based graphical displays and the ability to generate a precise set of relational queries from the visual specifications. The visual specifications can be rapidly and incrementally developed, giving the analyst visual feedback as they construct complex queries and visualizations.

Index Terms—Database visualization, database analysis, visualization formalism, multidimensional databases.

1 INTRODUCTION

IN the last several years, large databases have become common in a variety of applications. Corporations are creating large data warehouses of historical data on key aspects of their operations. International research projects such as the Human Genome Project [20] and Digital Sky Survey [31] are generating massive databases of scientific data.

generated from the resulting tables. Visual Insights recently released a new interface for visually exploring projections of data cubes using linked views of bar charts, scatterplots, and parallel coordinate displays [14].

In this paper, we present Polaris, an interface for the exploration of multidimensional databases that extends the Pivot Table interface to directly generate a rich, expressive