



Scalable, Dynamic Analysis and Visualization for Genomic Datasets

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Laboratory for BIOINFORMATICS
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Science has become **data-centric**

- Enormous amounts of data being observed
- Data then analyzed by experts in the field or computer scientists/statisticians
- Most disciplines – critical to have human insight during data analysis => **visualization**



Data growing in genomics: systems-level studies

- Cells & organisms are complex systems
- Many biological processes & diseases result from multiple changes on molecular level
- To understand and cure cancer & other diseases, need to observe and model cellular processes on a **systems level**

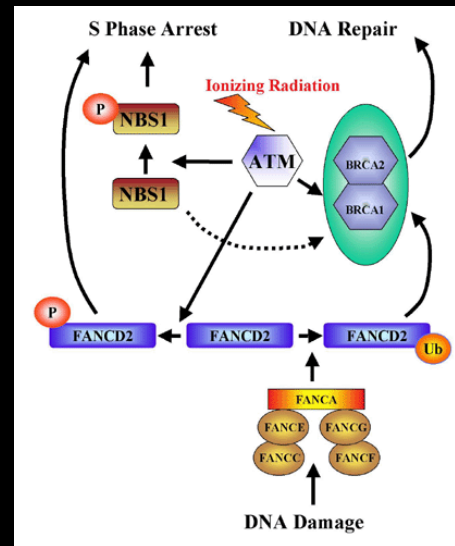
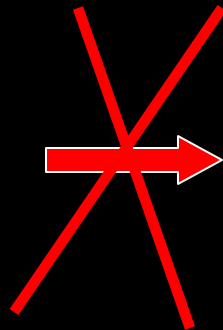




Data-Knowledge gap in biology



Explosion of
genomic **DATA**



KNOWLEDGE of components
and inter-relationships that lead to
function



Outline

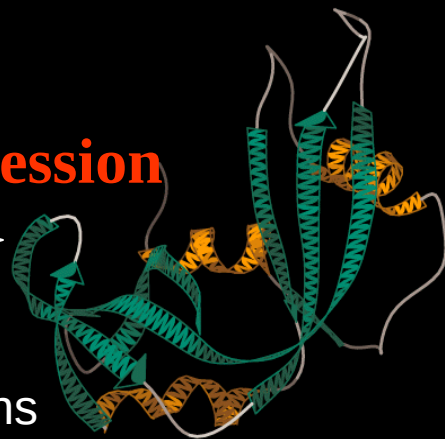
- Background
- Visualization and Analysis Solutions
 - ForestView
 - SPELL
 - GOLEM
- Integration of Solutions

Isn't genomics "done" with the human genome?

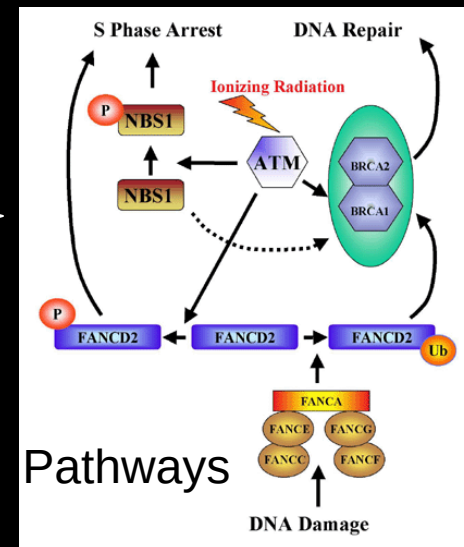


DNA

Gene Expression



Proteins



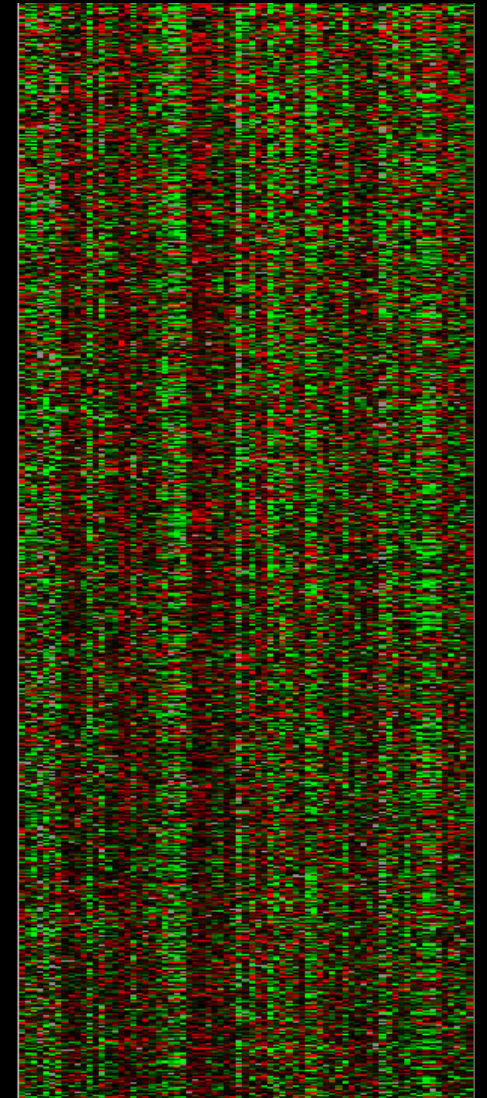
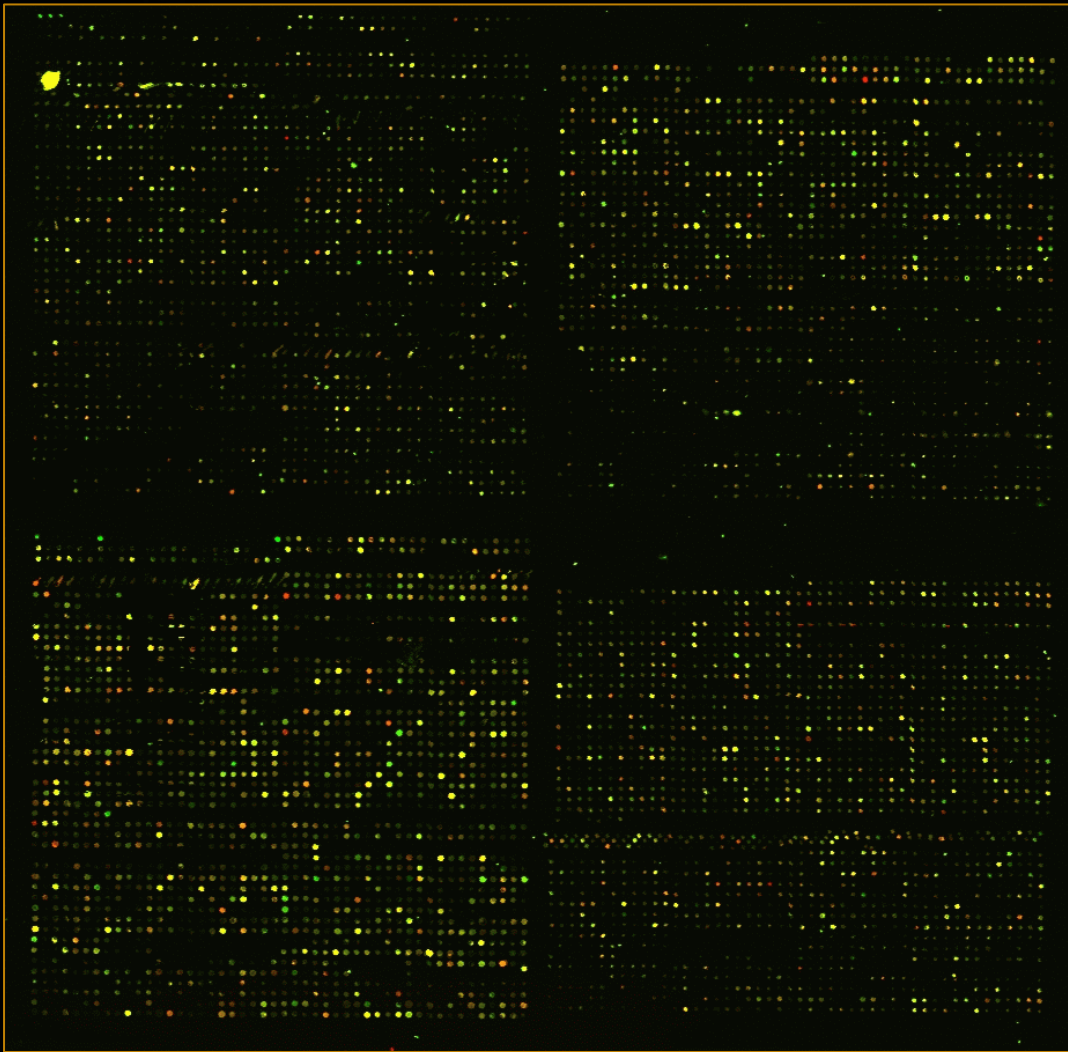
Pathways



Phenotypes



Microarray Results



Raw Image from Spellman et al., 98

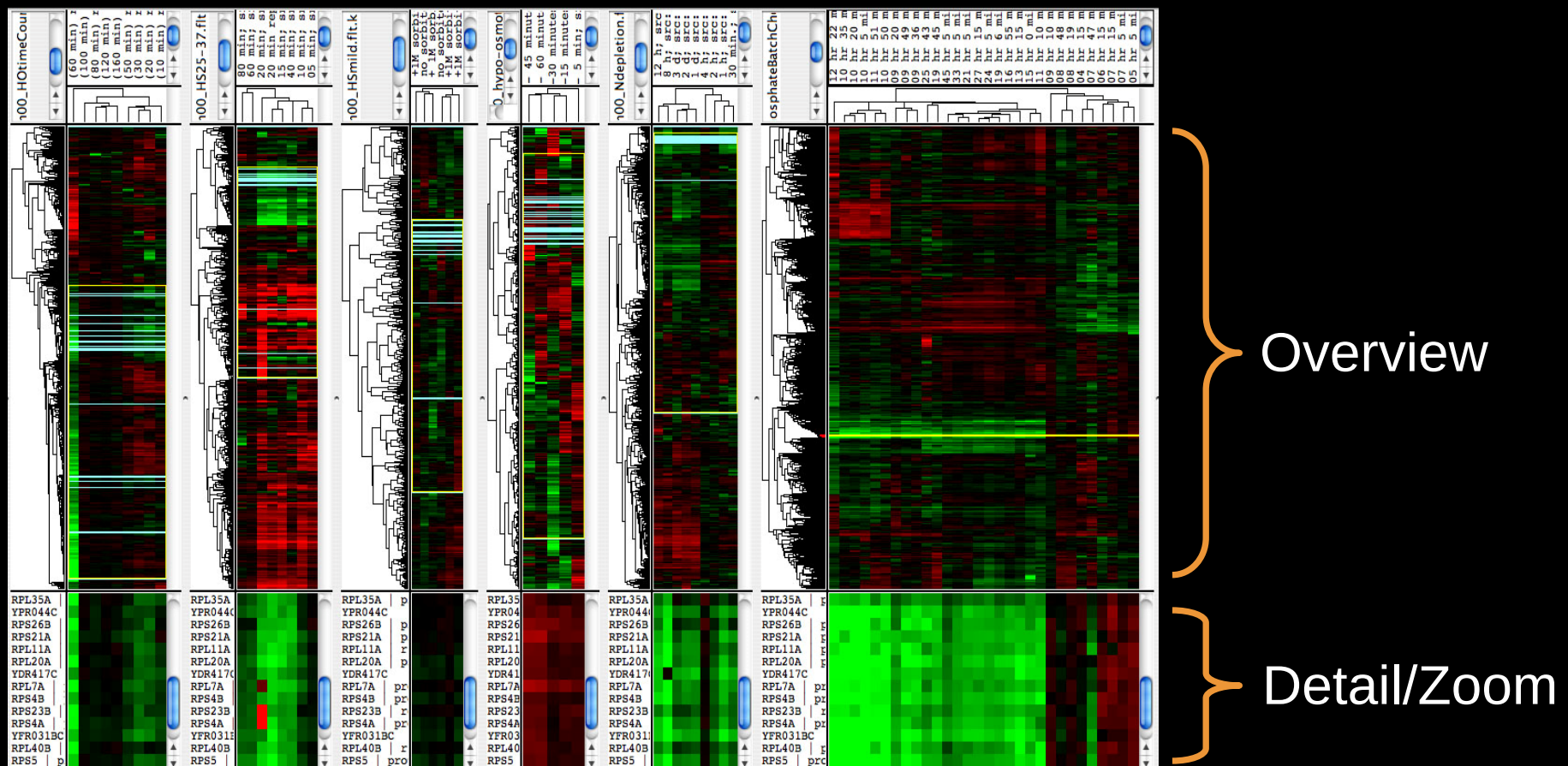


Challenges of Microarray Analysis

- LOTS of data
 - Hundreds of millions of measurements exploring cancer
- Strong need for collaboration
- Biologist's insight critical analysis => visualization-based analysis is critical
- Limited by effective visualization capacities
 - Screen space at a premium
 - Need to identify relevant data subsets
- Most analysis small-scale due to these issues
- Future of discovery will require larger-scale, integrated analysis and visualization

ForestView

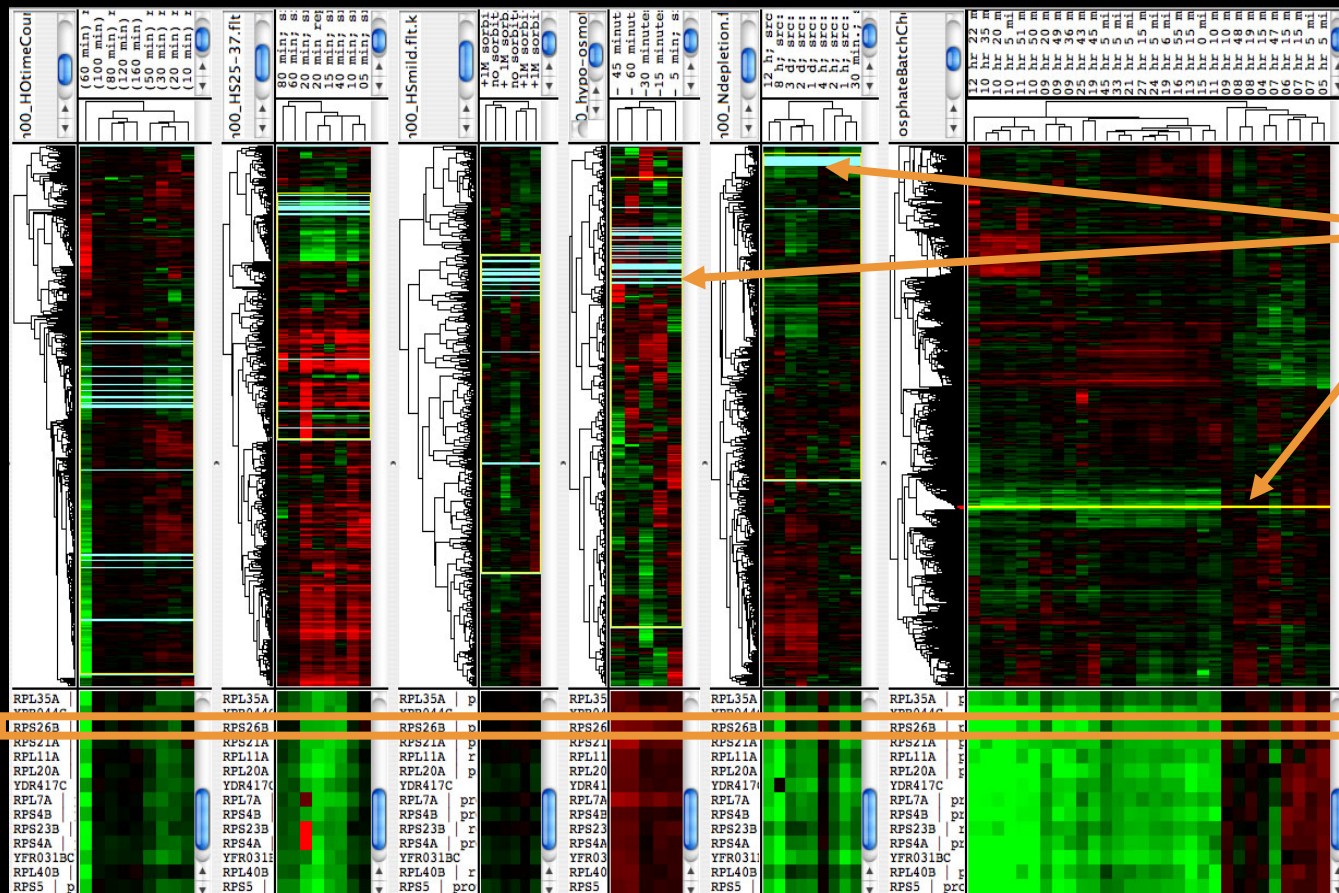
- Overview+Detail+Setting through Horizontal Integration
 - Places data in the larger context of other available data



Six diverse datasets linked together

ForestView

- Overview+Detail+Setting through Horizontal Integration
 - Places data in the larger context of other available data



Selections
linked across
datasets

Overview

Details aligned
horizontally

Detail/Zoom

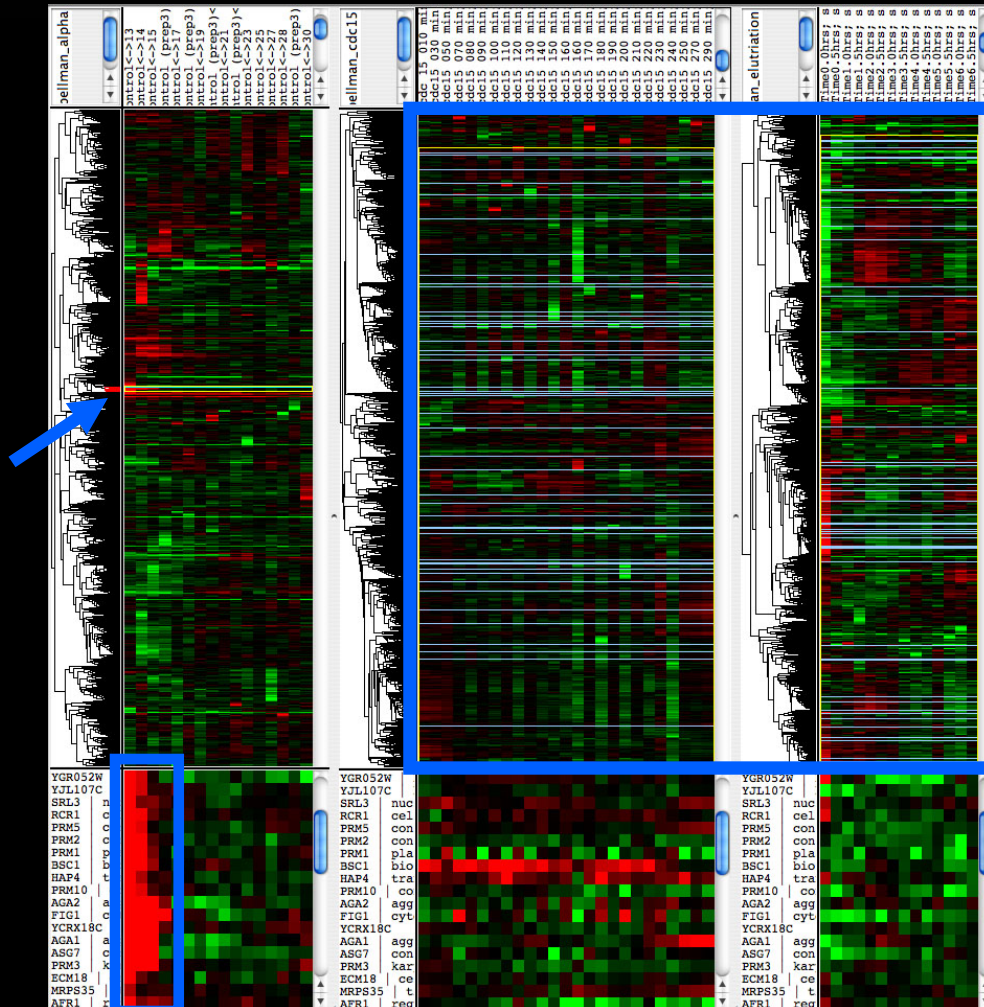
Six diverse datasets linked together

ForestView Observation

alpha-factor

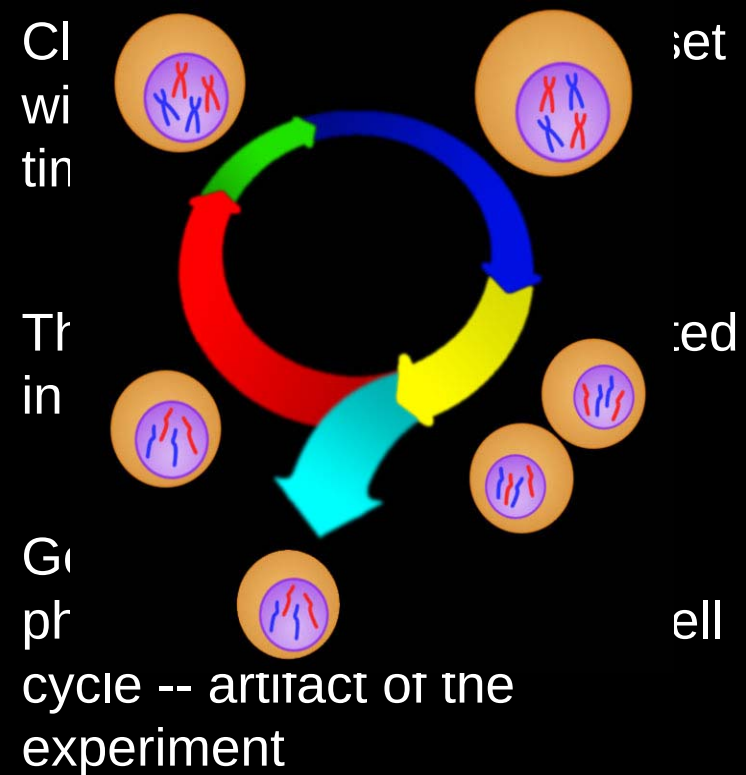
cdc15

elutriation

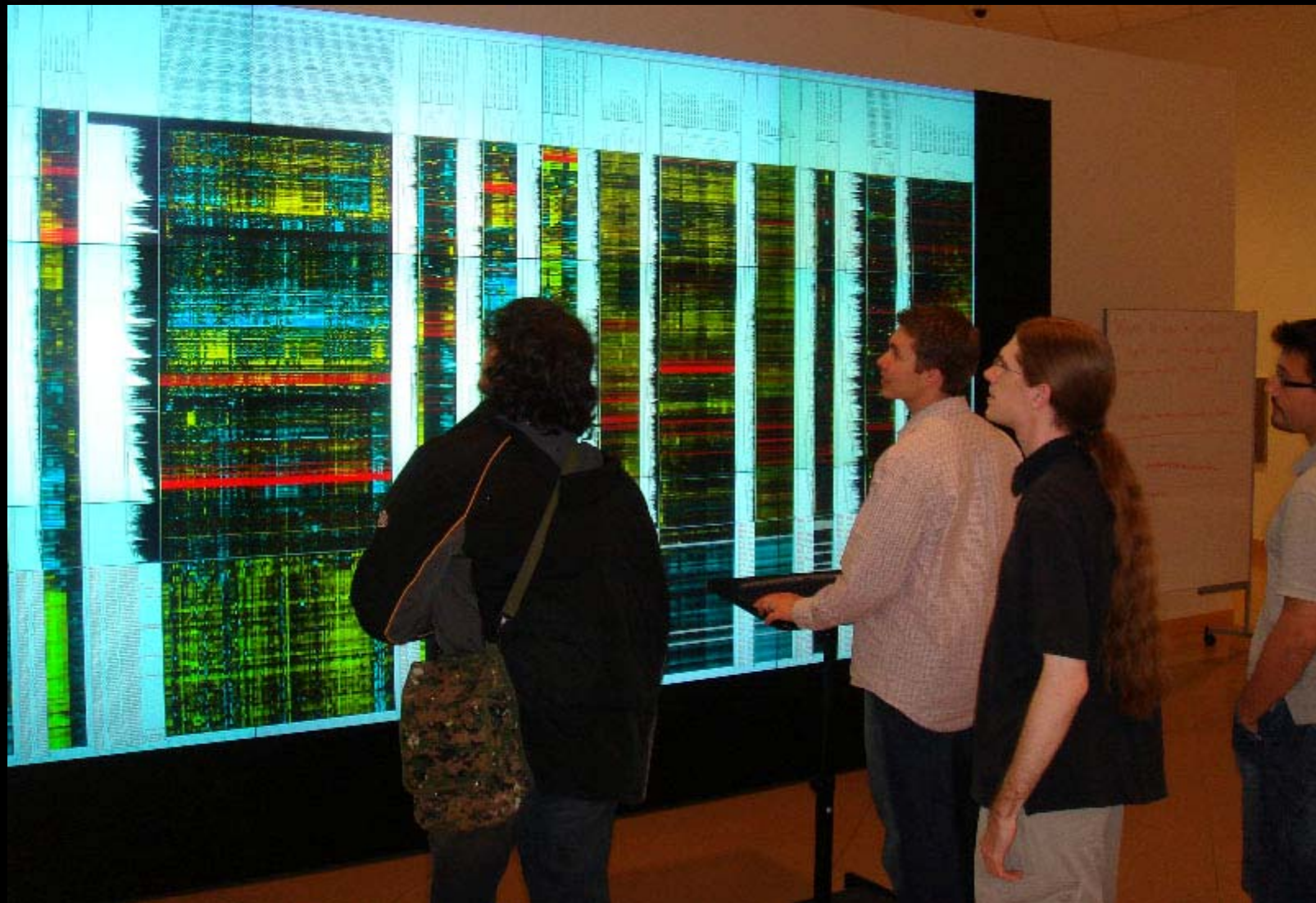


Spellman cell cycle datasets

Three datasets studying cell cycle phases -- cancer can be related to cell cycle defects



ForestView Scalability



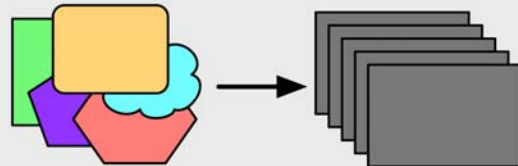


Analysis Across Multiple Datasets

- ForestView increases amount of data simultaneously viewed...
- But still much more data available
- Need to identify relevant datasets, and identify novel genes related to functions
- Our solution:
- SPELL (Serial Patterns of Expression Levels Locator) – a similarity search and visualization engine

System Overview

Data Collection & Normalization



User Specifies Query Context

SPELL (Yeast)

Gene Name(s):

Main SPELL Algorithm

Calculate correlations
in a standardized,
noise-reduced space

Weight dataset relevance
based on co-expression of the
query genes

Identify additional genes
co-expressed with the query
set given the dataset
relevance weights.

Display Search Results to User

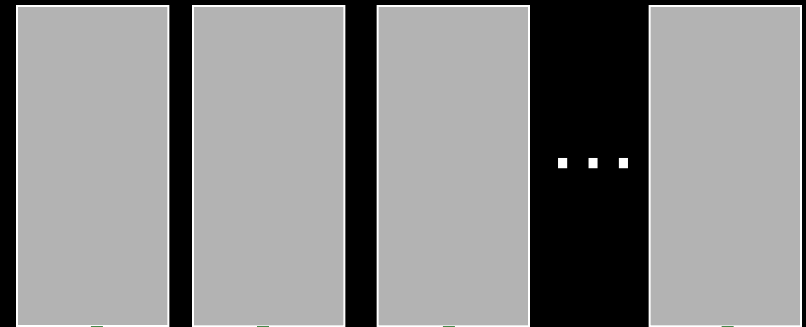
Dataset	Causton HG et al., 2001	Shenoy M et al., 2004	Rodriguez- Cohen BA et al., 2004	Sabeti N et al., 2004	Fleming JA et al., 2002	O'Rourke SM et al., 2004	Roberts CJ et al., 2000	Taniguchi Y et al., 2000	Shenoy T et al., 2001
Description	Alkaline response	Oxidative stress response	nutrient starvation	pup1 and pup2 mutants with deacetylase inhibition and histone deletion mutants	Proteasome inhibition with PS-341	HOG MAPK pathway	Phenol response	Galactose response	Galactose response
Contribution	3.4%	3.0%	2.8%	2.8%	2.7%	2.6%	2.4%	2.3%	2.3%
ACS#	Gene								
2.7	EUP1								
2.7	EUP2								
2.6	SGL1								
2.6	RPT1								
2.6	PHB1								
2.6	PHB2								
2.6	PHB3								
2.6	PHB4								
2.6	PHB5								
2.6	PHB6								
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2.6	PHB8								
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2.6	PHB99								
2.6	PHB100								



Algorithm Details

1. Transform data through signal re-weighting

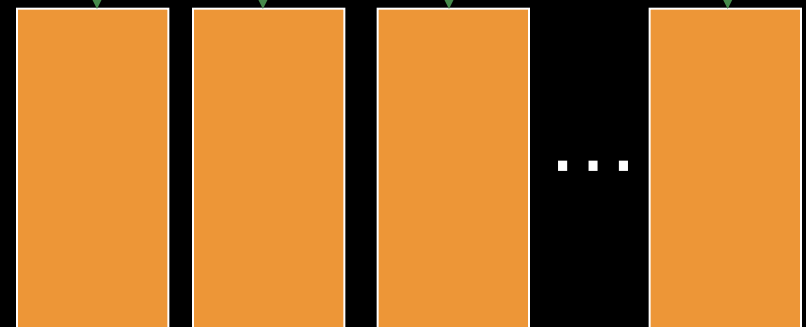
Original
Data



SVD

$$X_{m \times n} = U_{m \times n} \Sigma_{n \times n} V_{n \times n}^t$$

Signal Re-
weighted
Data





Algorithm Details

1. Transform data through signal re-weighting
2. Given a query, determine dataset weights

User specifies N query genes

SPELL (Yeast)

Gene Name(s):

Avg z-score between query pairs is dataset weight

$$w_d = \left(\frac{2}{N(N-1)} \right) \sum_{i=1}^{N-1} \sum_{j=i+1}^N (z'_{q_i, q_j})$$





Algorithm Details

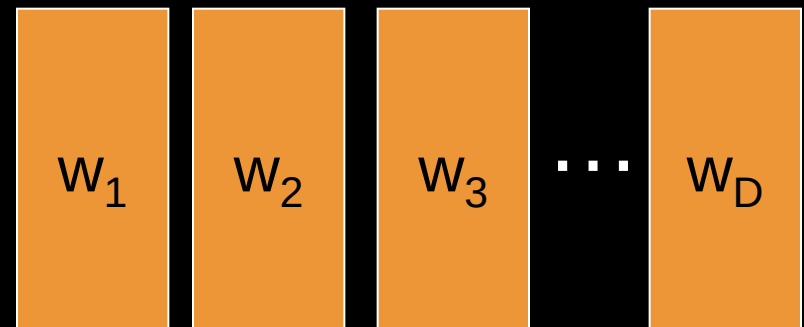
1. Transform data through signal re-weighting
2. Given a query, determine dataset weights
3. Identify additional co-expressed genes

SPELL (Yeast)

Gene Name(s):

Scores for all other genes are weighted average z-scores

$$s_x = \left(\frac{1}{DN} \right) \sum_{d=1}^D \sum_{i=1}^N w_d z'_{x,q_i}$$



Algorithm Details

1. Transform data through signal re-weighting
2. Given a query, determine dataset weights
3. Identify additional co-expressed genes
4. Display results

Dataset	Gasch AP et al., 2000	Caba E et al., 2005	Saldanha AJ et al., 2004	Bro C et al., 2003 show all →
Description	Diamide treatment time course	Genotoxic stress	Limitation by Phosphate	Lithium response
Contribution ²	4.3%	4.1%	3.8%	3.2%
ACS ² Gene				
☒ --	CTR9			
☒ --	MED2			
☐ 1.4	BIR1			
☐ 1.4	LSM3			
☐ 1.4	TAF12			
☐ 1.4	ENT1			
☐ 1.4	TFG1			
☐ 1.4	RSC58			
☐ 1.4	YNG2			
☐ 1.3	NUP2			
☐ 1.3	ENT5			
☐ 1.3	ARP8			
☐ 1.3	DJP1			
☐ 1.3	PRP3			
☐ 1.3	VPS16			
☐ 1.3	FHL1			
☐ 1.3	SRB4			
☐ 1.3	SET2			

The SPELL logo features the word "SPELL" in a stylized, serif font. Behind the letters are three lit candles of varying heights, casting a warm glow. The entire logo is set against a dark, textured background.

Search results² for DEM1, CDC28

Refine Search

Show Expression Levels

show
all →

regulator null
mutant
2.5%

[illegible]

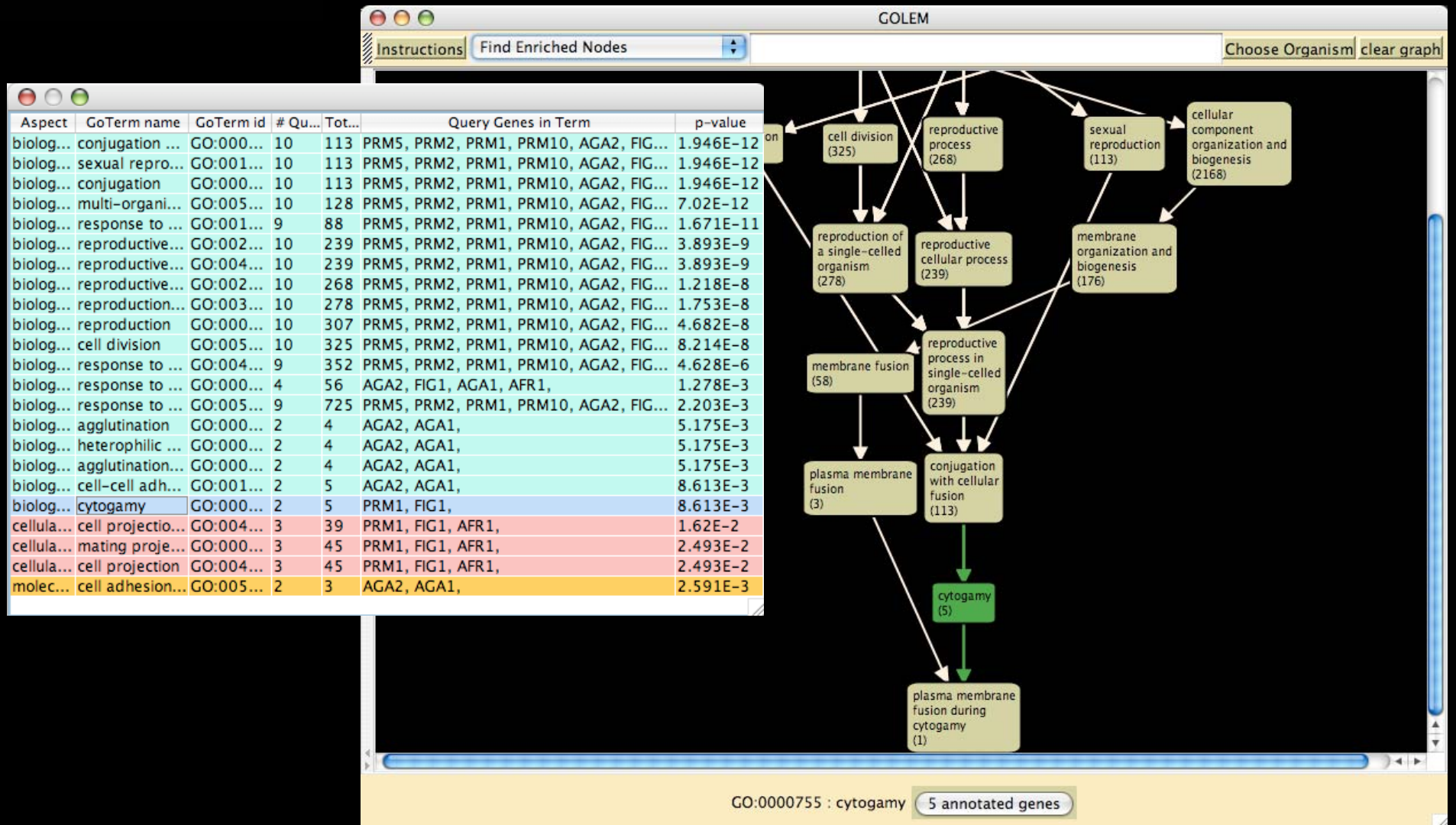


GOLEM: going beyond genomic data

- Also need analysis tools for verification and interpretation of search or visual analysis results
- Gene Ontology (GO) - curated hierarchical structure that represents known biology
- GOLEM - tool for finding GO enrichment, viewing ontology structure

GOLEM

- Using genes from the cell cycle example:





The future: dynamic, integrated search, analysis and visualization

SPELL query used to identify datasets relevant to a gene set

SPELL (Yeast)

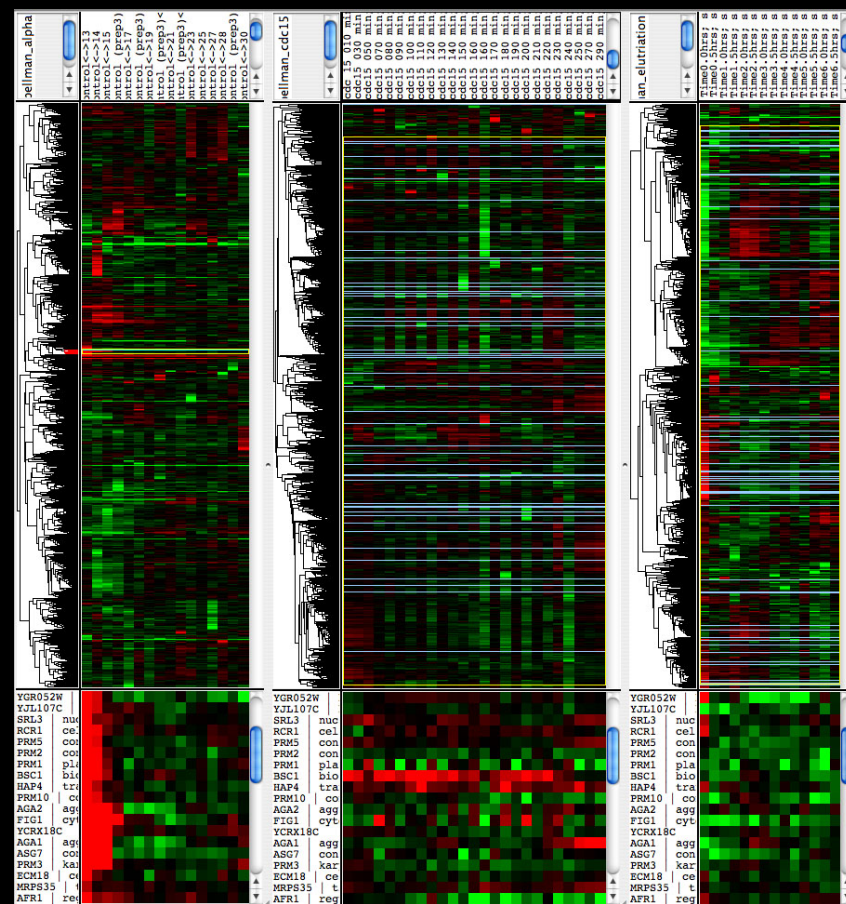
Gene Name(s):

Datasets explored in ForestView, related genes selected

Results used to further explore current data, or refine SPELL search

Aspect	GoTerm name	GoTerm id	# Qu...	Tot...	Query Genes in Term	p-value
biolog...	conjugation ...	GO:000...	10	113	PRM5, PRM2, PRM1, PRM10, AGA2, FIG...	1.946E-12
biolog...	sexual repro...	GO:001...	10	113	PRM5, PRM2, PRM1, PRM10, AGA2, FIG...	1.946E-12
biolog...	conjugation	GO:000...	10	113	PRM5, PRM2, PRM1, PRM10, AGA2, FIG...	1.946E-12
biolog...	multi-organi...	GO:005...	10	128	PRM5, PRM2, PRM1, PRM10, AGA2, FIG...	7.02E-12
biolog...	response to ...	GO:001...	9	88	PRM5, PRM2, PRM1, PRM10, AGA2, FIG...	1.671E-11
biolog...	reproductive...	GO:002...	10	239	PRM5, PRM2, PRM1, PRM10, AGA2, FIG...	3.893E-9
biolog...	reproductive...	GO:004...	10	239	PRM5, PRM2, PRM1, PRM10, AGA2, FIG...	3.893E-9
biolog...	reproductive...	GO:002...	10	268	PRM5, PRM2, PRM1, PRM10, AGA2, FIG...	1.218E-8
biolog...	reproduction...	GO:003...	10	278	PRM5, PRM2, PRM1, PRM10, AGA2, FIG...	1.753E-8
biolog...	reproduction...	GO:000...	10	307	PRM5, PRM2, PRM1, PRM10, AGA2, FIG...	4.682E-8
biolog...	cell division	GO:005...	10	325	PRM5, PRM2, PRM1, PRM10, AGA2, FIG...	8.214E-8
biolog...	response to ...	GO:004...	9	352	PRM5, PRM2, PRM1, PRM10, AGA2, FIG...	4.628E-6
biolog...	response to ...	GO:000...	4	56	AGA2, FIG1, AGA1, AFR1,	1.278E-3
biolog...	response to ...	GO:005...	9	725	PRM5, PRM2, PRM1, PRM10, AGA2, FIG...	2.203E-3
biolog...	agglutination	GO:000...	2	4	AGA2, AGA1,	5.175E-3
biolog...	heterophilic ...	GO:000...	2	4	AGA2, AGA1,	5.175E-3
biolog...	agglutination...	GO:000...	2	4	AGA2, AGA1,	5.175E-3
biolog...	cell-cell adh...	GO:001...	2	5	AGA2, AGA1,	8.613E-3
biolog...	cytogamy	GO:000...	2	5	PRM1, FIG1,	8.613E-3
cellula...	cell projectio...	GO:004...	3	39	PRM1, FIG1, AFR1,	1.62E-2
cellula...	mating proje...	GO:000...	3	45	PRM1, FIG1, AFR1,	2.493E-2
cellula...	cell projection...	GO:004...	3	45	PRM1, FIG1, AFR1,	2.493E-2
molec...	cell adhesion...	GO:005...	2	3	AGA2, AGA1,	2.591E-3

GOLEM used to identify functional enrichments of clusters





Conclusions

- Scalable, dynamic visualization-based analysis enables novel biological discoveries
- Input from biology researchers critical in analysis
- Integration of multiple data sets is important in modeling and analyzing biological data
- Integration of visualization and analysis/search is important in genomics



Acknowledgements

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<http://function.princeton.edu>



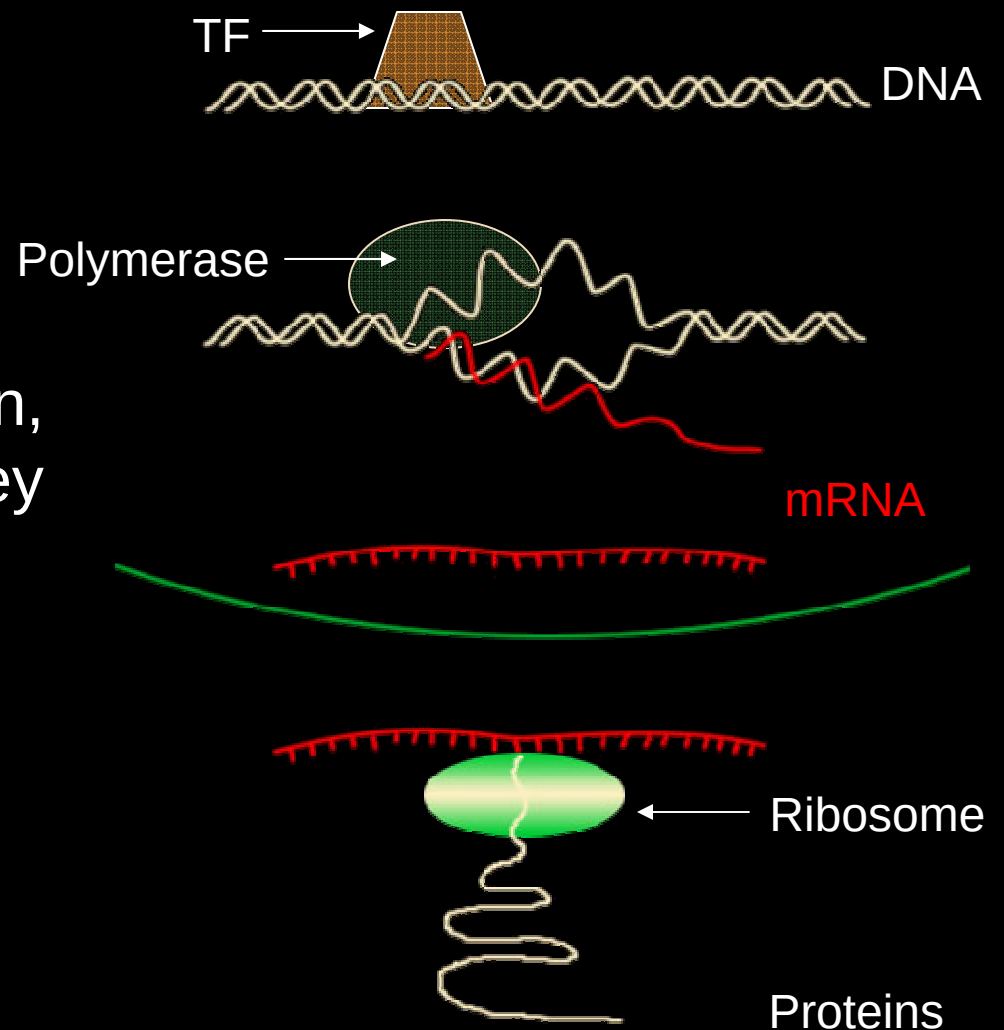
Collaborator:

Kai Li



Central Dogma

- Common cellular mechanisms to create proteins
- Understanding the function, coordination of proteins key to understand disease
- Gene expression microarrays give us a picture of cellular states





Basic Microarray Methodology

