

Integrative analysis of interaction networks

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JTB2010 - Nov.23.2009



Donnelly Centre
for **Cellular + Biomolecular Research**



UNIVERSITY OF
TORONTO



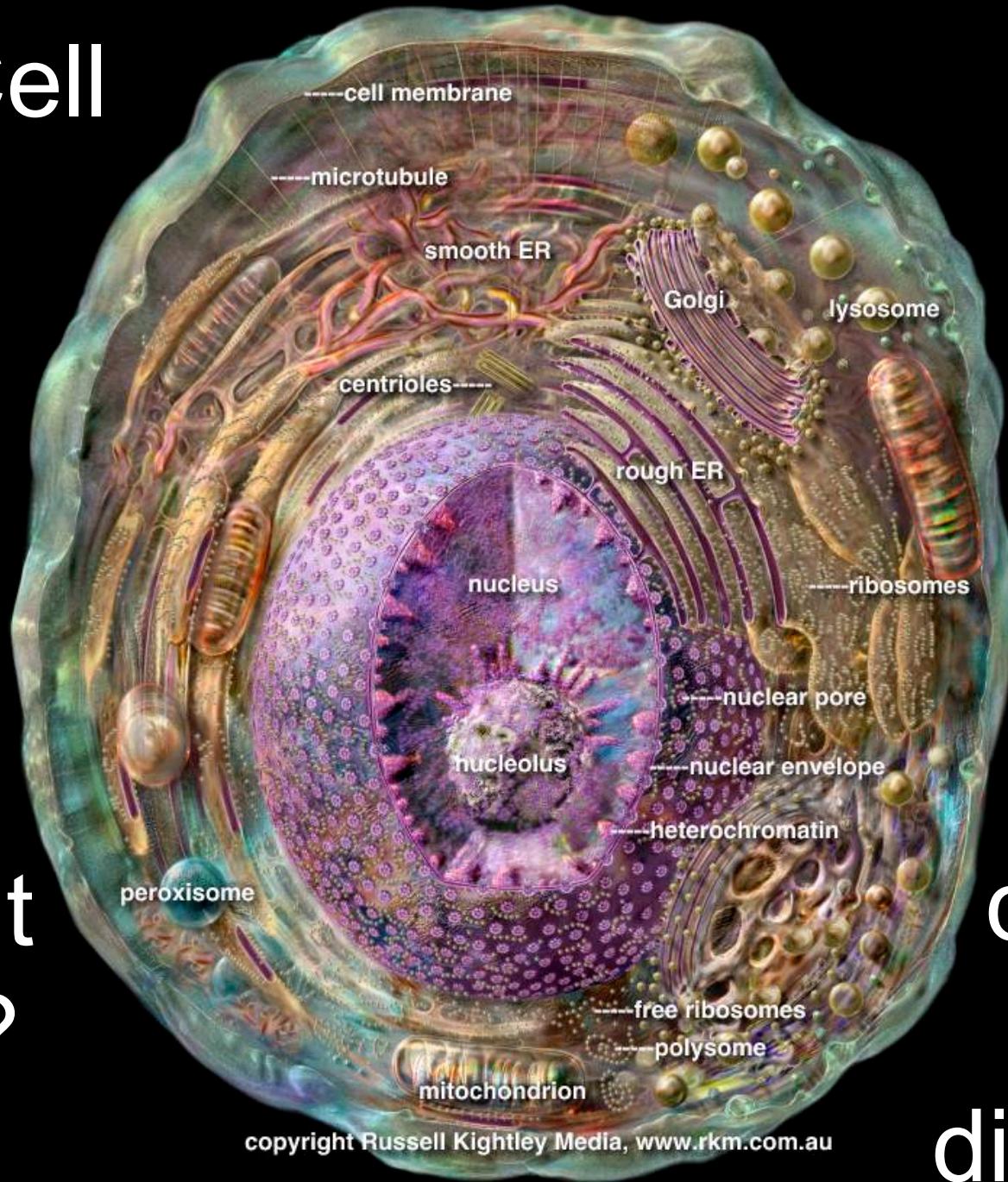
Outline

- Data integration using networks
- Network analysis
- Network data
- Network visualization and analysis
- Analyzing molecular profiles

Data integration using networks

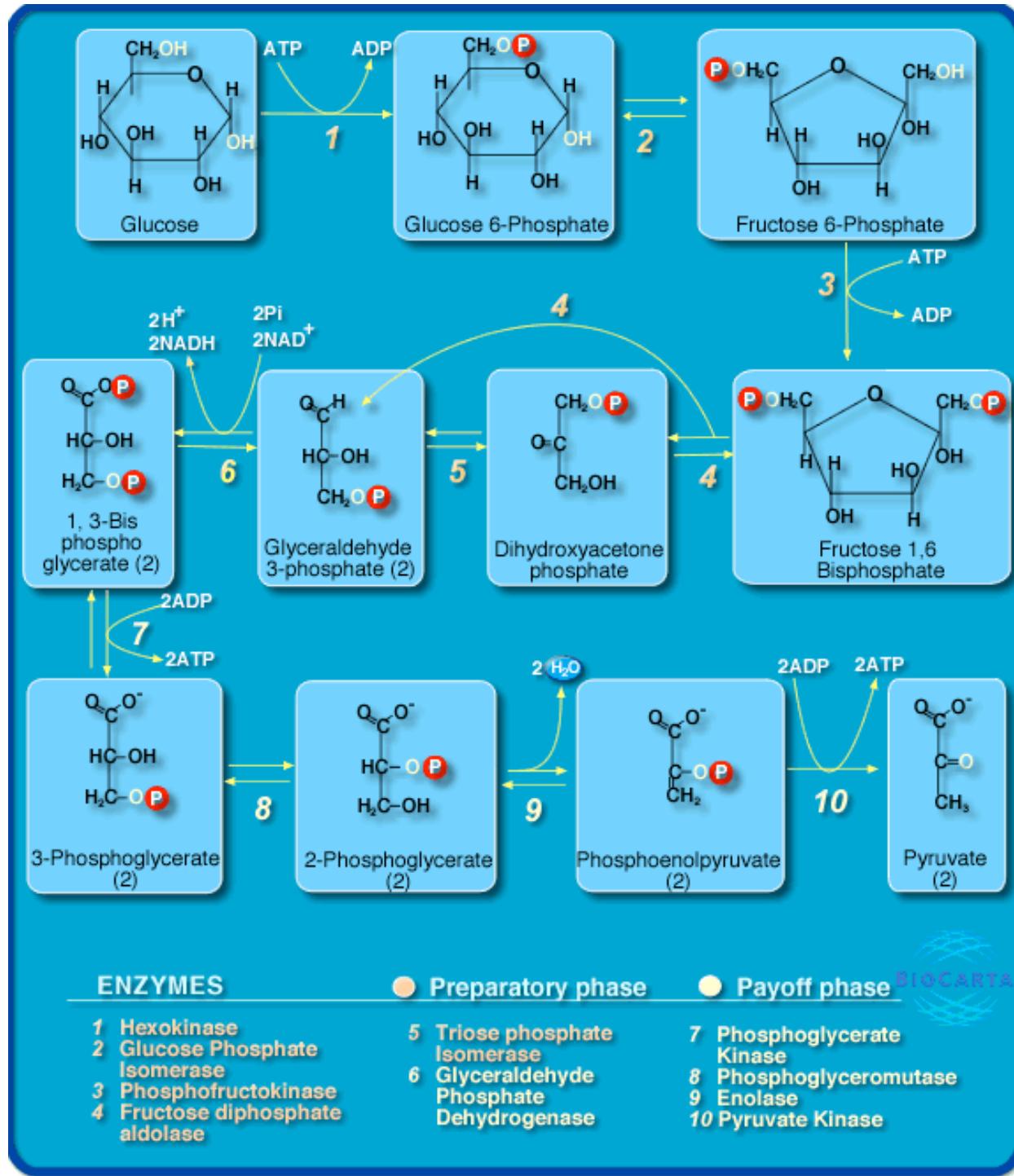
The Cell

How
does it
work?



copyright Russell Kightley Media, www.rkm.com.au

How
does it
fail in
disease?



Signaling Pathway

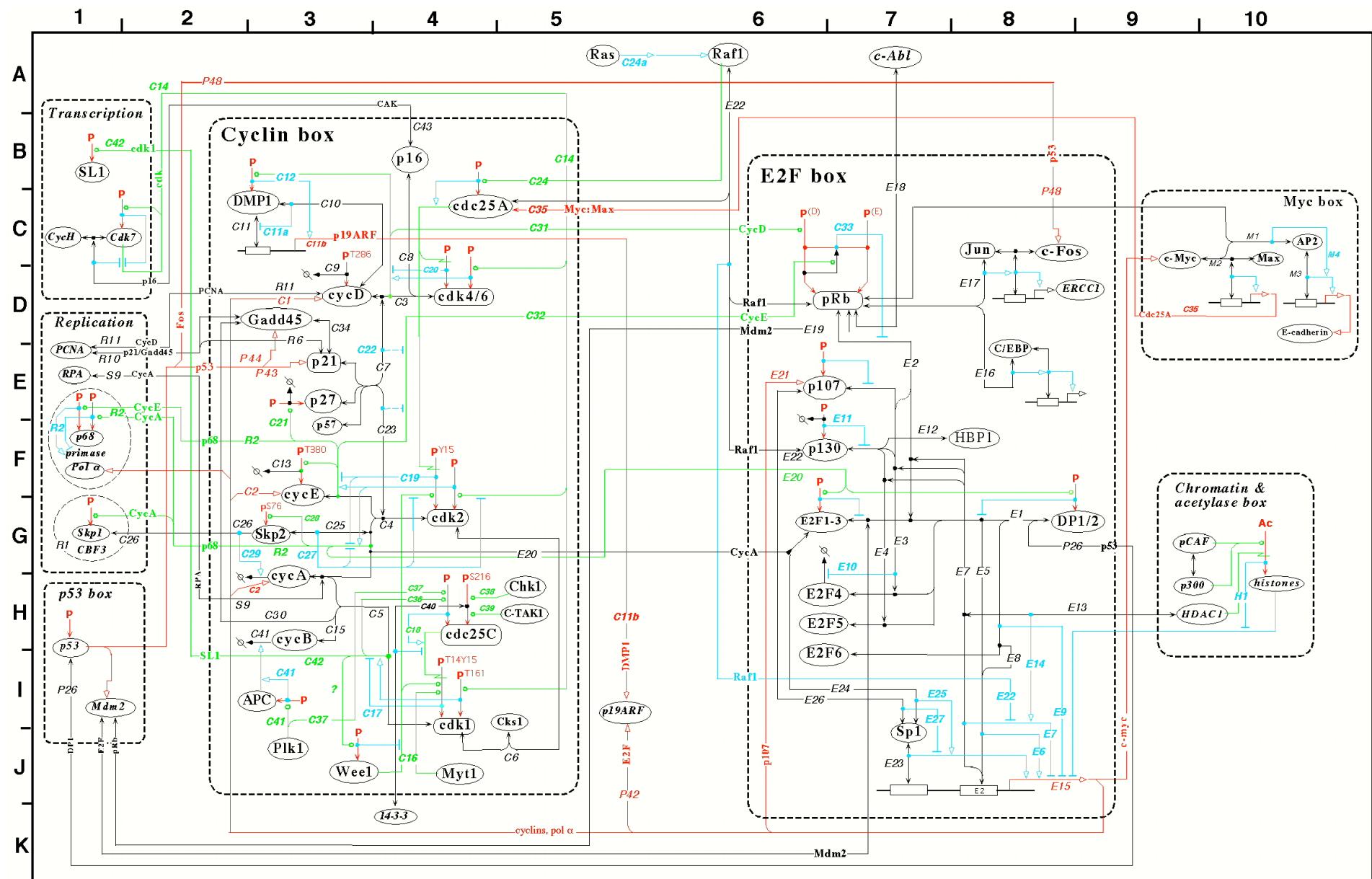
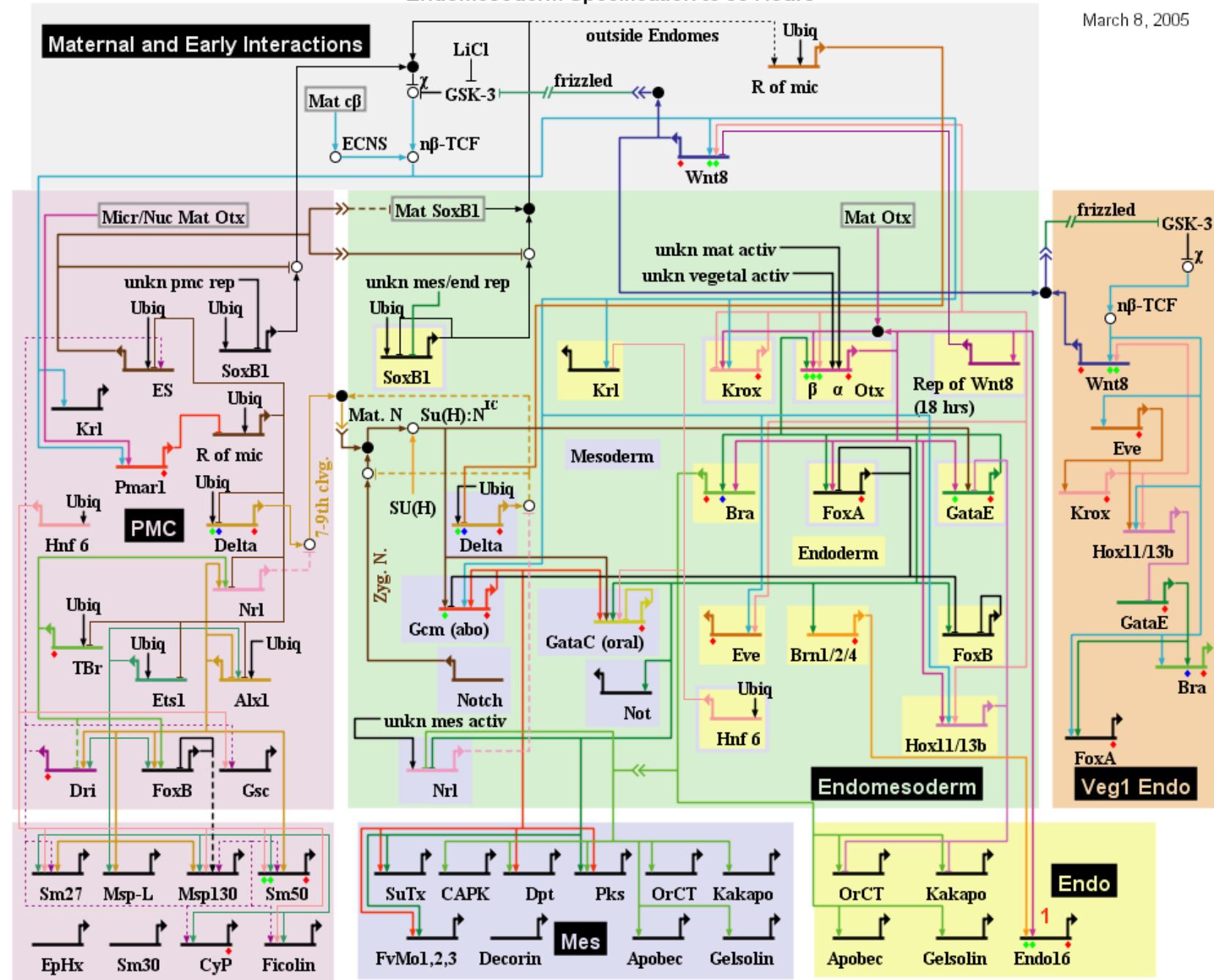


Figure 6A: The Cyclin - E2F cell cycle control system (version 3a - June 8, 1999)

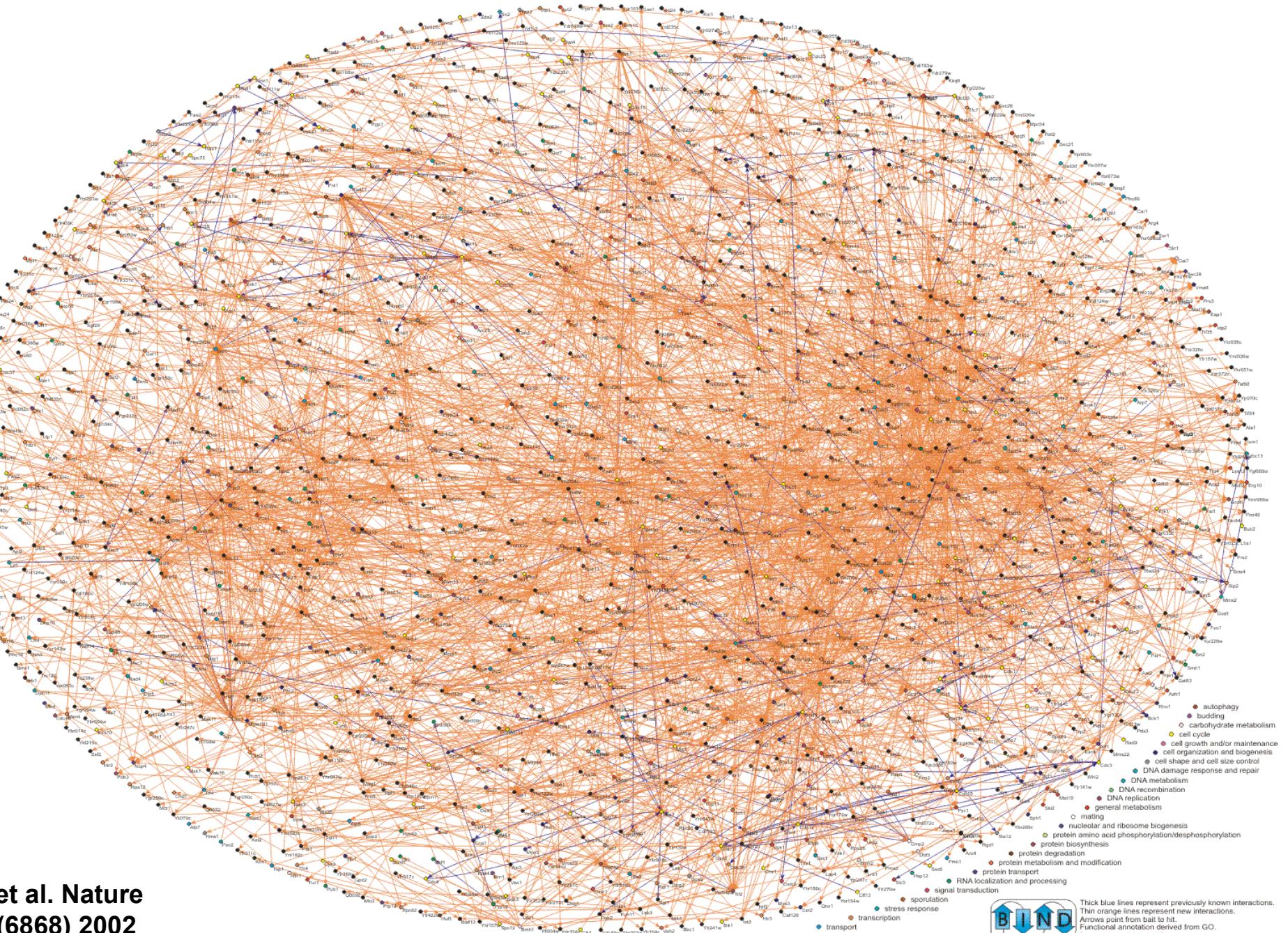
http://discover.nci.nih.gov/kohnk/interaction_maps.html

Endomesoderm Specification to 30 Hours

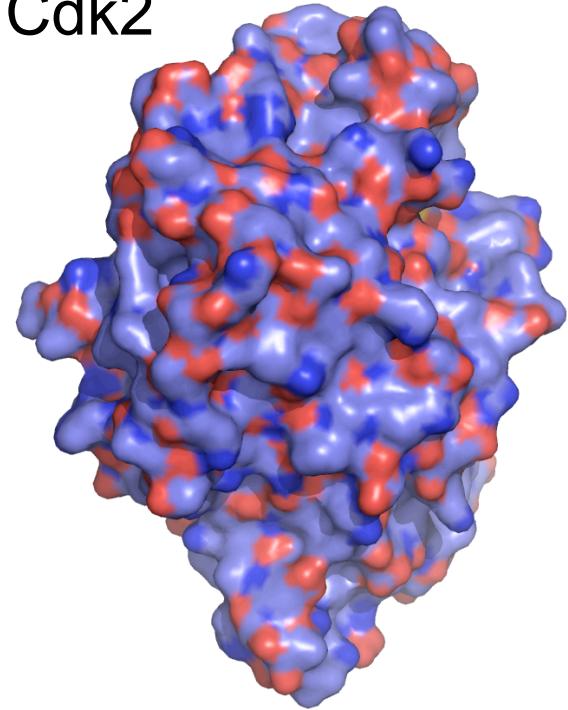
March 8, 2005



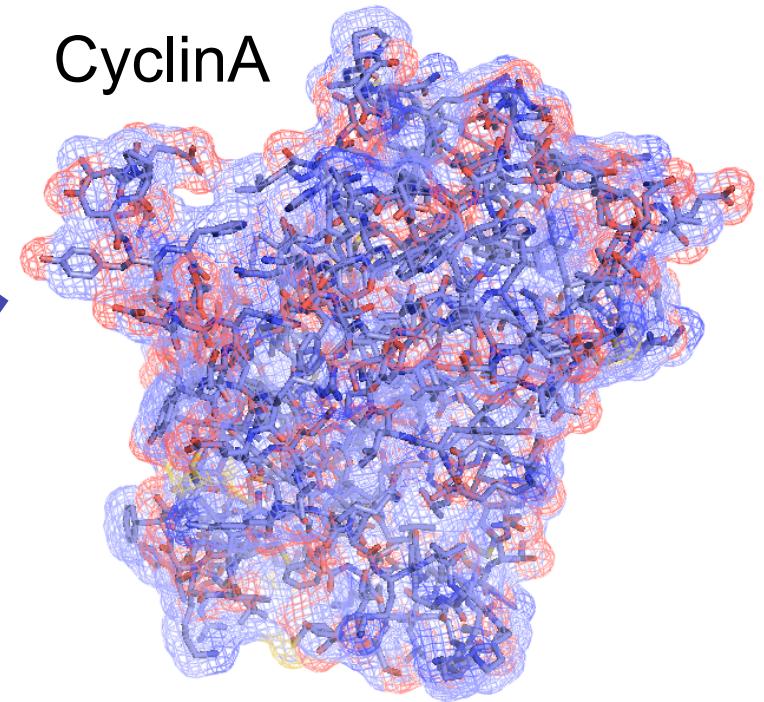
Systematic identification of protein complexes in *Saccharomyces cerevisiae* by mass spectrometry



Cdk2

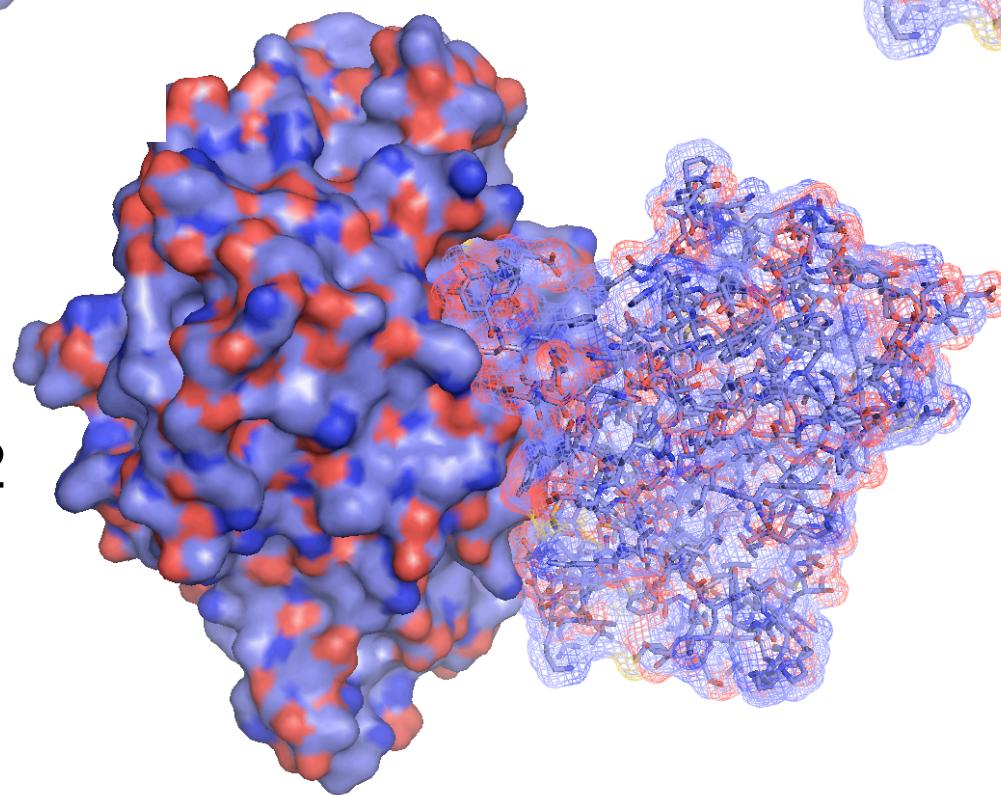


CyclinA



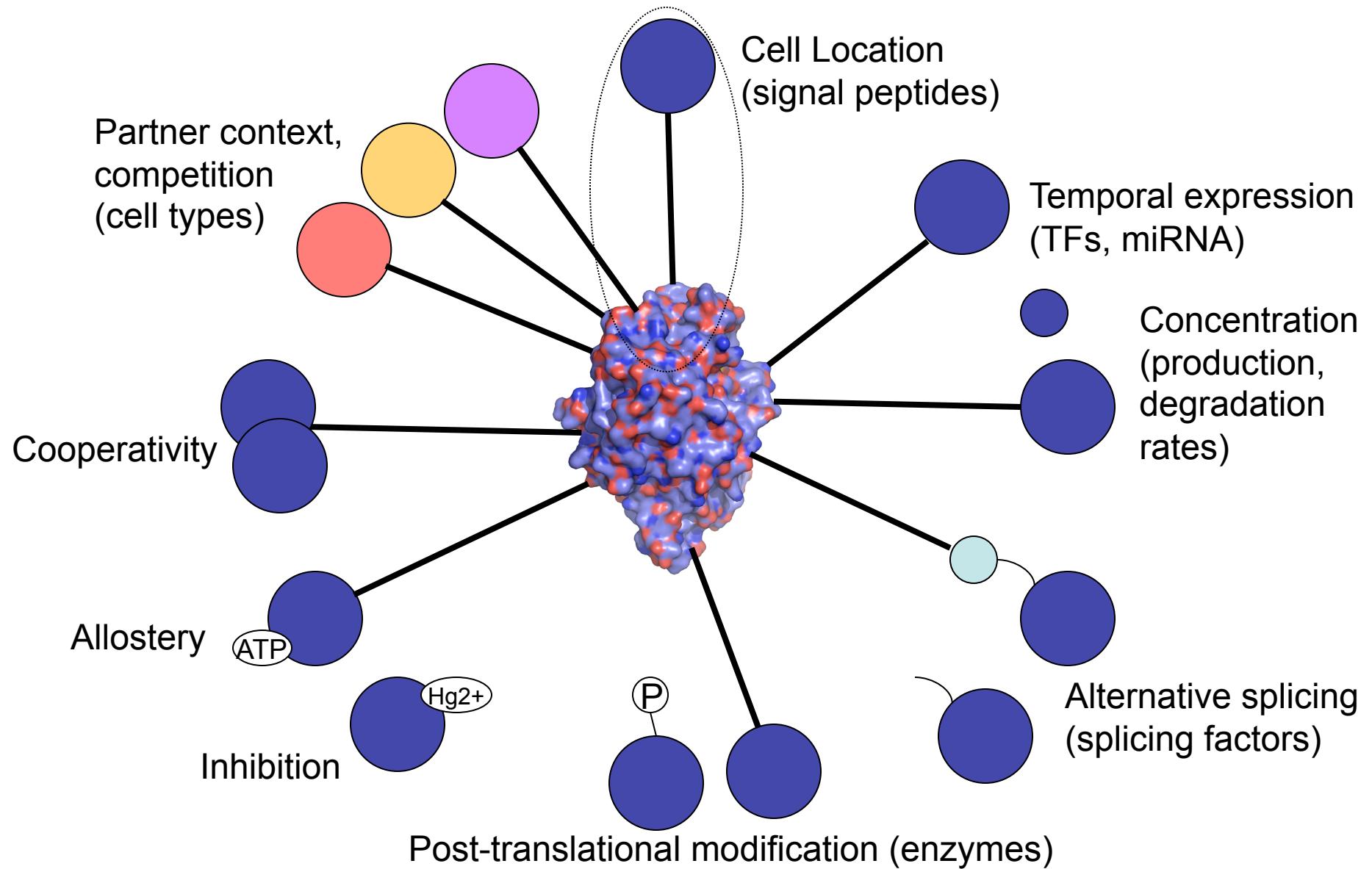
Specific Interaction

CyclinA-Cdk2
complex

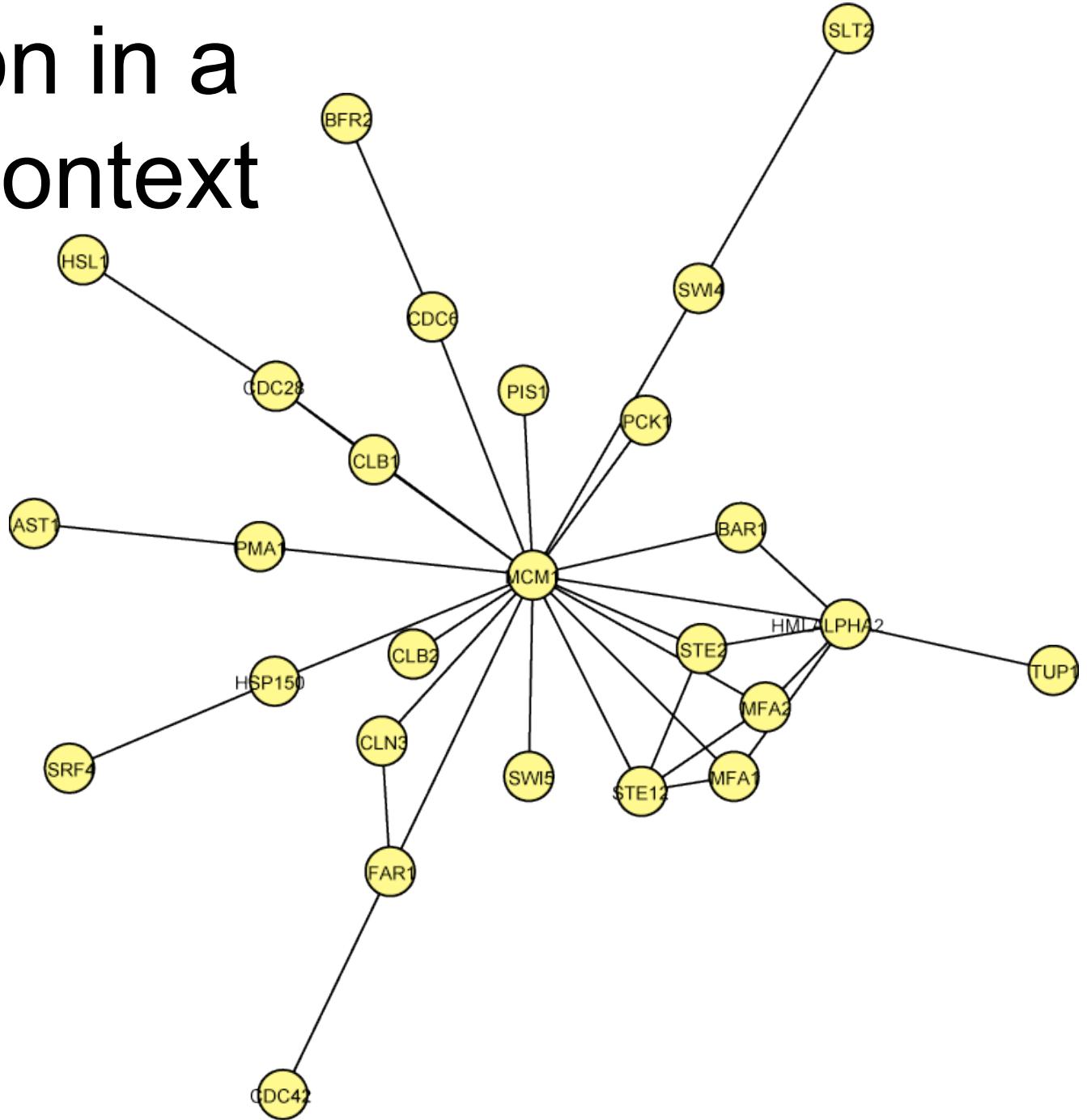


1FIN

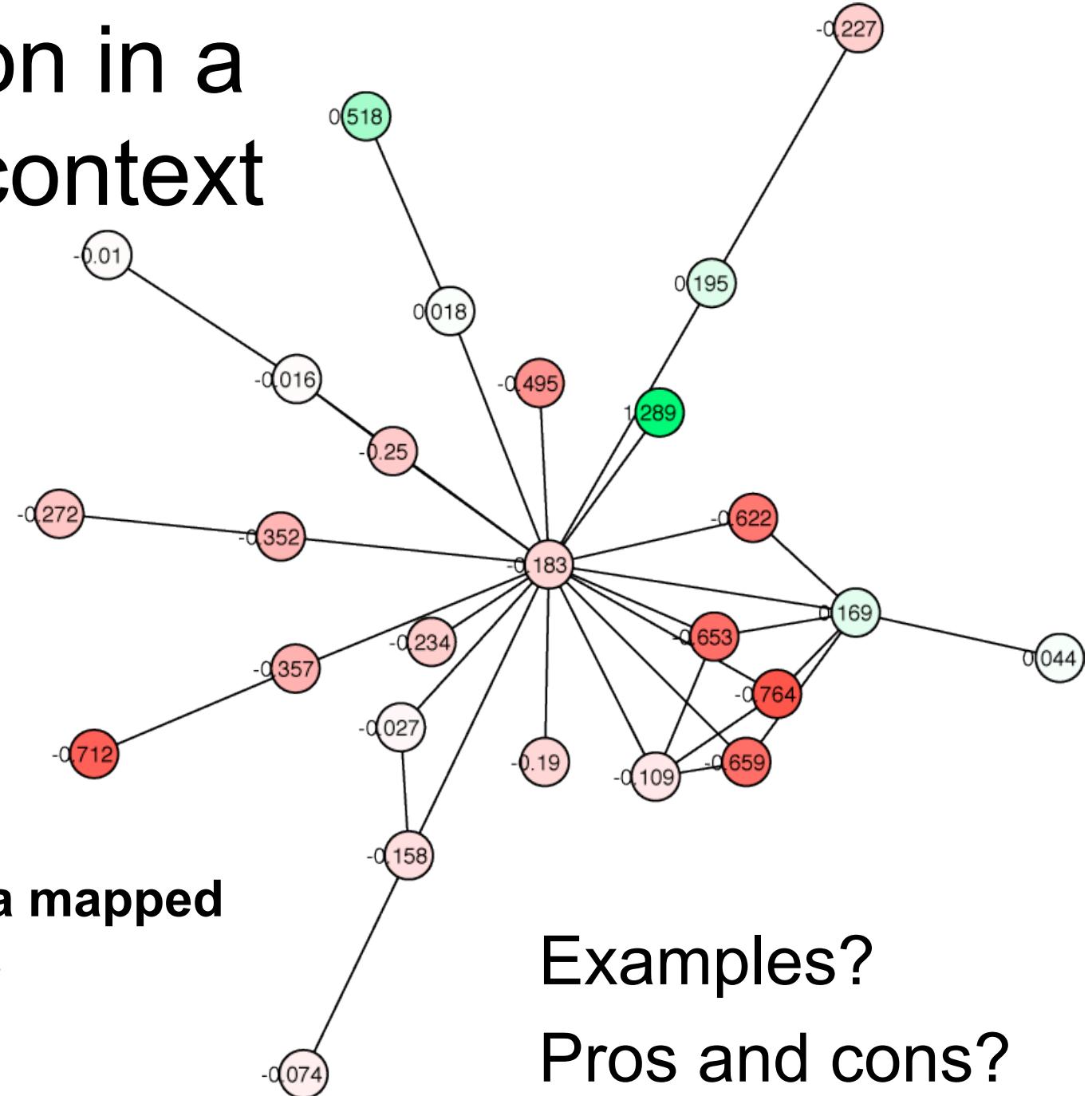
Cellular Context



Integration in a network context



Integration in a network context



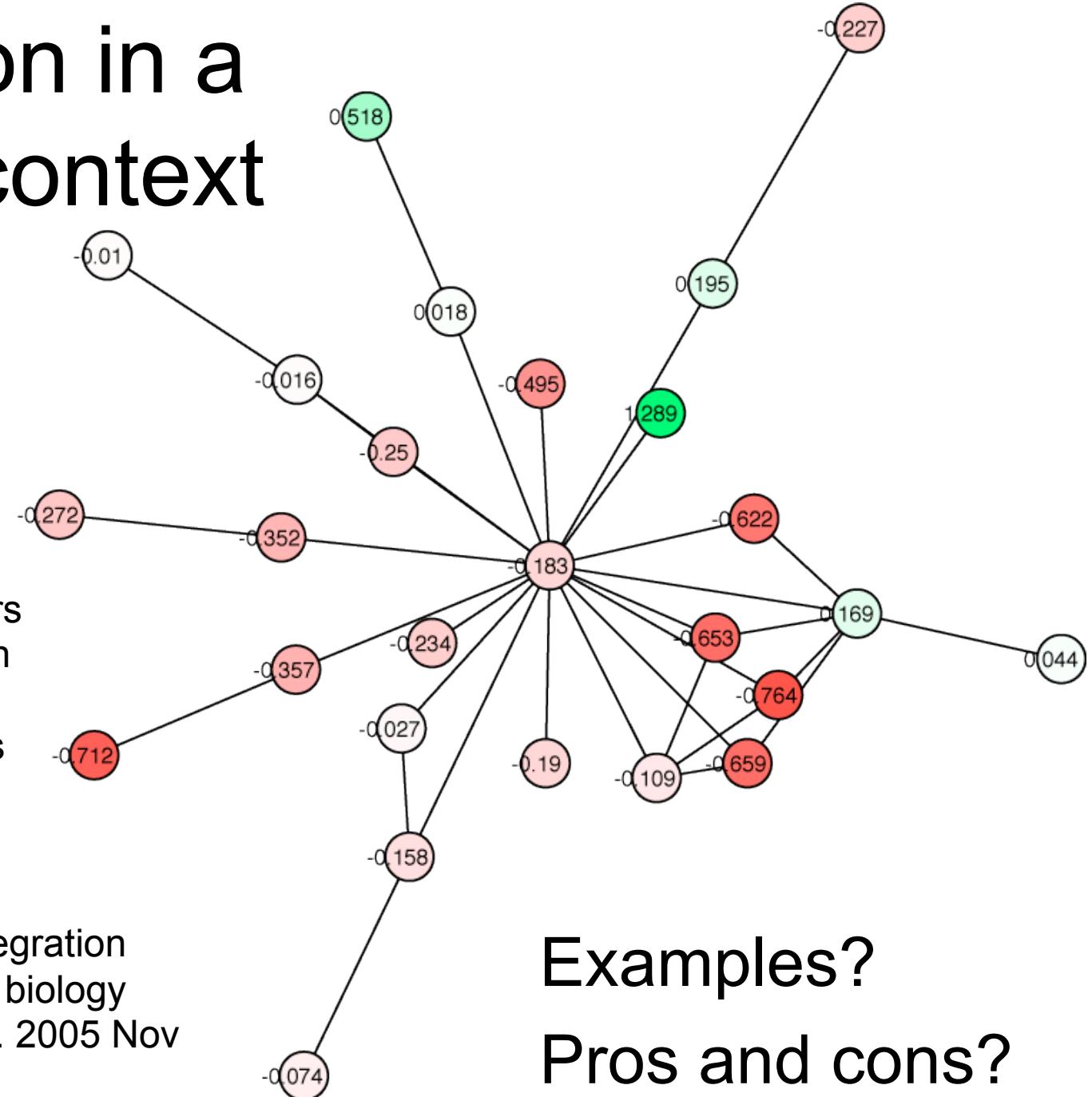
Integration in a network context

Advantages:

- Interpretable
- Broader coverage
- Error reduction

Challenges:

- Must carefully match data sets to avoid errors e.g. different interaction experiments
- Consider data set bias
- Consider binary vs. discrete vs. continuous



Hwang D et al. A data integration methodology for systems biology
Proc Natl Acad Sci U S A. 2005 Nov 29;102(48):17296-301

Examples?
Pros and cons?



Data Integration



Network Analysis

Why Network Analysis?

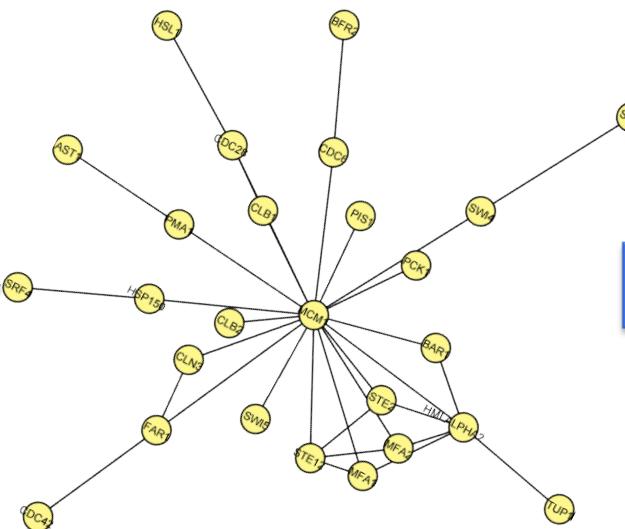
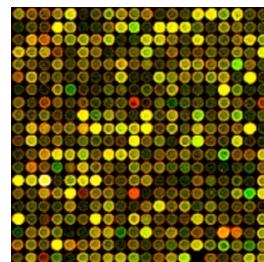
Intuitive to Biologists

- Provide a biological context for results
- More efficient than searching databases gene-by-gene
- Intuitive display for sharing data

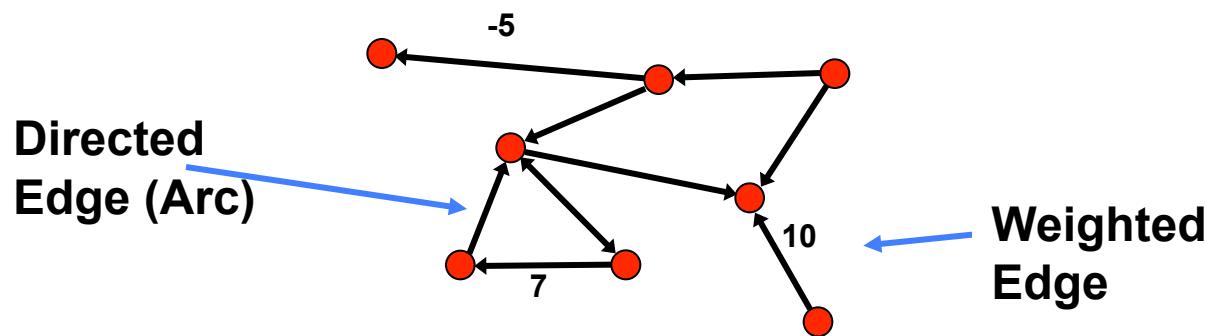
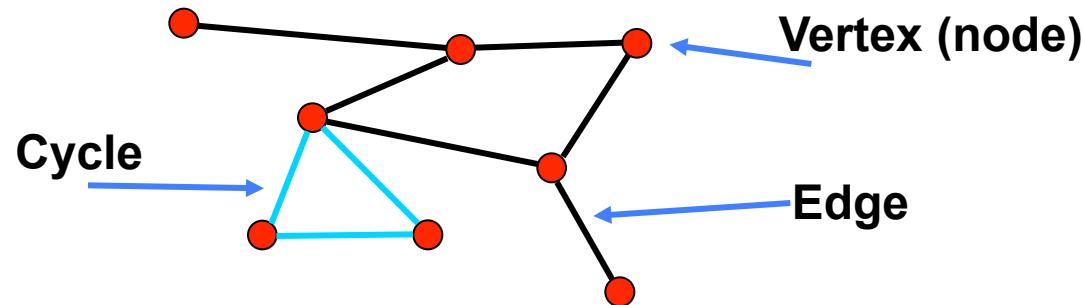
Computationally Query to Answer Specific Questions

- Visualize multiple data types on a network
- Cluster, Find active pathways, Compare, Search

Eureka! New pathway gene!



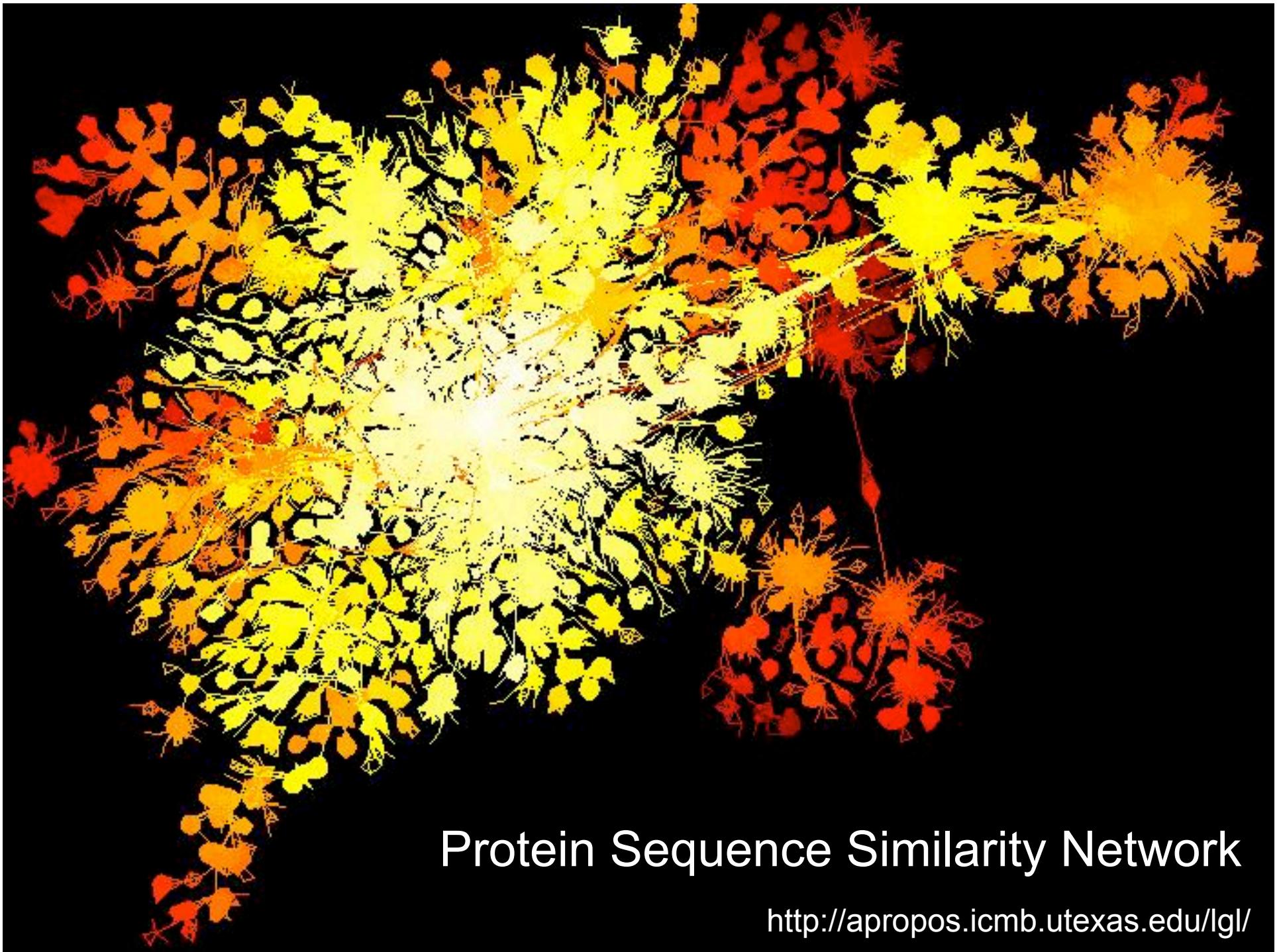
Graph Theory



We map molecular interaction networks to graphs

Mapping Biology to a Network

- A simple mapping
 - one compound/node, one interaction/edge
- A more realistic mapping
 - Cell localization, cell cycle, cell type, taxonomy
 - Only represent physiologically relevant interaction networks
- Edges can represent other relationships
- **Critical:** understand the mapping for network analysis

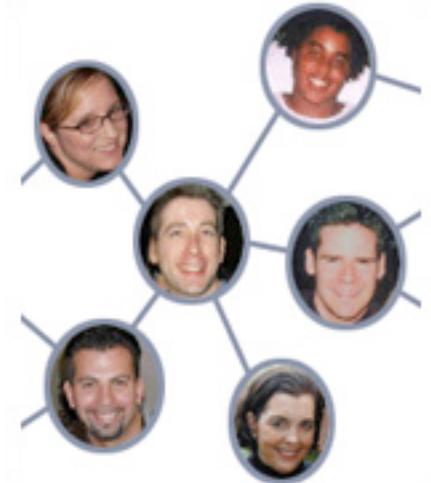


Protein Sequence Similarity Network

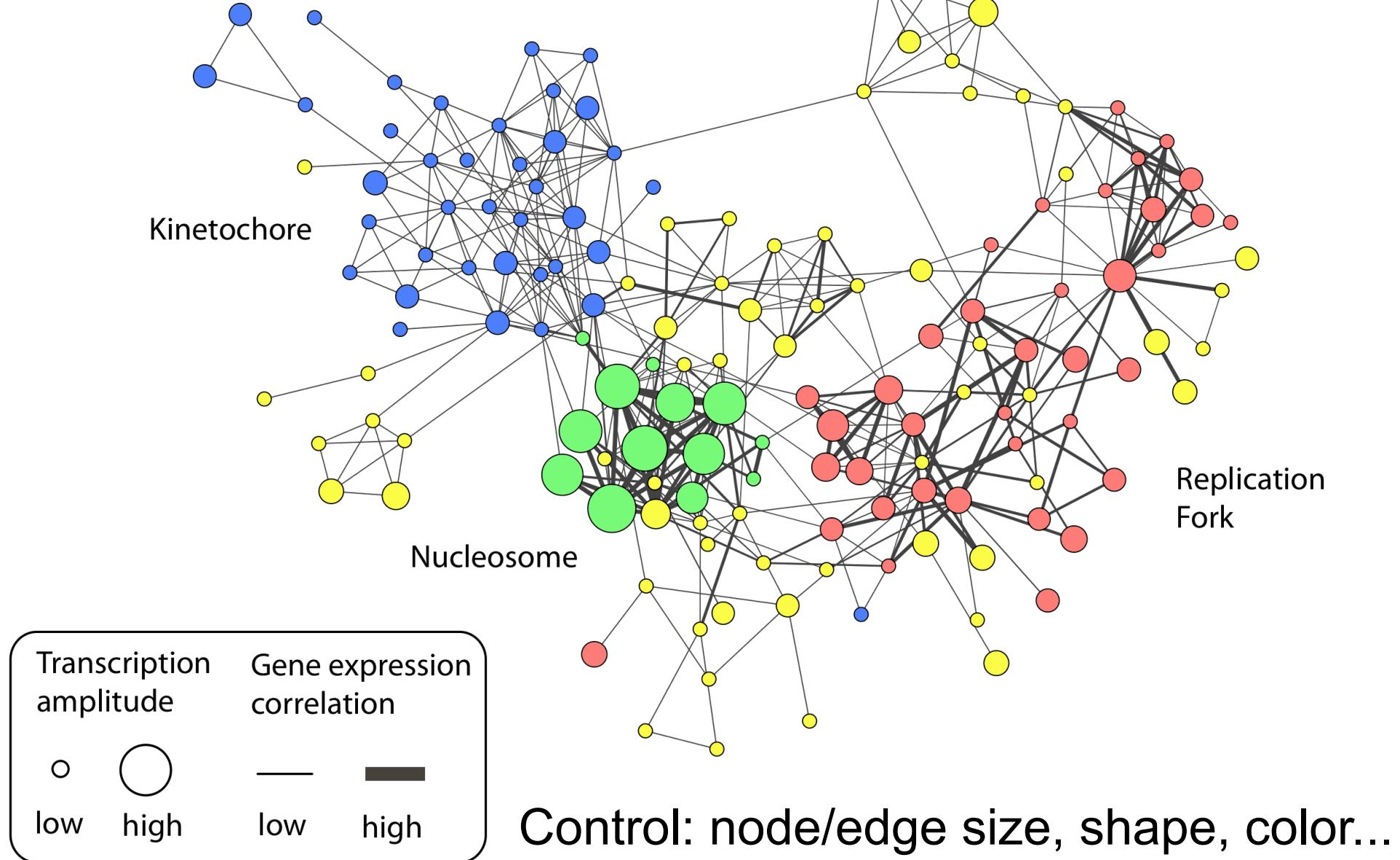
<http://apropos.icmb.utexas.edu/lgl/>

Six Degrees of Separation

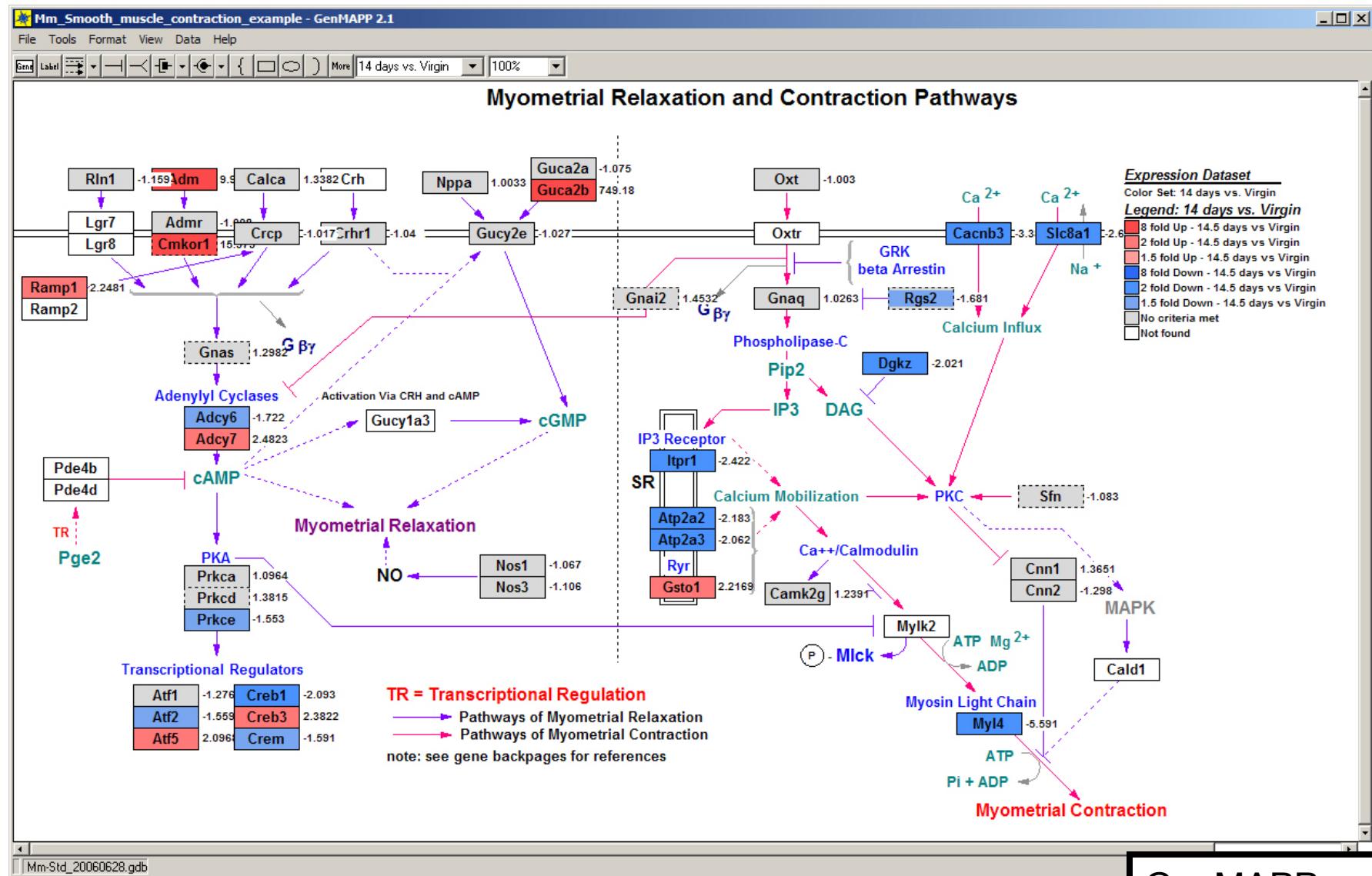
- Everyone in the world is connected by at most six links
- Which path should we take?
- Shortest path by breadth first search
 - If two nodes are connected, will find the shortest path between them
- Are two proteins connected? If so, how?
- Biologically relevant?



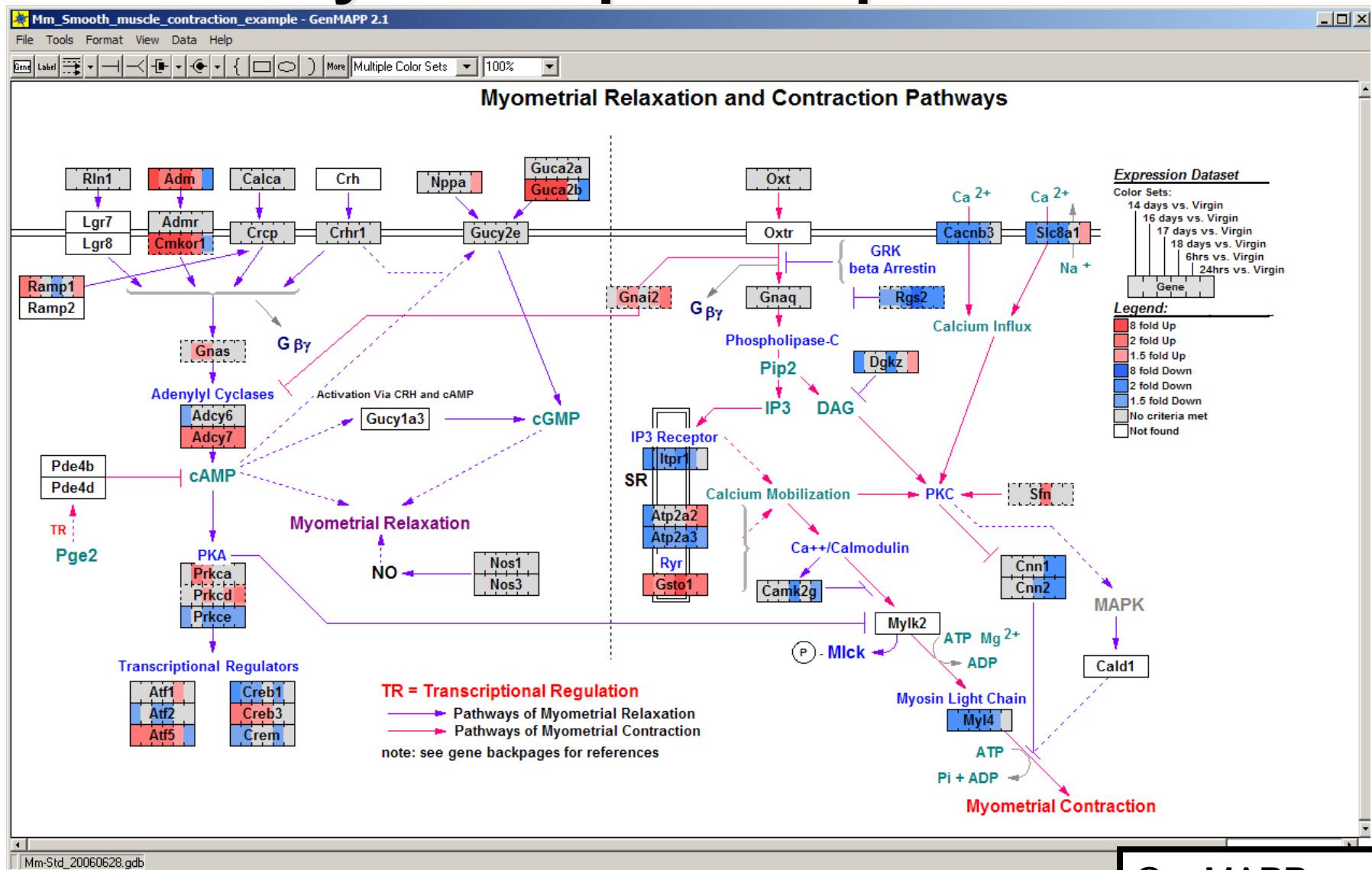
Visualize multiple data types on a network



Visualizing Time Course Data on Pathways: Single Comparison View

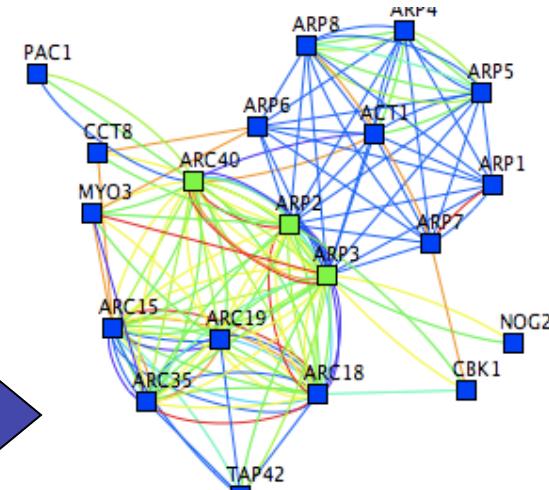


Visualizing Time Course Data on Pathways: Multiple Comparison View



Predicting Gene Function

arp2
arp3
arc40



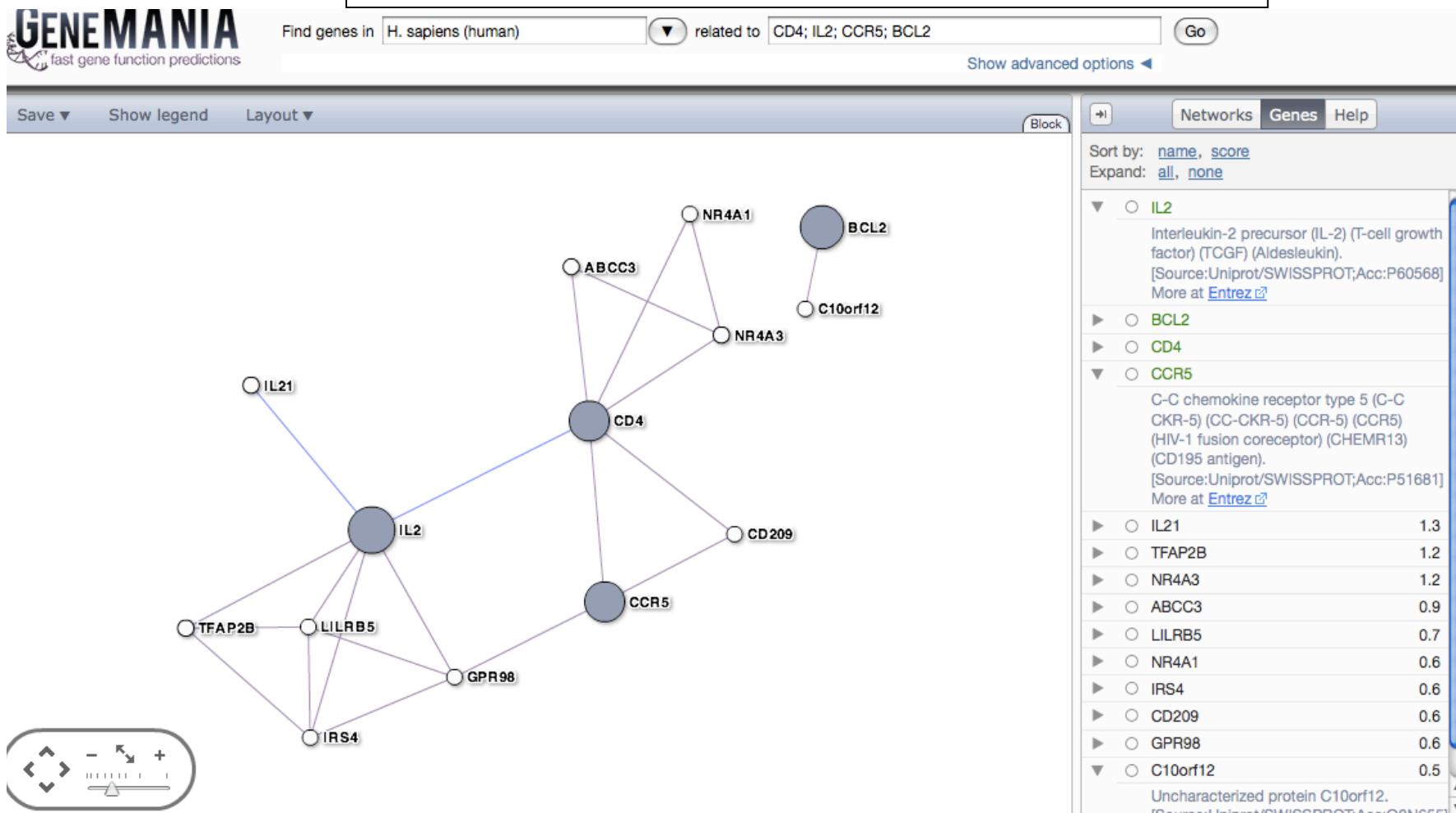
- STRING
 - <http://string.embl.de/>
- bioPIXIE
 - <http://pixie.princeton.edu/pixie/>
- GeneMania
 - <http://www.genemania.org>

Top-Scoring Genes

ARC15	0.09026
ARC19	0.08677
ARC35	0.08414
ARC18	0.07793
ARC40	0.03239
ARP8	0.02344
ARP5	0.02293
ARP6	0.02031
TAP42	0.02017
ACT1	0.01854
ARP4	0.01841
ARP1	0.01752
NOG2	0.01676
PAC1	0.01563
ARP7	0.01561
MYO3	0.01551

Fraser AG, Marcotte EM - A probabilistic view of gene function - Nat Genet. 2004 Jun;36(6):559-64

<http://www.genemania.org>



[About](#), [CCBR](#), © 2009

- Guilt-by-association principle
- Biological networks are combined intelligently to optimize prediction accuracy
- Algorithm is more fast and accurate than its peers

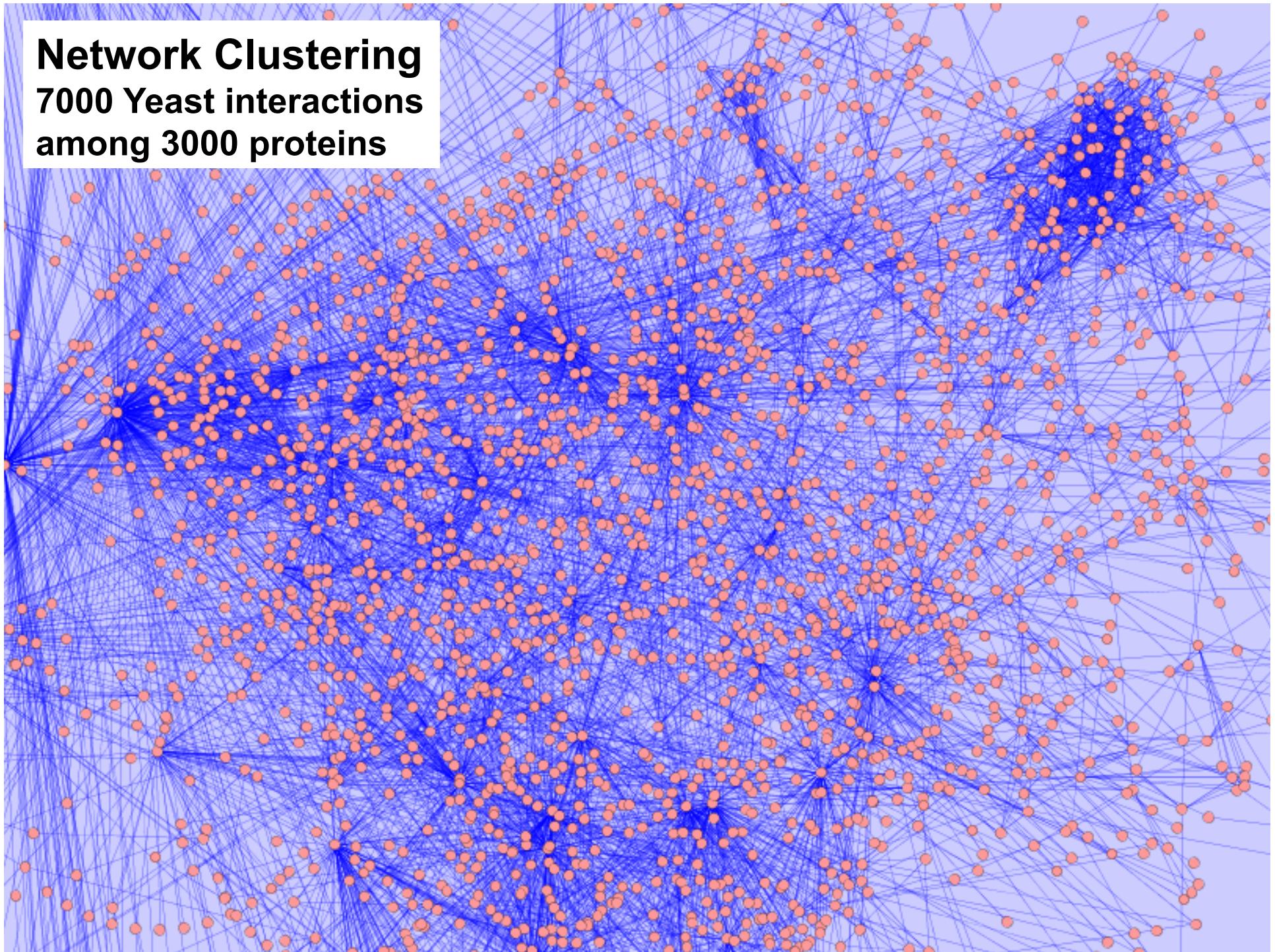
Gene Function Prediction

Quaid Morris (CCBR)
Rashad Badrawi, Ovi Comes, Sylva Donaldson,
Christian Lopes, Farzana Kazi, Jason Montojo,
Harold Rodriguez, Khalid Zuberi

Graph Clustering - MCODE Plugin

- Clusters in a protein-protein interaction network have been shown to represent protein complexes and parts of pathways
- Clusters in a protein similarity network represent protein families
- Network clustering is available through the MCODE Cytoscape plugin

Network Clustering
7000 Yeast interactions
among 3000 proteins

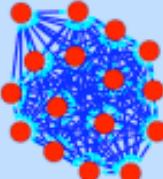
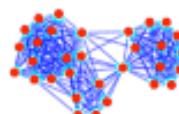
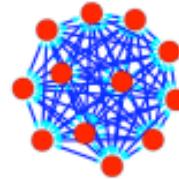
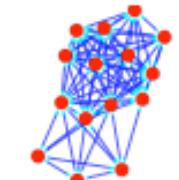
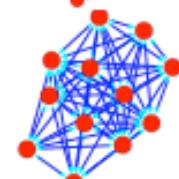


MCODE Results Summary				
Rank	Score	Size	Names	Complex
1	7.25	16,116	YGR232W, YDL007W, YKL145W, YFR052W, YFR004W, YLR421C, YOR261C, YDL147W, YDR427W, YHR200W, YER021W, YOR117W, YDL097C, YOR259C, YPR108W, YDR394W YPL093W, YBL004W, YOR272W, YNL110C, YKL009W, YFL002C, YOL077C, YPL126W, YIL035C, YLR409C, YLR129W, YOR061W, YKR060W, YCR057C, YDR449C, YOR039W, YJL109C, YPL012W, YGR103W, YLR449W, YOR206W, YKL014C, YLL008W, YKL172W, YNL002C, YLR002C, YGL111W, YOL041C, YGL019W, YOR145C, YPR016C YGL011C, YOL038W, YPR103W, YMR314W, YBL041W, YOR362C, YER012W, YJL001W, YML092C, YGR253C, YER094C, YGR135W	
2	6.387	31,198	YPL043W, YMR290C, YER006W, YKR081C, YDR496C, YDL031W, YNL061W, YNL132W, YLR222C, YLR197W, YMR049C, YHR052W, YJL069C, YKL099C, YDL014W, YPR187W, YPR010C, YPR110C, YNL248C, YOR341W, YNR003C, YKL144C, YOR207C, YPR190C, YNL113W, YOR116C, YBR154C	
3	5.417	12,65	YPL043W, YMR290C, YER006W, YKR081C, YDR496C, YDL031W, YNL061W, YNL132W, YLR222C, YLR197W, YMR049C, YHR052W, YJL069C, YKL099C, YDL014W, YPR187W, YPR010C, YPR110C, YNL248C, YOR341W, YNR003C, YKL144C, YOR207C, YPR190C, YNL113W, YOR116C, YBR154C	
4	5	15,75	YPL043W, YMR290C, YER006W, YKR081C, YDR496C, YDL031W, YNL061W, YNL132W, YLR222C, YLR197W, YMR049C, YHR052W, YJL069C, YKL099C, YDL014W, YPR187W, YPR010C, YPR110C, YNL248C, YOR341W, YNR003C, YKL144C, YOR207C, YPR190C, YNL113W, YOR116C, YBR154C	
5	5	12,60	YPL043W, YMR290C, YER006W, YKR081C, YDR496C, YDL031W, YNL061W, YNL132W, YLR222C, YLR197W, YMR049C, YHR052W, YJL069C, YKL099C, YDL014W, YPR187W, YPR010C, YPR110C, YNL248C, YOR341W, YNR003C, YKL144C, YOR207C, YPR190C, YNL113W, YOR116C, YBR154C	

Create a new child network.

Save

Done

MCODE Results Summary				
Rank	Score	Size	Names	Complex
1	7.25	16,116	YGR232W, YDL007W, YKL145W, YFR052W, YFR004W, YLR421C, YOR261C, YDL147W, YDR427W, YHR200W, YER021W, YOR117W, YDL097C, YOR259C, YPR108W, YDR394W	
Ribosome			YPL093W, YBL004W, YOR272W, YNL110C, YKL009W, YFL002C, YOL077C, YPL126W, YIL035C, YLR409C, YLR129W, YOR061W, YKR060W, YCR057C, YDR449C, YOR039W, YJL109C, YPL012W, YGR103W, YLR449W, YOR206W, YKL014C, YLL008W, YKL172W, YNL002C, YLR002C, YGL111W, YOL041C, YGL019W, YOR145C, YPR016C	
Proteasome 20S		31,198	YGL011C, YOL038W, YPR103W, YMR314W, YBL041W, YOR362C, YER012W, YJL001W, YML092C, YGR253C, YER094C, YGR135W	
RNA Splicing	3	5.417	YPL043W, YMR290C, YER006W, YKR081C, YDR496C, YDL031W, YNL061W, YNL132W, YLR222C, YLR197W, YMR049C, YHR052W, YJL069C, YKL099C, YDL014W	
RNA Pol core	4	12,65	YPR187W, YPR010C, YPR110C, YNL248C, YOR341W, YNR003C, YKL144C, YOR207C, YPR190C, YNL113W, YOR116C, YBR154C	
	5	15,75		
	5	12,60		

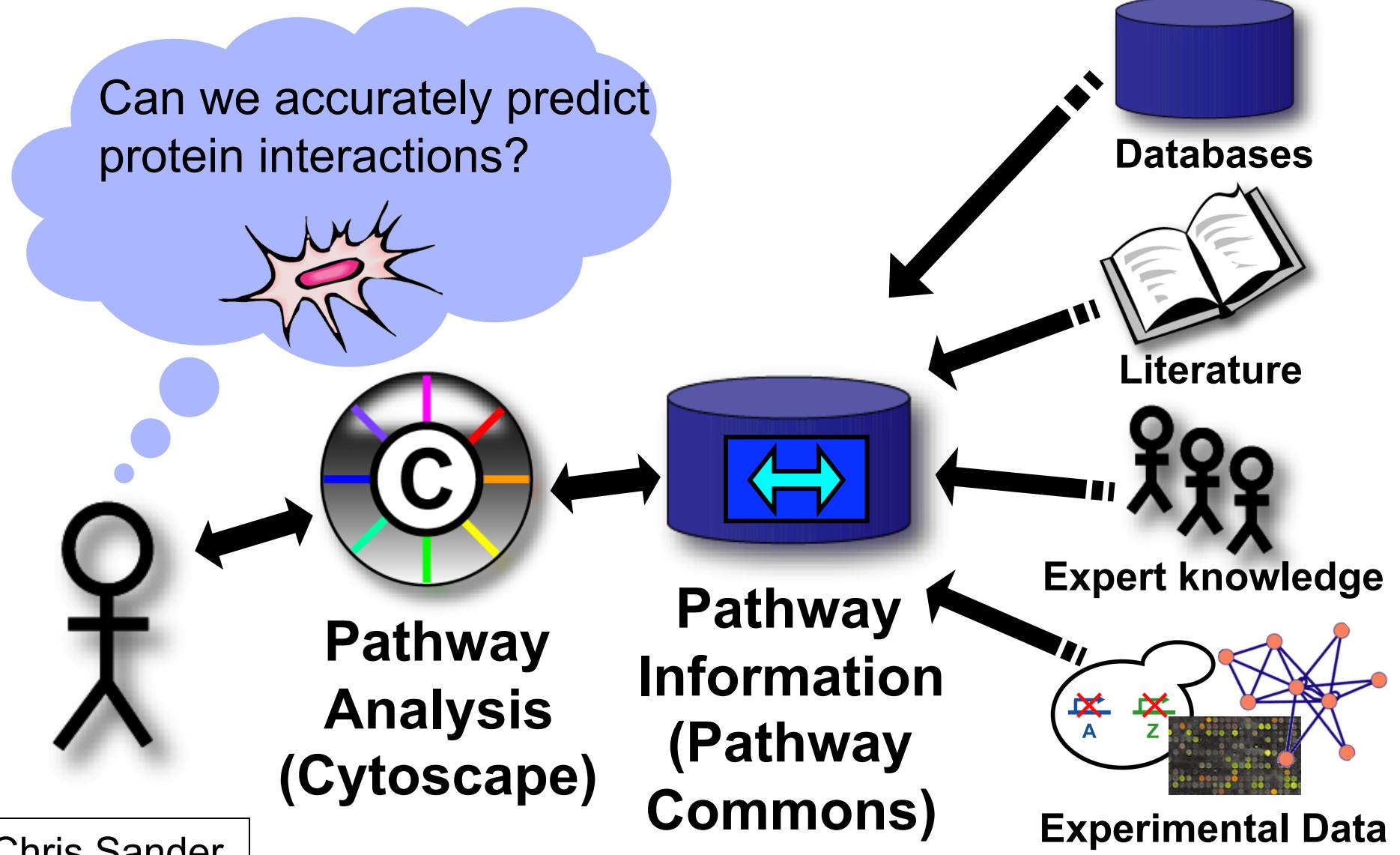
Create a new child network.

Save

Done

Network Data

Cell map exploration and analysis



>300 Pathway Databases!

Navigation	
Protein-Protein Interactions	
Metabolic Pathways	
Signaling Pathways	
Pathway Diagrams	
Transcription Factors / Gene Regulatory Networks	
Protein-Compound Interactions	
Genetic Interaction Networks	
Protein Sequence Focused	
Other	
Search	
Organisms	
All	
Availability	
All	
Standards	
All	
Reset	Search
Statistics	
Analyze Pathguide	
Contact	
Comments, Questions, Suggestions are Always Welcome!	

Complete Listing of All Pathguide Resources

Pathguide contains information about 222 biological pathway resources. Click on a link to go to the resource home page or 'Details' for a description page. Databases that are free and those supporting BioPAX, CellML, PSI-MI, or SBML standards are respectively indicated.

If you know of a pathway resource that is not listed here, or have other questions or comments, please [send us an e-mail](#).

Get the Stats
Detailed Pathguide resource statistics now available

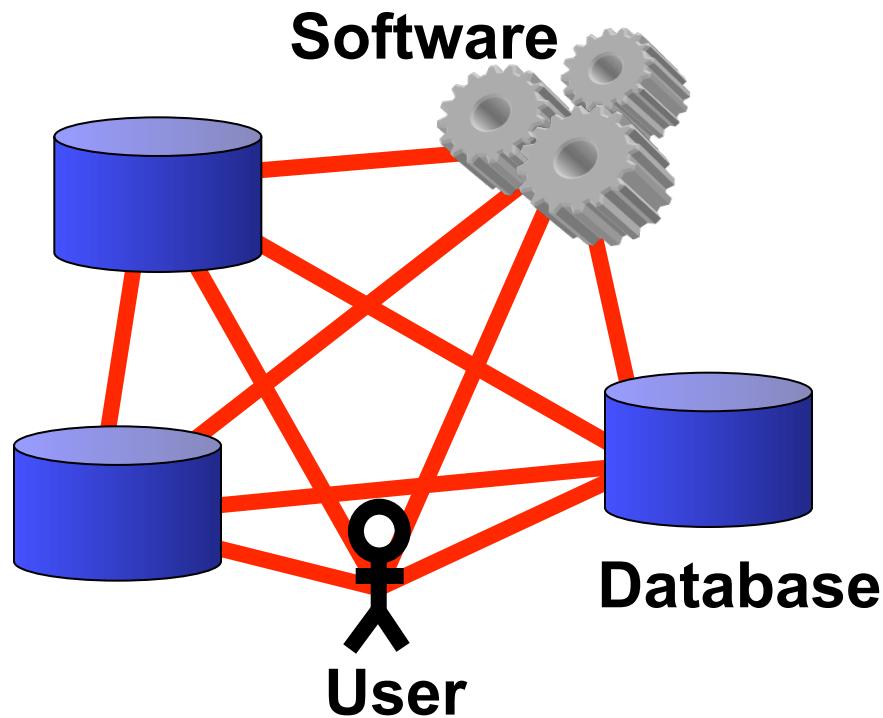
Pathguide Published
Please cite the [Pathguide](#)

Protein-Protein Interactions

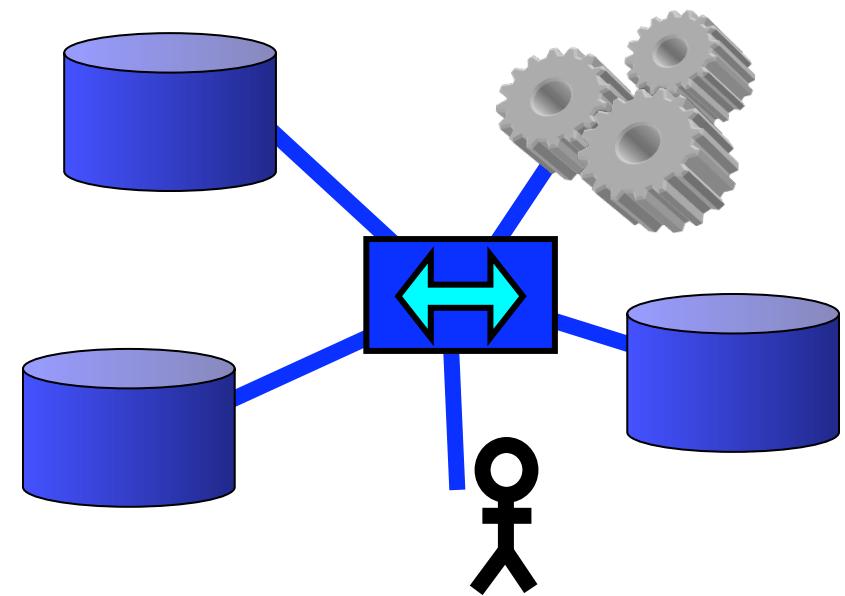
Database Name (Order: alphabetically by web popularity)	Full Record	Availability	Standards
3DID - 3D interacting domains	Details	Free	
ABCdb - Archaea and Bacteria ABC transporter database	Details	Free	
AfCS - Alliance for Cellular Signaling Molecule Pages Database	Details	Free	
AllFuse - Functional Associations of Proteins in Complete Genomes	Details	Free	
ASEdb - Alanine Scanning Energetics Database	Details	Free	
ASPD - Artificial Selected Proteins/Peptides Database	Details	?	
BID - Binding Interface Database	Details	Free	
BIND - Biomolecular Interaction Network Database	Details	Free	PSI-MI
BindingDB - The Binding Database	Details	Free	
BioGRID - General Repository for Interaction Datasets	Details	book	PSI-MI
BRITE - Biomolecular Relations in Information Transmission and Expression	Details	Free	
CA1Neuron - Pathways of the hippocampal CA1 neuron	Details	Free	
Cancer Cell Map - The Cancer Cell Map	Details	Free	BioPAX
CSP - Cytokine Signaling Pathway Database	Details	Free	
CTDB - Calmodulin Target Database	Details	Free	
DDIB - Database of Domain Interactions and Bindings	Details	Free	
DIP - Database of Interacting Proteins	Details	book	PSI-MI
Doodle - Database of oligomerization	Details	Free	
DopaNet - DopaNet	Details	Free	
DRC - Database of Ribosomal Cache	Details	Free	
DSM - Dynamic Signaling Maps	Details	Free	
FIMM - Functional Molecular Interaction Map	Details	Free	
FusionDB - Prokaryote Gene Fusion Database	Details	Free	

- Varied formats, representation, coverage
- Pathway data extremely difficult to combine and use

Solution: Standard Exchange Formats



>100 DBs and tools
Tower of Babel



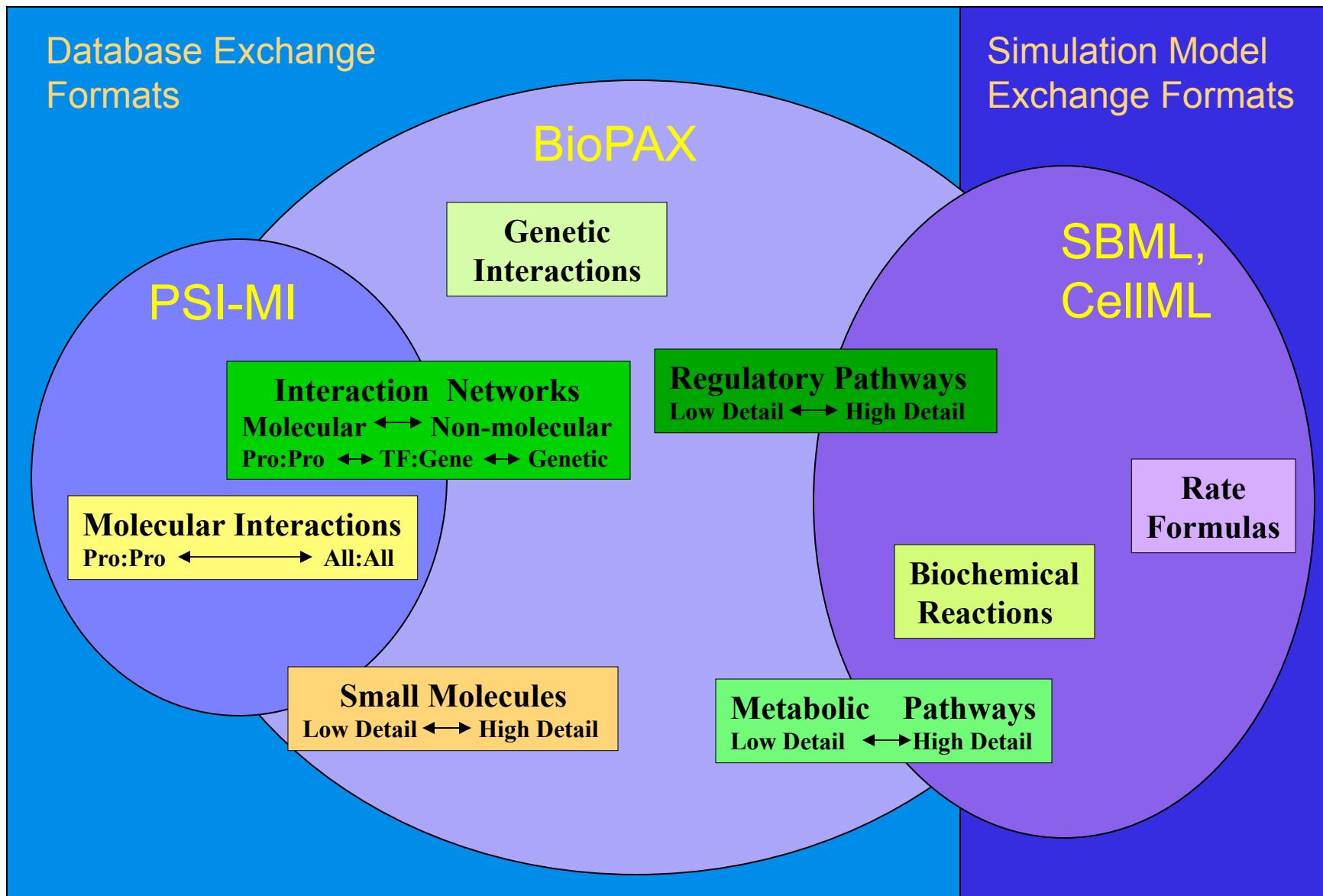
With Data
Exchange Format

Reduces work, promotes collaboration, increases accessibility

Interaction and Pathway Data Exchange Formats

- **PSI-MI** <http://psidev.sourceforge.net>
 - Molecular interactions - protein-protein interaction focus
 - Peer reviewed, HUPO community standard
- **BioPAX** <http://www.biopax.org>
 - Biological pathways
 - Community ontology in OWL, Protégé
- **SBML** <http://www.sbml.org>
 - Widely adopted for representing mathematical models of biological processes e.g. biochemical reaction networks
- **CellML** <http://www.cellml.org>
 - Math models of biological processes

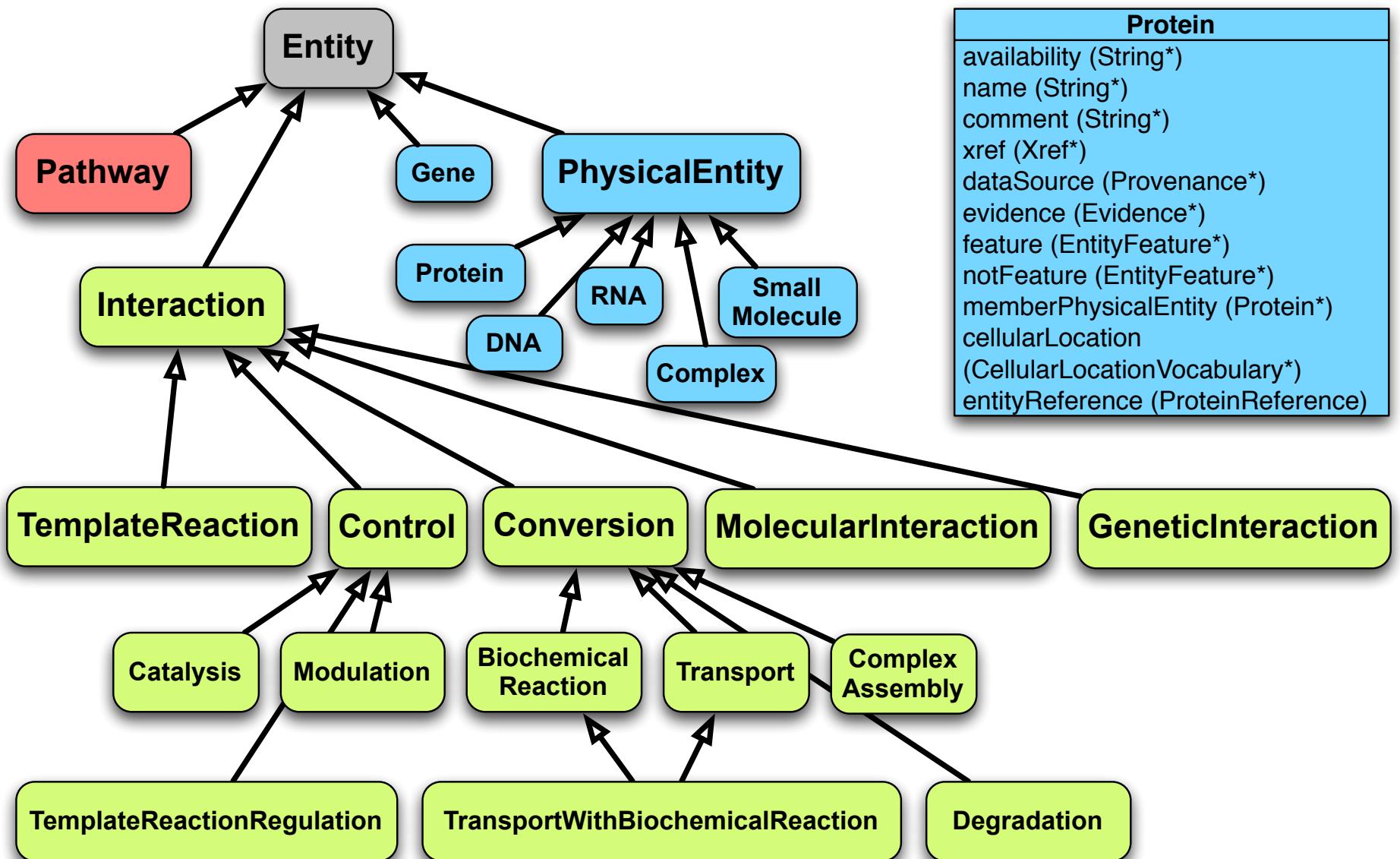
Biological Network Exchange Formats

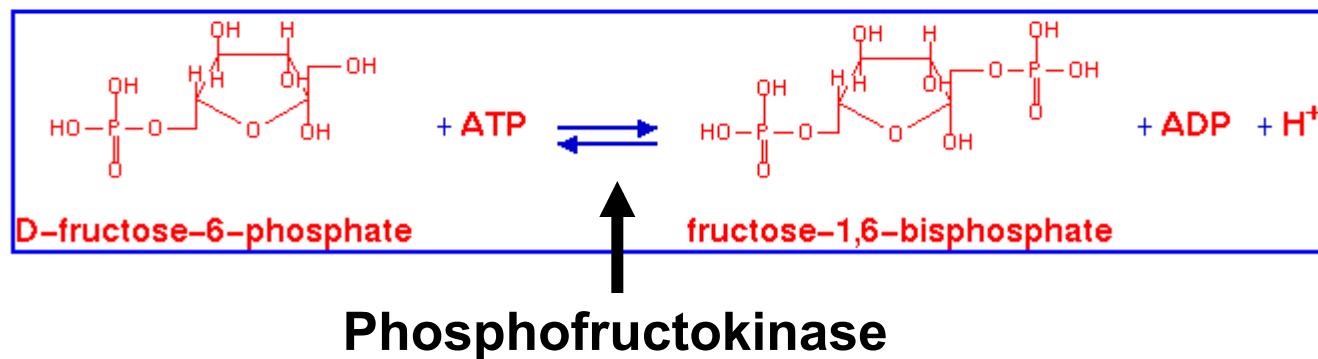


BioPAX Pathway Language

- Represent:
 - Metabolic pathways
 - Signaling pathways
 - Protein-protein, molecular interactions
 - Gene regulatory pathways
 - Genetic interactions
- Community effort: pathway databases distribute pathway information in standard format
 - Over 100 people, database groups, standard efforts

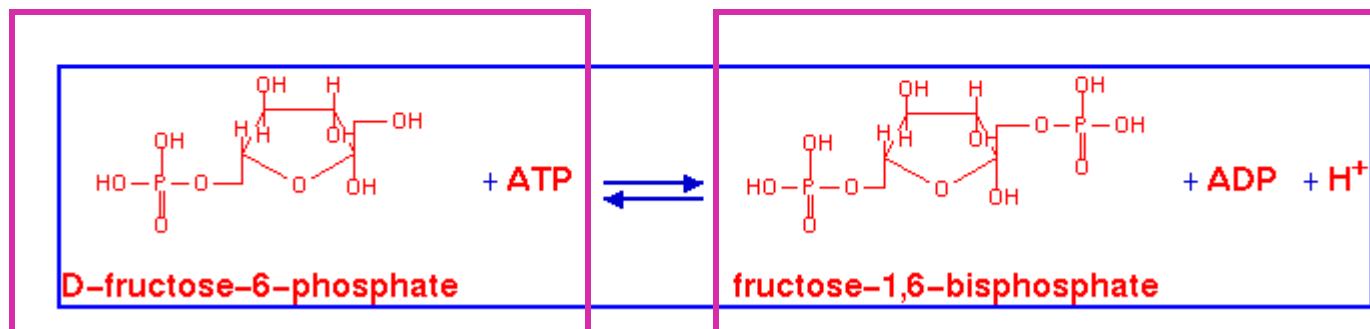
BioPAX Class Structure





Biochemical Reaction Glycolysis Pathway

Source: BioCyc.org

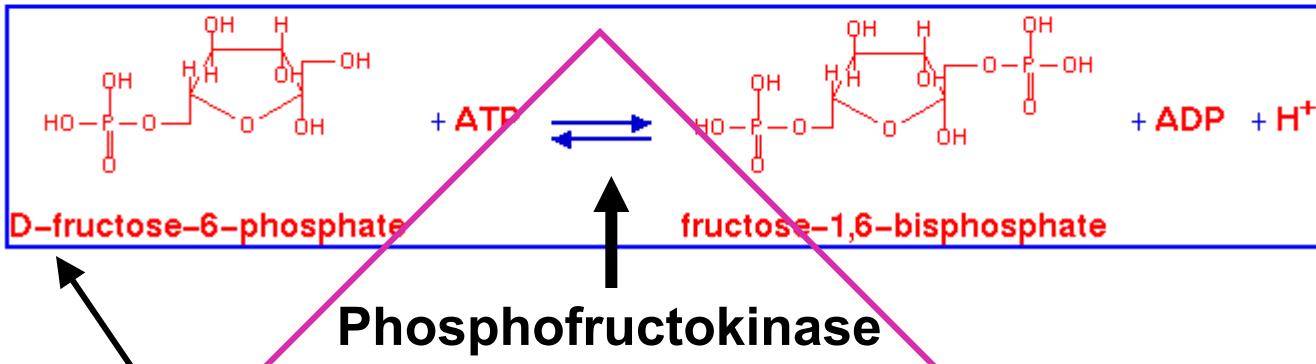


Left

Right

biochemicalReaction	
☐	PARTICIPANTS
☐	RIGHT
☐	SYNONYMS
☐	SHORT-NAME
☐	SPONTANEOUS
☐	COMMENT
☐	AVAILABILITY
☐	LEFT
☐	DATA-SOURCE
☐	NAME
☐	XREF
☐	DELTA-H
☐	DELTA-S
☐	EC-NUMBER
☐	KEQ
☐	DELTA-G

EC # 2.7.1.11



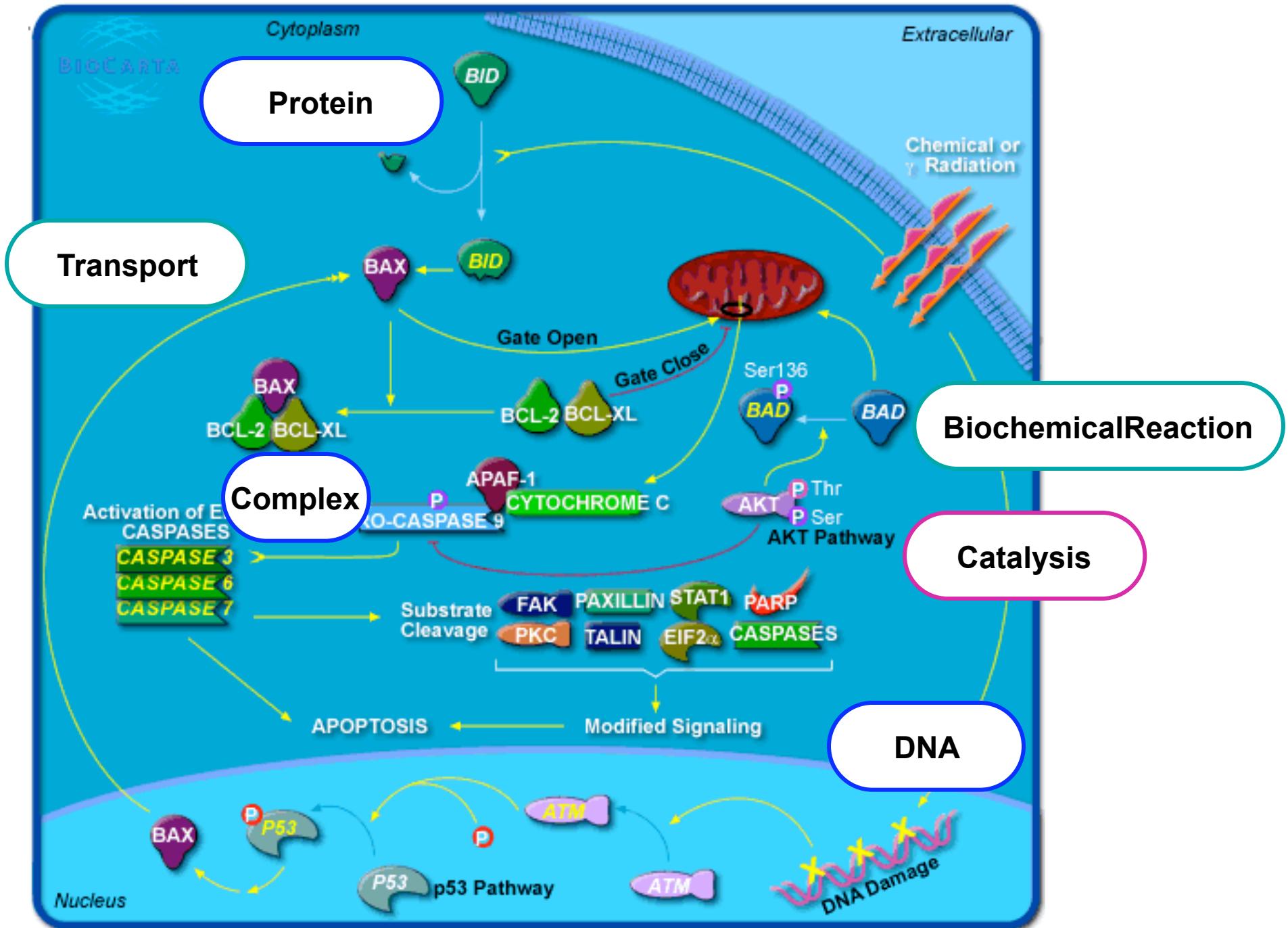
Phosphofructokinase

Controller

Controlled

catalysis	
0	CONTROLLED
0	COMMENT
0	PARTICIPANTS
0	AVAILABILITY
0	CONTROL-TYPE
0	DATA-SOURCE
0	CONTROLLER
0	SYNONYMS
0	SHORT-NAME
0	NAME
0	XREF
0	DIRECTION
0	COFACTOR

Direction: reversible



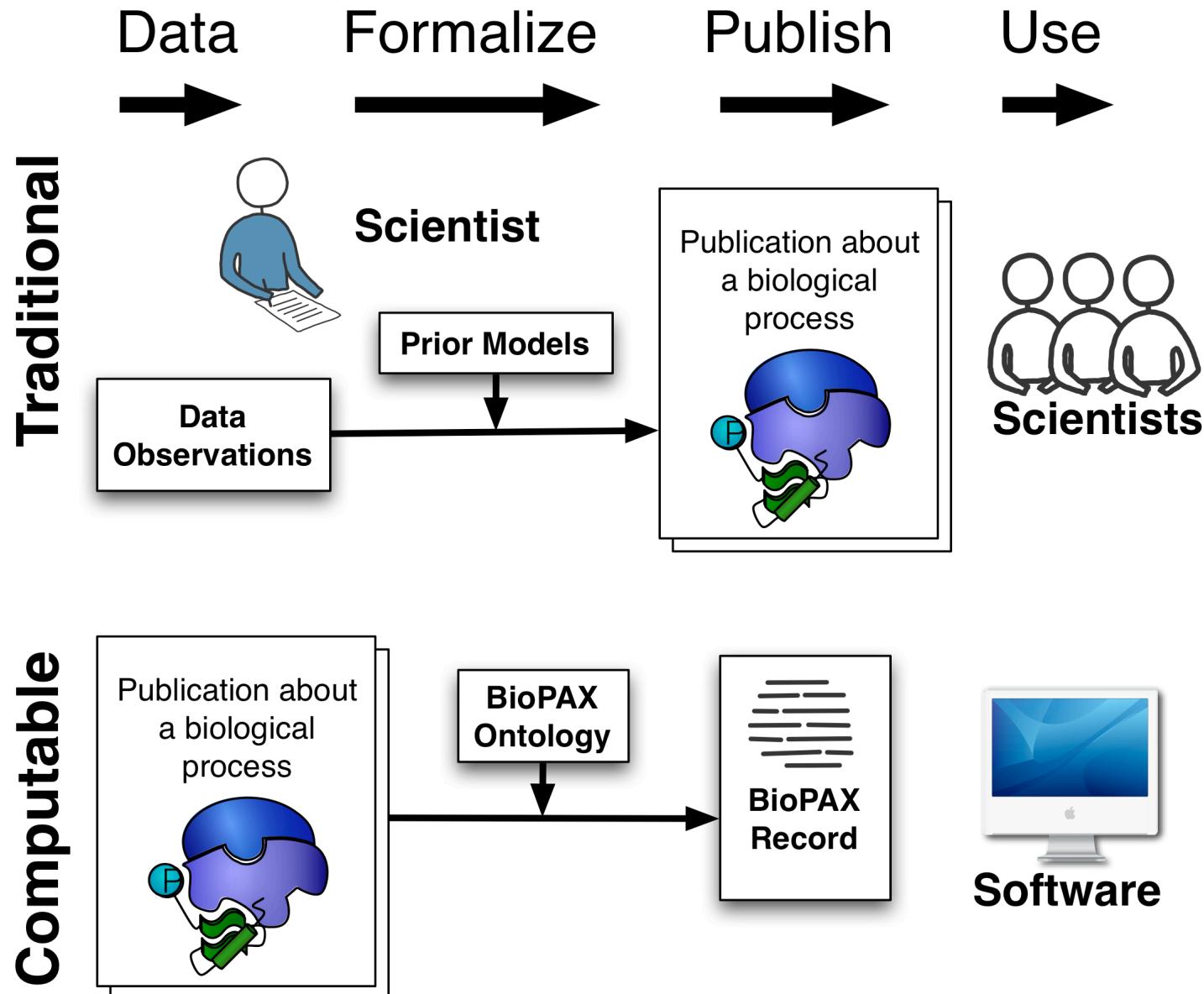
Controlled Vocabularies (CVs)

- BioPAX uses existing CVs where available via openControlledVocabulary instances
 - Cellular location: Gene Ontology (GO) component
 - PSI-MI CVs for:
 - Protein post-translational modifications
 - Interaction detection experimental methods
 - Experimental form
 - PATO phenotypic quality ontology
 - Some database providers use their own CVs
 - E.g. BioCyc evidence codes
- More at the Ontology Lookup Service
 - <http://www.ebi.ac.uk/ontology-lookup/>

OWL (XML) Snippet

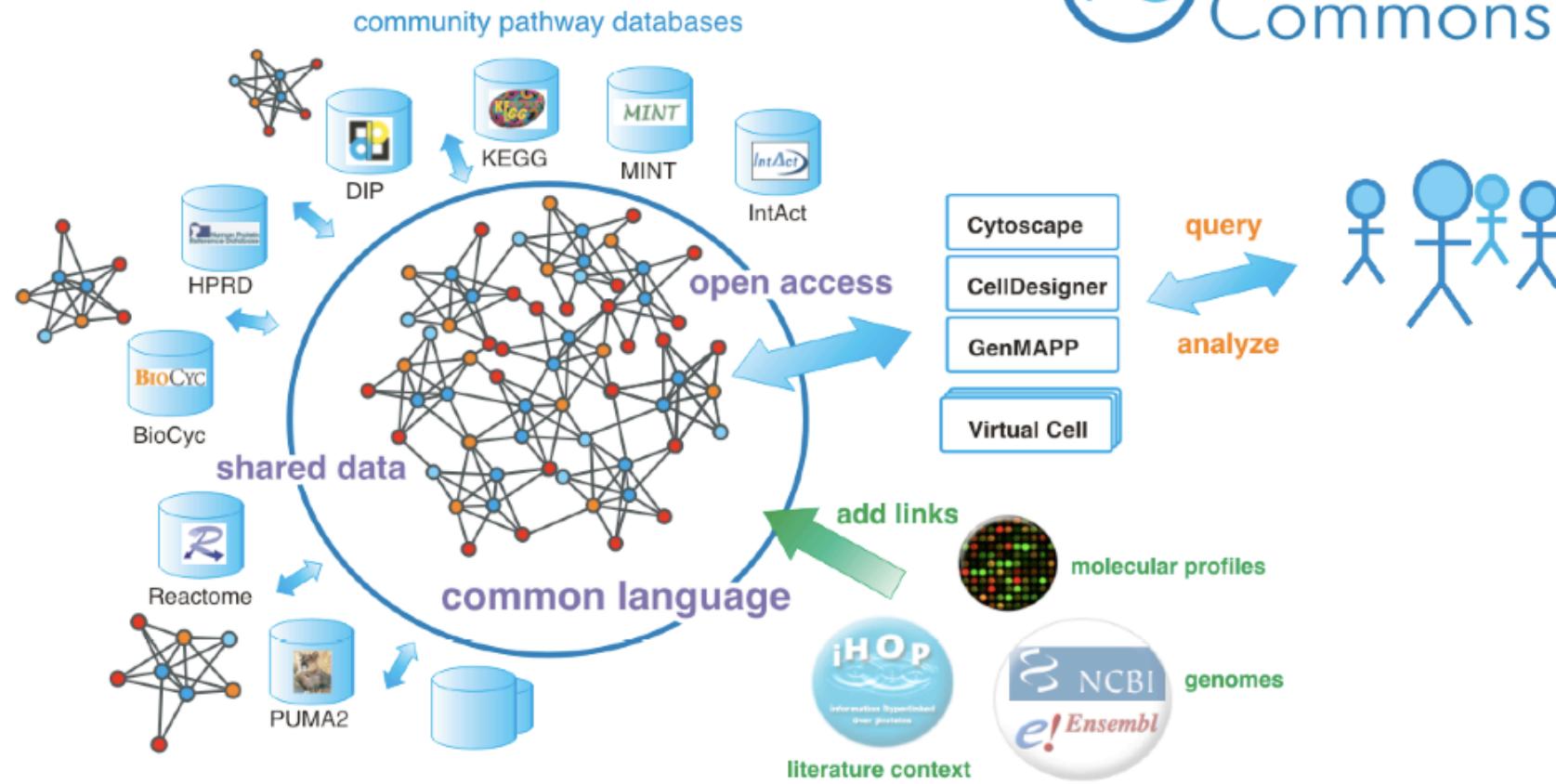
```
<bp:biochemicalReaction rdf:ID="biochemicalReaction37">
  <bp:DATA-SOURCE rdf:resource="#datasource14"/>
  <bp:LEFT>
    <bp:physicalEntityParticipant rdf:ID="physicalEntityParticipant26">
      <bp:STOICHIOMETRIC-COEFFICIENT>1.0</bp:STOICHIOMETRIC-COEFFICIENT>
      <bp:PHYSICAL-ENTITY>
        <bp:smallMolecule rdf:ID="smallMolecule27">
          <bp:SHORT-NAME rdf:datatype="http://www.w3.org/2001/XMLSchema#string"
            >a-D-glu-6-p</bp:SHORT-NAME>
          <bp:CHEMICAL-FORMULA rdf:datatype="http://www.w3.org/2001/XMLSchema#string"
            >C6H13O9P</bp:CHEMICAL-FORMULA>
          <bp:SYNONYMS rdf:datatype="http://www.w3.org/2001/XMLSchema#string"
            >&lt;FONT FACE="Symbol">a&lt;/FONT>-D-glucose-6-phosphate</bp:SYNONYMS>
          <bp:XREF>
            <bp:unificationxref rdf:ID="unificationxref30">
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                >C00668</bp:ID>
              <bp:DB rdf:datatype="http://www.w3.org/2001/XMLSchema#string"
                >KEGG</bp:DB>
            </bp:unificationxref>
          </bp:XREF>
        <bp:XREF rdf:resource="#unificationxref29"/>
        <bp:MOLECULAR-WEIGHT>260.14</bp:MOLECULAR-WEIGHT>
        <bp:AVAILABILITY rdf:datatype="http://www.w3.org/2001/XMLSchema#string"
          >see http://www.amaze.ulb.ac.be/</bp:AVAILABILITY>
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        <bp:SYNONYMS rdf:datatype="http://www.w3.org/2001/XMLSchema#string"
          >alpha-D-glucose-6-p</bp:SYNONYMS>
        <bp:STRUCTURE>
          <bp:chemicalStructure rdf:ID="chemicalstructure28">
            <bp:STRUCTURE-FORMAT>SMILES</bp:STRUCTURE-FORMAT>
            <bp:STRUCTURE-DATA>C(O[P(=O)(O)O])[CH]1([CH](O)[CH](O)[CH](O)[CH](O)O1)</bp:STRUCTURE-DATA>
          </bp:chemicalStructure>
        </bp:STRUCTURE>
        <bp:NAME>alpha-D-glucose 6-phosphate</bp:NAME>
        <bp:SYNONYMS rdf:datatype="http://www.w3.org/2001/XMLSchema#string"
          >alpha-D-glucose-6-phosphate</bp:SYNONYMS>
        <bp:SYNONYMS rdf:datatype="http://www.w3.org/2001/XMLSchema#string"
          > D-glucose-6-P</bp:SYNONYMS>
        <bp:DATA-SOURCE rdf:resource="#KB_439584_Individual_47"/>
        </bp:smallMolecule>
      </bp:PHYSICAL-ENTITY>
      <bp:CELLULAR-LOCATION rdf:resource="#openControlledvocabulary15"/>
    </bp:physicalEntityParticipant>
  </bp:LEFT>
  <bp:DELTA-G rdf:datatype="http://www.w3.org/2001/XMLSchema#double"
    >0.4</bp:DELTA-G>
  <bp:SYNONYMS rdf:datatype="http://www.w3.org/2001/XMLSchema#string"
    >alpha-D-Glucose 6-phosphate &lt;=> beta-D-Fructose 6-phosphate </bp:SYNONYMS>
  <bp:RIGHT>
    <bp:physicalEntityParticipant rdf:ID="physicalEntityParticipant38">
      <bp:CELLULAR-LOCATION rdf:resource="#openControlledvocabulary15"/>
    <bp:PHYSICAL-ENTITY>
      <bp:smallMolecule rdf:ID="smallMolecule39">
```

Pathway Information Processing



Aim: Convenient Access to Pathway Information

<http://www.pathwaycommons.org>



Facilitate creation and communication of pathway data
Aggregate pathway data in the public domain
Provide easy access for pathway analysis

Long term: Converge to integrated cell map

Pathway Commons

Search

Home | Filter | FAQ | About | Credits

Pathway Commons is a convenient point of access to biological pathway information collected from public pathway databases, which you can browse or search. Pathways include biochemical reactions, complex assembly, transport and catalysis events, and physical interactions involving proteins, DNA, RNA, small molecules and complexes. [more...](#)

Pathway Commons Quick Stats:

Number of Pathways:	921
Number of Interactions:	9,924
Number of Physical Entities:	15,515
Number of Organisms:	10

Search Pathway Commons:

To get started, enter a gene name, gene identifier or pathway name in the text box above. For example: [p53](#), [P38398](#) or [mTOR](#).

To restrict your search to specific data sources or specific organisms, update your [global filter settings](#).

Pathway Commons currently contains the following data sources:

 [Cancer Cell Map, Release: 1.0 \[19-May-06\]](#)
[Browse](#)

 [HumanCyc, Release: 10.5 \[18-Sep-06\]](#)
[Browse](#)

 [NCI / Nature Pathway Interaction Database](#)
[01-Jan-07]
[Browse](#)

 [Reactome, Release: 19 \[16-Nov-06\]](#)
[Browse](#)

Pathway Commons is hosted by the [Computational Biology Center](#) at [Memorial Sloan-Kettering Cancer Center](#) and the [University of Toronto](#).

Pathway Commons is a work in progress. We welcome your feedback. Email us at: pc-info@pathwaycommons.org.

[Home](#) | [Filter](#) | [FAQ](#) | [About](#) | [Credits](#) | [Results](#)

Searched for: p53

Pathway Commons completed your search for "p53" and found **22** relevant records:

Narrow Results by Type:

- [All Types \(45\)](#)
- [Pathway \(22\)](#)
- [Protein \(23\)](#)

Narrow Results by Data Source:

- [All Data Sources \(22\)](#)
- [Cancer Cell Map \(2\)](#)
- [NCI / Nature Pathway Interaction Database \(3\)](#)
- [Reactome \(17\)](#)

[\[Update Filter Settings\]](#)

Showing Results 1 - 10 of 22 | [Next 10](#)

[Pathway: Transcriptional activation of p53 responsive genes](#)

Summary:

p53 causes G1 arrest by inducing the expression of a cell cycle inhibitor, p21 (El-Deiry et al, 1993; Harper et al, 1993; Xiong et al, 1993). P21 binds and inactivates Cyclin-Cdk complexes that mediate G1/S progression, resulting in lack of phosphorylation of Rb, E2F sequestration and cell cycle arrest at the G1/S transition. Mice with a homozygous deletion of p21 gene are deficient in their ability to undergo a G1/S arrest in response to DNA damage (Deng et al, 1995).

Data Sources:

- Reactome

- ... **p53** causes G1 arrest by inducing the expression of a cell cycle inhibitor, p21 (El-Deiry et al, 1993; Harper et al, 1993; Xiong et al, 1993).

[Pathway: Stabilization of p53](#)

- ... ATM also regulates the phosphorylation of **p53** at other sites, especially Ser-20, by activating other serine/threonine kinases in response to IR (Chehab et al, 2000 ...)

[Pathway: p53-Dependent G1 DNA Damage Response](#)

- Most of the damage-induced modifications of **p53** are dependent on the ATM kinase. ... The first link between ATM and **p53** was predicted based on the earlier studies that showed that AT cells exhibit a reduced and delayed induction of **p53** following exposure to IR (Kastan et al, 1992 and Khanna and Lavin, 1993). ... **p53** is a short-lived protein ...

[Pathway: p53-Dependent G1/S DNA damage checkpoint](#)

- The arrest at G1/S checkpoint is mediated by the action of a widely known tumor suppressor protein, **p53**. ... Loss of **p53** functions, as a result of mutations in cancer prevent the G1/S checkpoint (Kuerbitz et al, 1992). ... **P53** is rapidly induced in response to damaged DNA.

[Pathway: p53-Independent G1/S DNA damage checkpoint](#)

- The G1 arrest induced by DNA damage has been ascribed to the transcription factor and tumor suppressor protein **p53**.

[Pathway: G1/S DNA Damage Checkpoints](#)

- In the G1 phase there are two types of DNA damage responses, the p53-dependent and the p53-independent pathways. ... The p53-dependent responses inhibit CDKs through the up-regulation of genes encoding CKIs mediated by the **p53** protein, whereas the p53-independent mechanisms inhibit CDKs through the inhibitory T14Y15 phosphorylation of Cdk2.

[Pathway: Cell Cycle Checkpoints](#)

<http://pathwaycommons.org>

Pathway Commons Status

Pathway Commons Quick Stats:

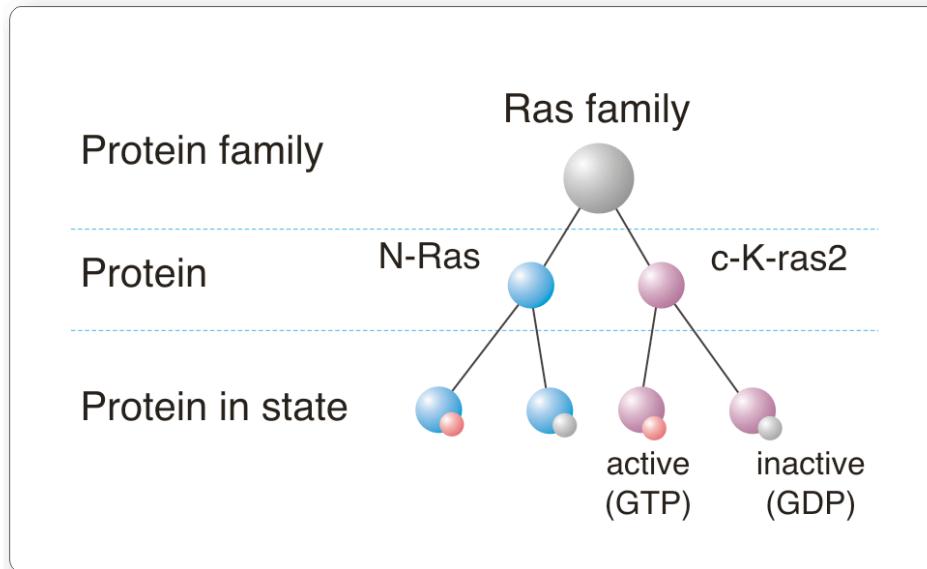
Number of Pathways: 1,449
Number of Interactions: 421,395
Number of Physical Entities: 88,509
Number of Organisms: 441

- Signaling
- Metabolism
- Molecular Interactions
- Future
 - Genetic Interactions
 - Gene Regulation



Towards an Integrated Cell Map

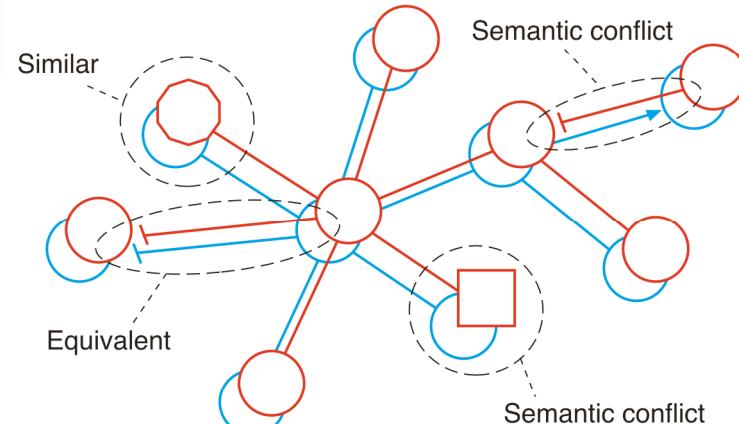
- Semantic pathway integration is difficult



Physical entities

Determining equivalent entities is critical

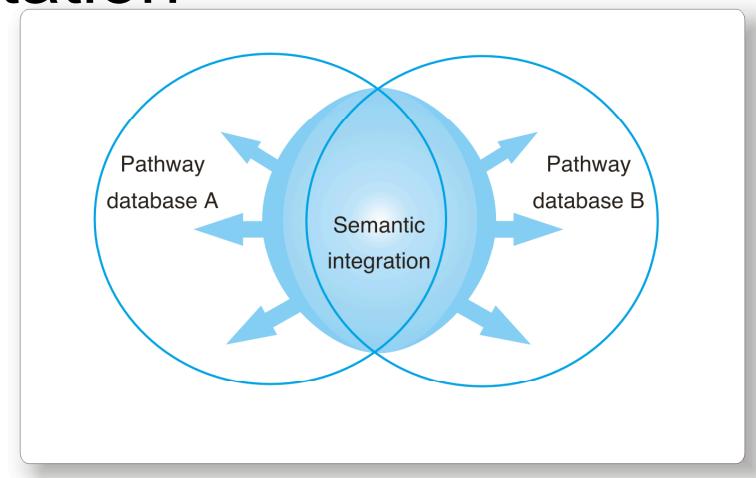
Relationships



Practical Semantic Integration

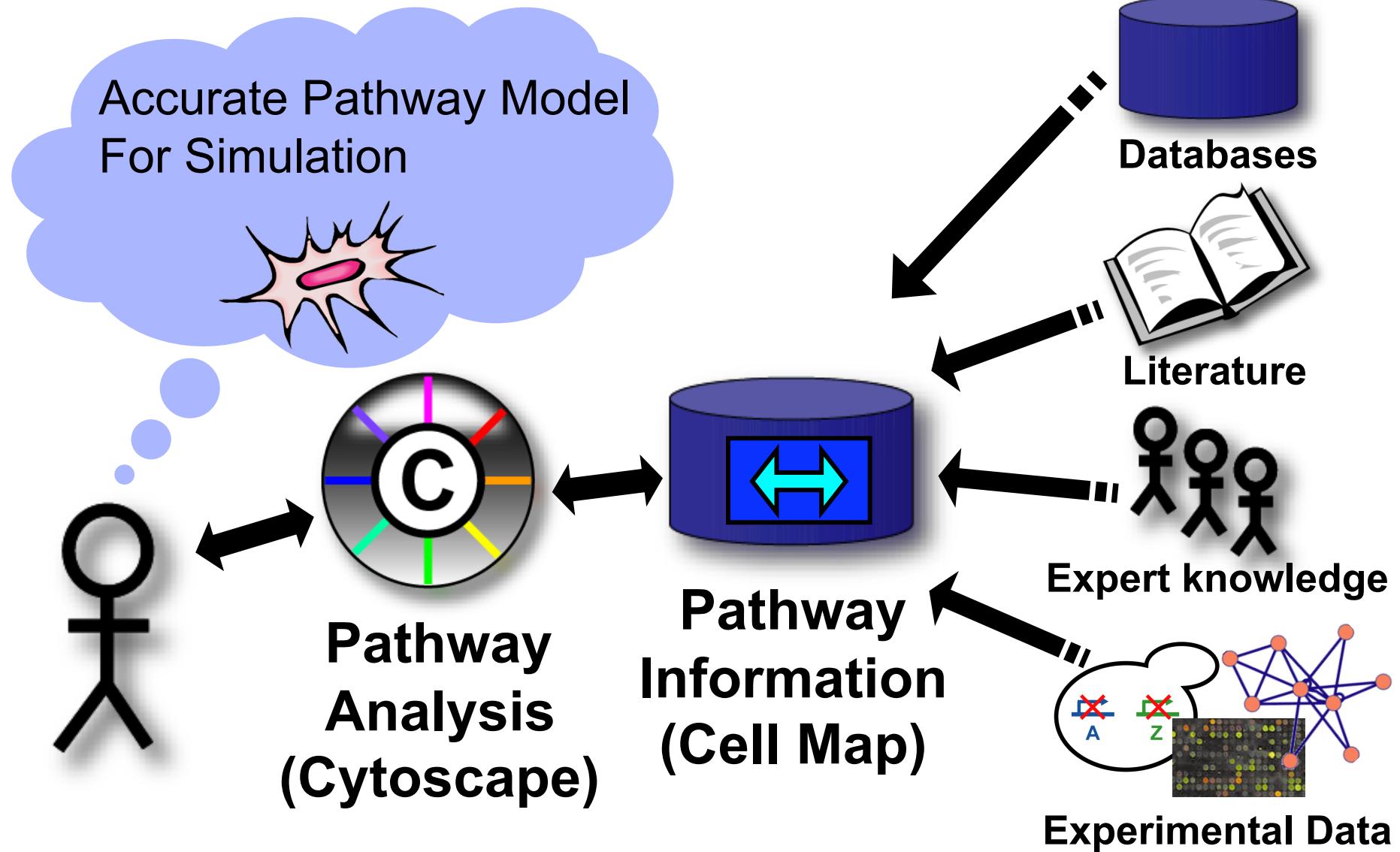
- Minimize errors
 - Integrate only where possible with high accuracy
 - Detect and flag conflicts, errors for users, no revision
 - Promote best-practices to minimize future errors
 - Interaction confidence algorithms
 - Validation software
 - Allow users to filter and select trusted sources
- Converge to standard representation
 - Community process

Doable: hundreds of curators
globally in >200 databases
(GDP) - make it more efficient



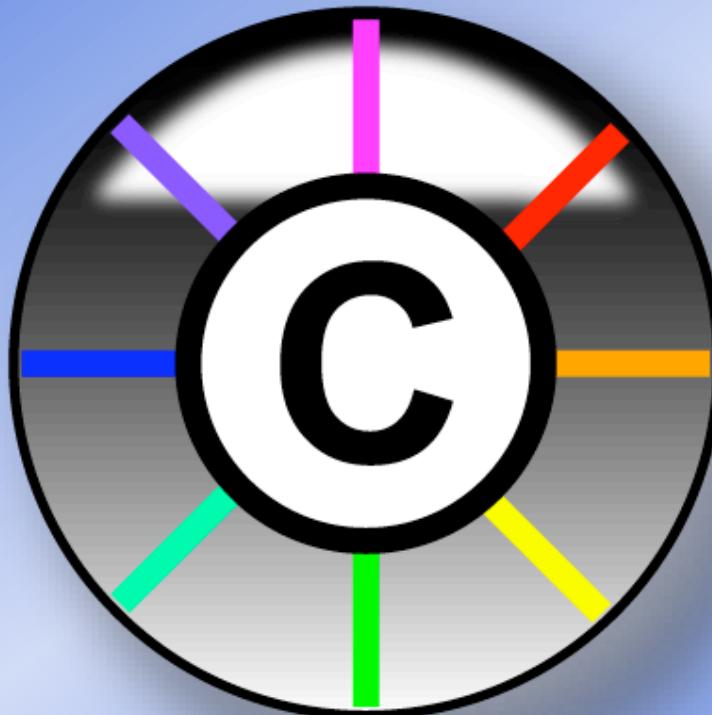
Network Visualization and Analysis

Using Pathway Information

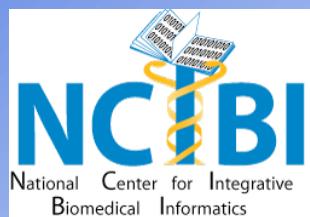




Cytoscape



UCSF



Agilent Technologies

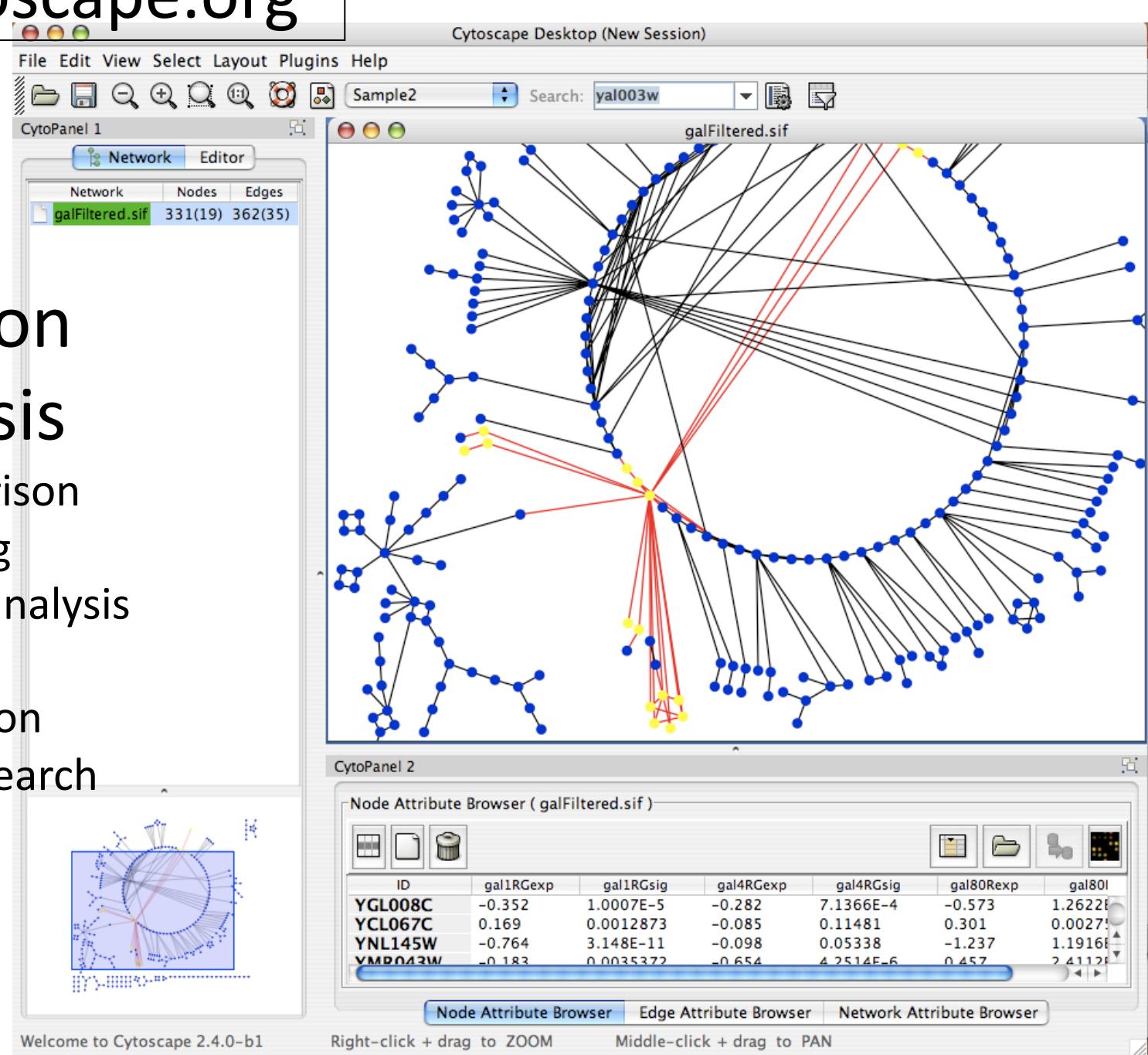


<http://cytoscape.org>

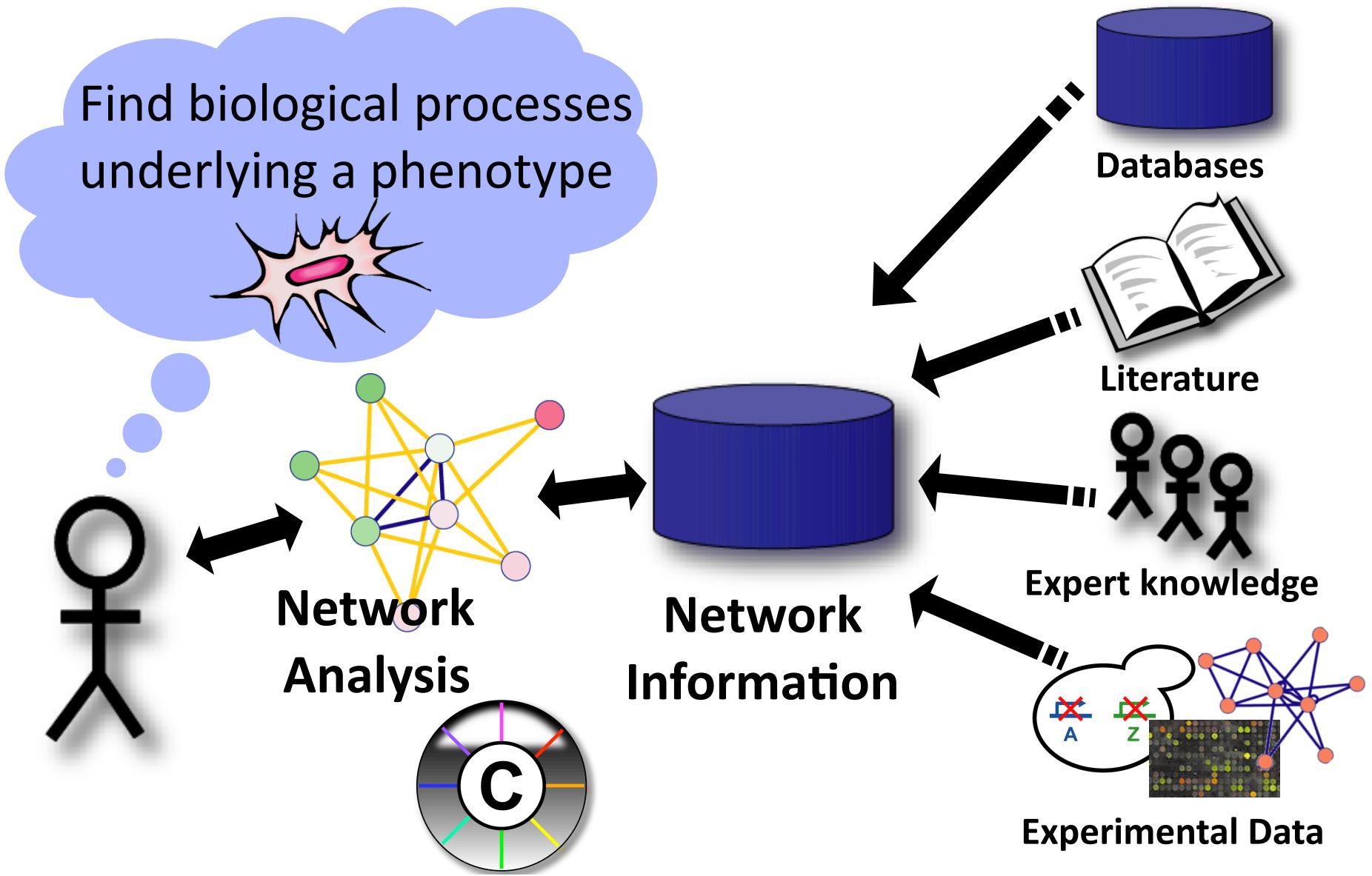
Network visualization and analysis

Pathway comparison
Literature mining
Gene Ontology analysis
Active modules
Complex detection
Network motif search

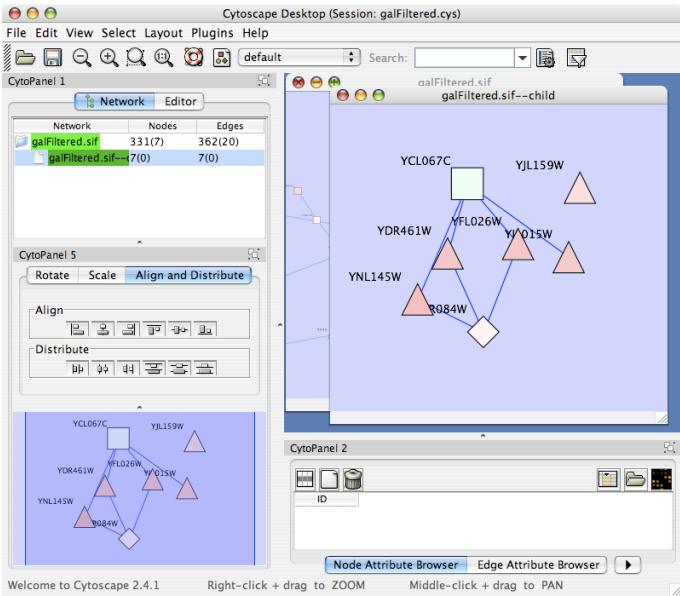
UCSD, ISB, Agilent,
MSKCC, Pasteur, UCSF,
Unilever, UToronto, U
Michigan



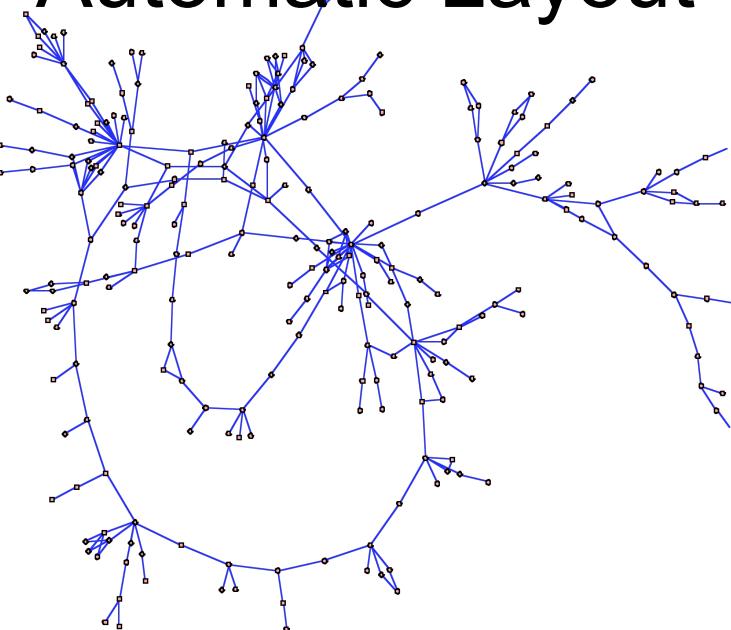
Network Analysis using Cytoscape



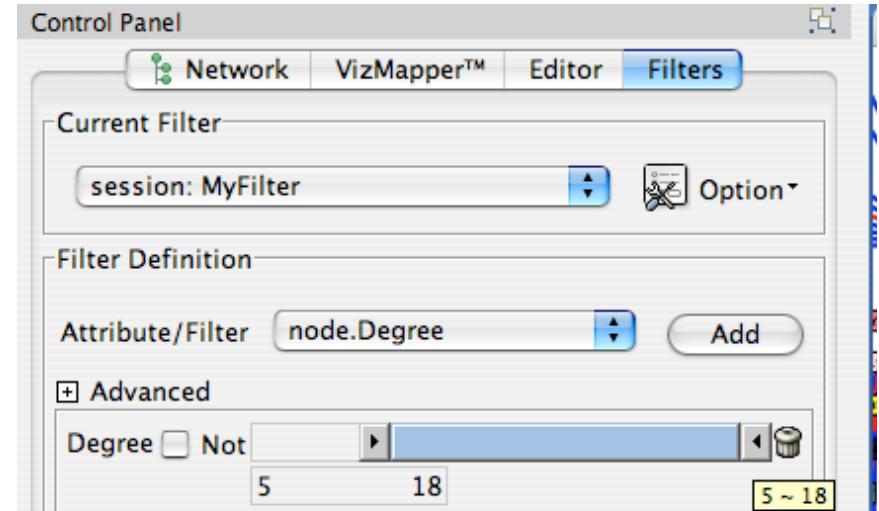
Manipulate Networks



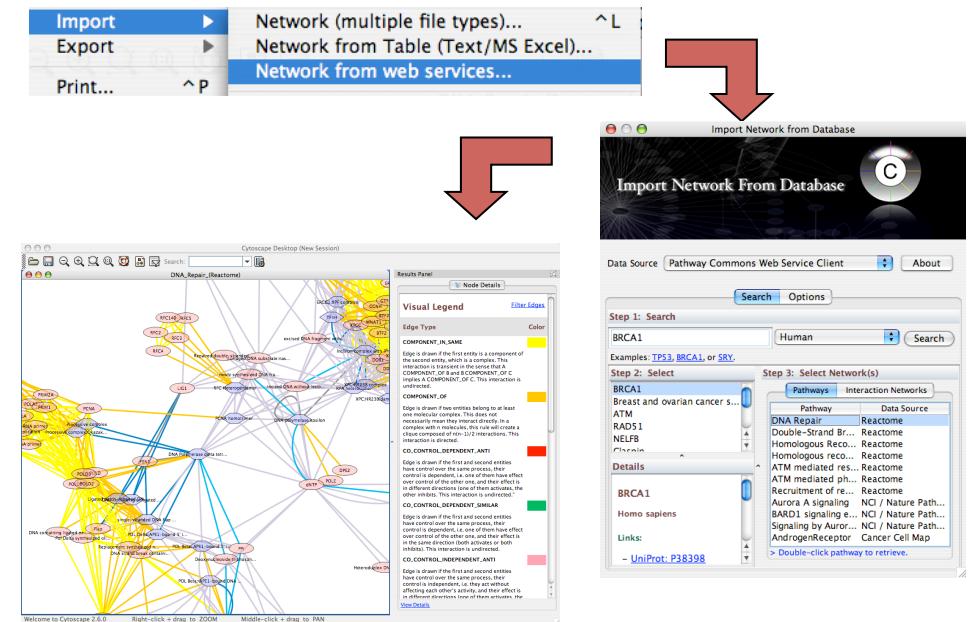
Automatic Layout



Filter/Query

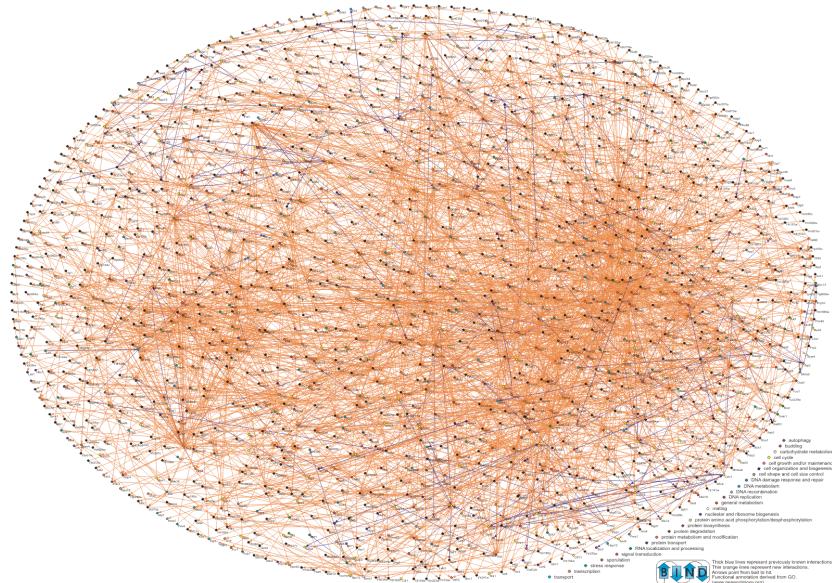


Interaction Database Search



Overview

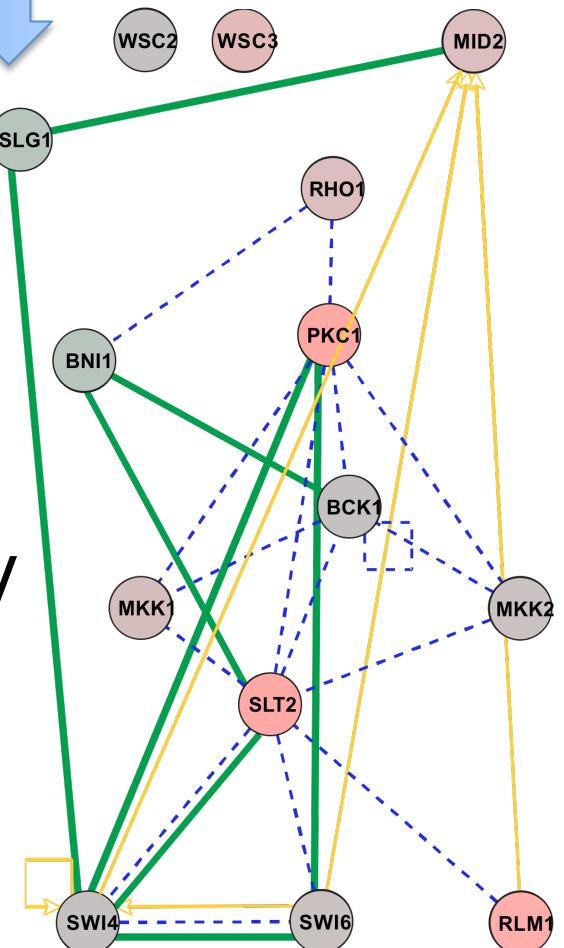
Systematic identification of protein complexes in *Saccharomyces cerevisiae* by mass spectrometry



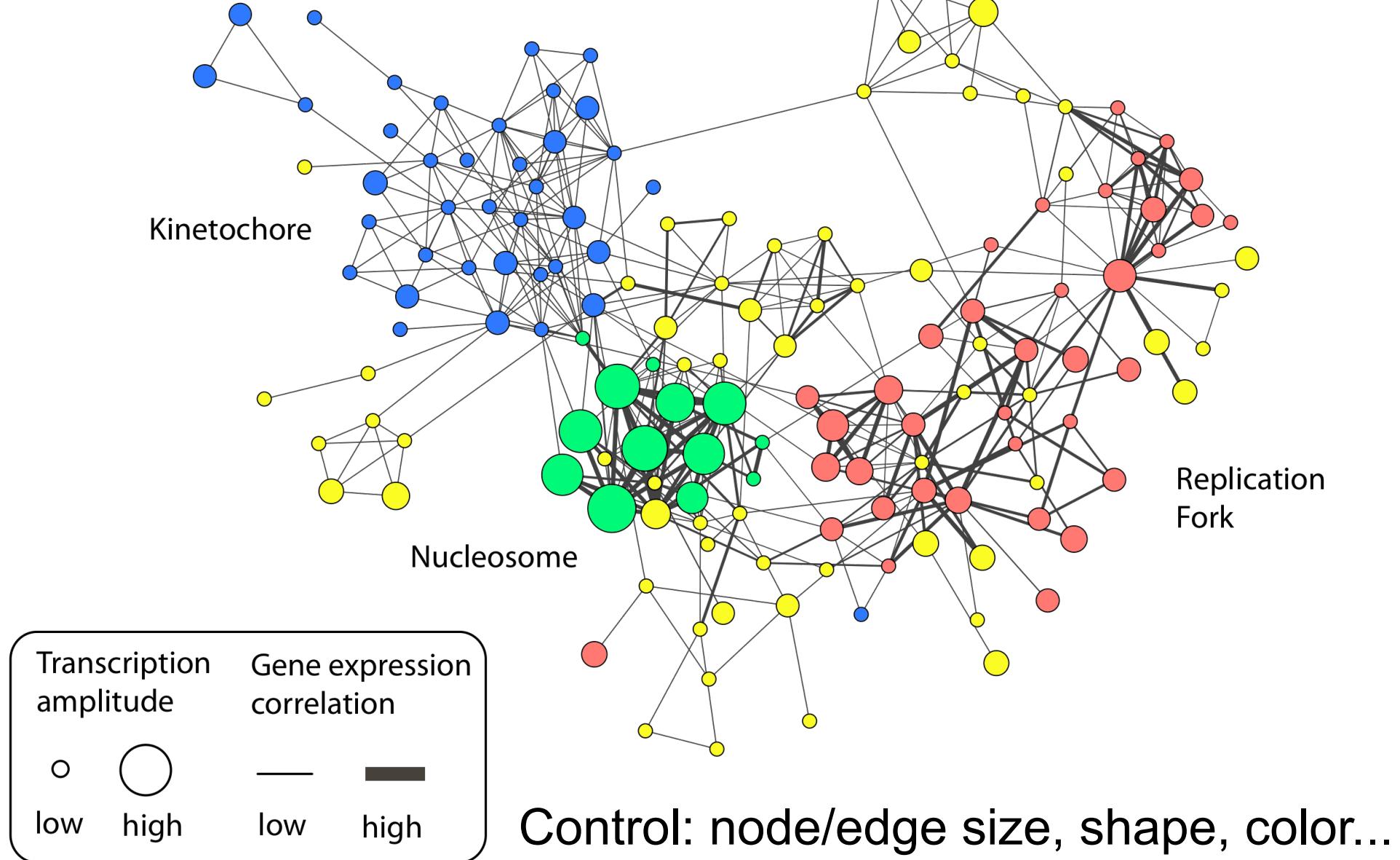
Zoom

Focus

PKC
Cell
Wall
Integrity



Visualize multiple data types on a network



Active Community

<http://www.cytoscape.org>

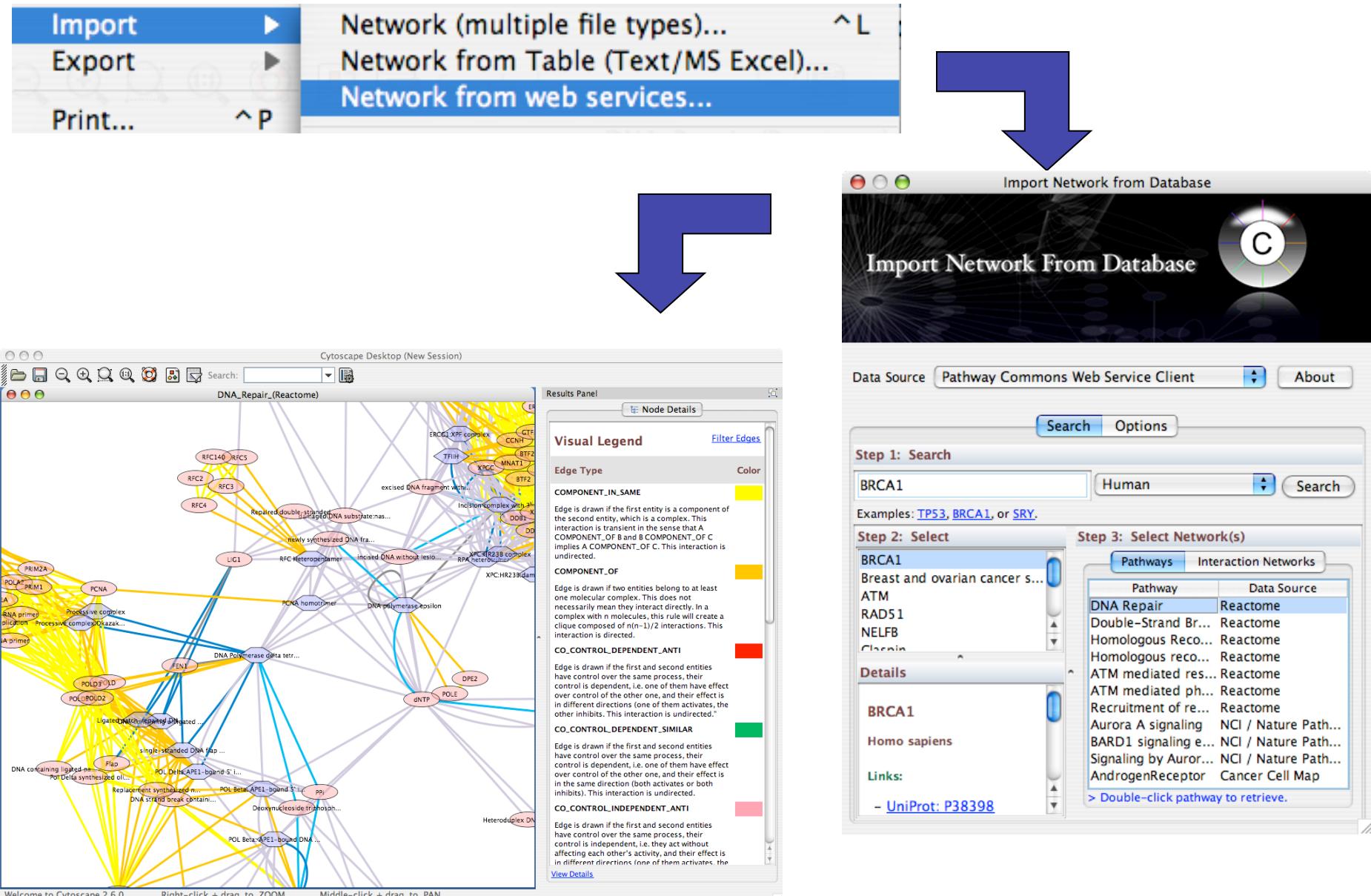
- Help
 - 8 tutorials, >10 case studies Cline MS et al. Integration of biological networks and gene expression data using Cytoscape Nat Protoc. 2007;2(10):2366-82
 - Mailing lists for discussion
 - Documentation, data sets
- 10,000s users, 2500 downloads/month
- >40 Plugins Extend Functionality
 - Build your own, requires programming
 - e.g. Retina Workbench

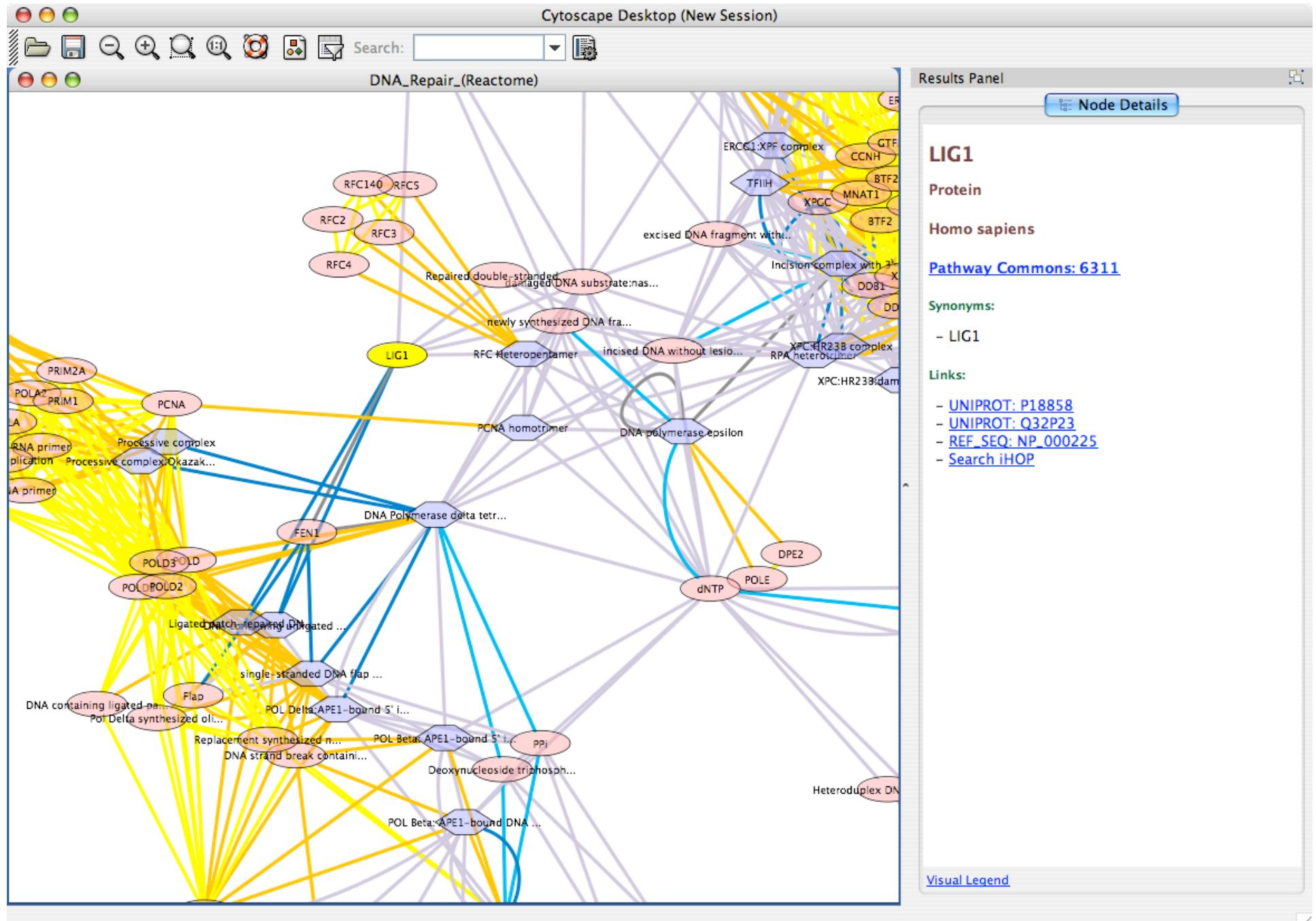
Analyzing Molecular Profiles

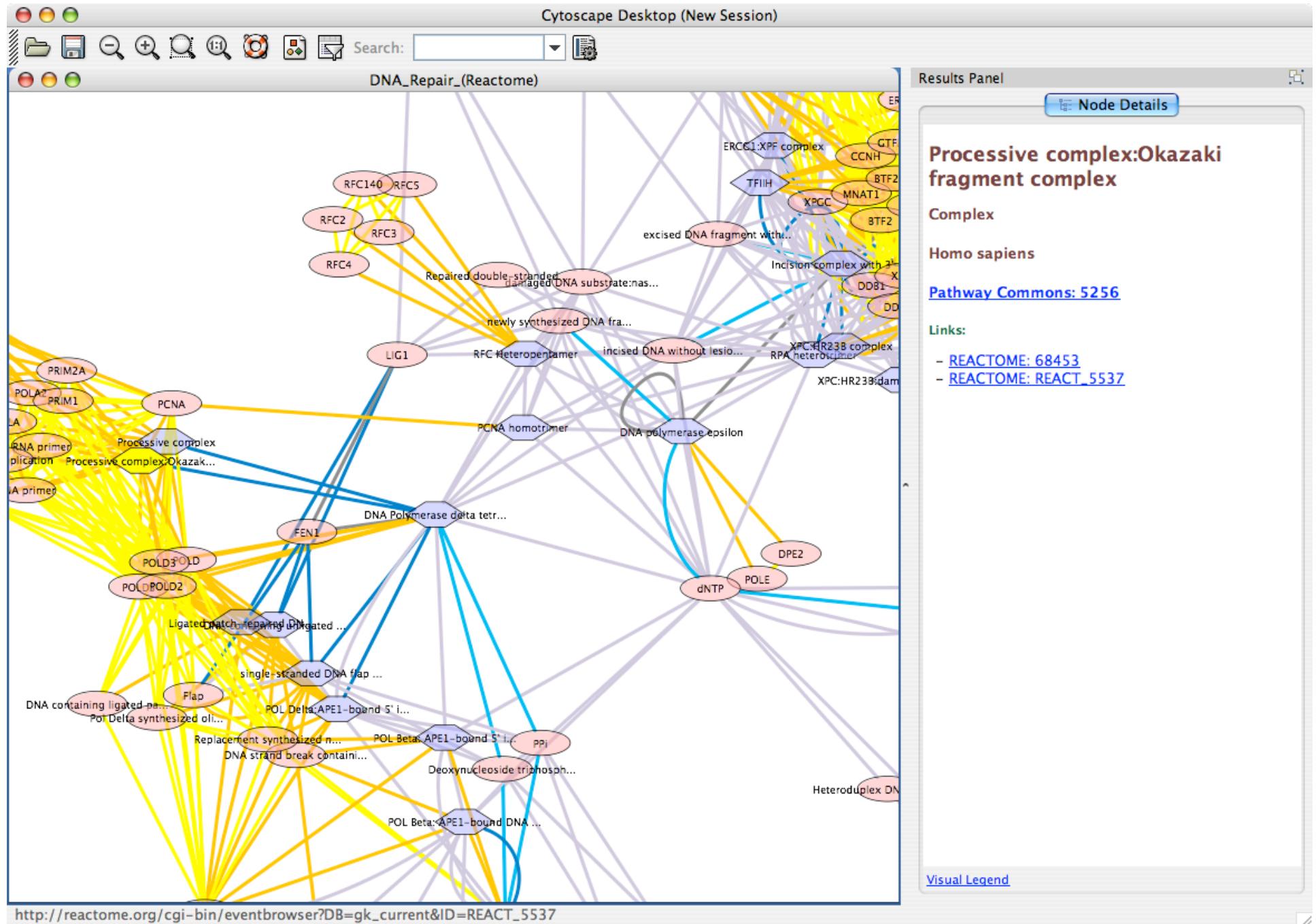
Analyzing gene expression data in a network context

- Input
 - Gene expression data
 - Network data
- Output
 - Visual diagram of expression data on network
 - Active network regions
- Outline
 - Where to find network data?
 - Interaction database (cPath)
 - Literature associations via text mining
 - Load expression data
 - Identify active pathways

Interaction Database Search

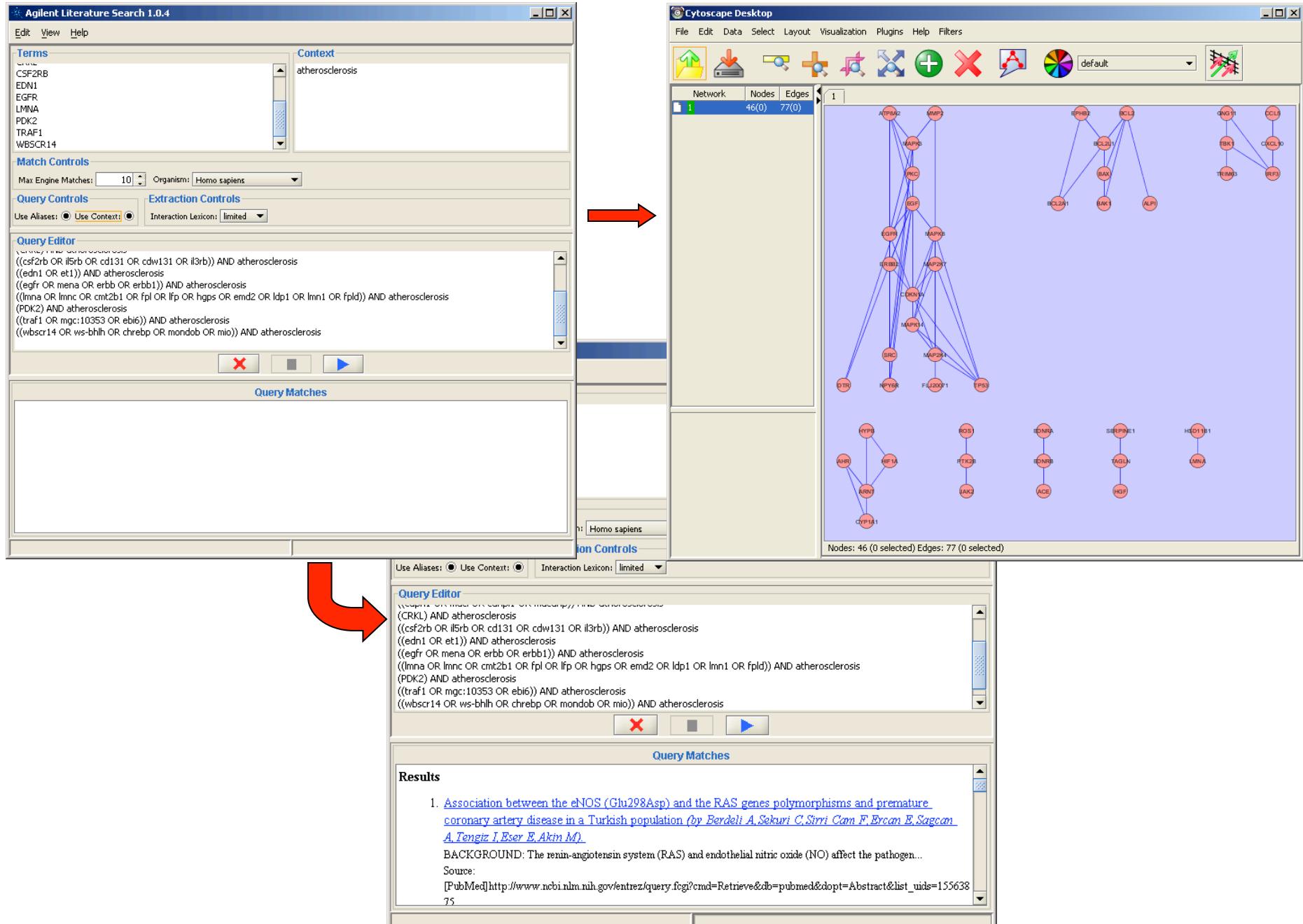






Text Mining

- Computationally extract gene relationships from text, usually PubMed abstracts
- Literature search tool, lots of network data
- BUT not perfect
 - Problems recognizing gene names
 - Natural language processing not perfect
- Agilent Literature Search Cytoscape plugin
- Others: E.g. iHOP
 - www.ihop-net.org/UniPub/iHOP/



Cytoscape Network produced by Literature Search.

Abstract from the scientific literature

Sentences for an edge

BCL2L1 -> BCL2A1 Agilent Literature Search Sentences

1. [High glucose inhibits apoptosis in human coronary artery smooth muscle cells by increasing bcl-xL and bfl-1/A1.](#)
 2. [High glucose induced phosphorylation of phosphatidylinositol 3-kinase \(PI 3-K\) and extracellular signal-regulated kinase \(ERK\)1/2 along with bcl-xL and bfl-1/A1 upregulation.](#)

Physiol. 2002 Aug;283(2):C422-8.
 at org
inhibits apoptosis in human coronary artery smooth muscle cells by increasing 1/A1.
 moto M, Okumura M, Kojima T, Maruyama T, Yasuda K.
 f Internal Medicine, Gifu University School of Medicine, Gifu 500-8705, Japan.

Clinical Queries
 LinkOut
 My NCBI (Cubby)
 Related Resources
 Order Documents
 NLM Catalog
 NLM Gateway
 TOXNET
 Consumer Health
 Clinical Alerts
 ClinicalTrials.gov
 PubMed Central

Cardiovascular disease is a serious complication in diabetic patients. To elucidate the precise mechanisms of atherosclerosis in diabetic patients, the effects of high glucose concentration (25 mM) on apoptosis regulation and bcl-2 family protein expression in human coronary artery smooth muscle cells (CASM) were examined. Treatment with a high level of glucose (25 mM) caused a significant decrease in apoptosis in CASM compared with the same cells treated with a physiologically normal glucose concentration (5.5 mM) (23.9 +/- 2.4% vs. 16.5 +/- 1.8%; P < 0.01). With respect to apoptosis regulation, treatment of CASM with high glucose concentration markedly increased mRNA expressions of bcl-xL and bfl-1/A1 compared with cells treated with normal glucose. High glucose induced phosphorylation of phosphatidylinositol 3-kinase (PI 3-K) and extracellular signal-regulated kinase (ERK)1/2 along with bcl-xL and bfl-1/A1 upregulation. These results suggest that high glucose suppresses apoptosis via upregulation of bcl-xL and bfl-1/A1 levels through PI 3-K and ERK1/2 pathways in CASM. High glucose-induced increase in the expression of antiapoptotic proteins may be important in the development of atherosclerosis in diabetic patients.

PMID: 12107051 [PubMed - indexed for MEDLINE]

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[Department of Health & Human Services](#)
[Privacy Statement](#) | [Freedom of Information Act](#) | [Disclaimer](#)

Mar 29 2005 17:30:14

Gene Expression/Network Integration

- Identifier (ID) mapping
 - Translation from network IDs to gene expression IDs e.g. Affymetrix probe IDs
 - Also: Unification, link out, query
 - Entrez gene IDs (genes), UniProt (proteins)
- Synergizer
 - llama.med.harvard.edu/cgi/synergizer/translate
- More ID mapping services available
 - <http://baderlab.org/IdentifierMapping>

Gene Expression/Network Integration

THE SYNERGIZER

The Synergizer database is a growing repository of gene and protein identifier synonym relationships. This tool facilitates the conversion of identifiers from one naming scheme (a.k.a "namespace") to another.

[load sample inputs](#)

Select species:

Select authority:

Select "FROM" namespace:

Select "TO" namespace: [854192]

(NB: The strings in [brackets] are representative IDs in the corresponding namespaces.)

File containing IDs to translate: [Browse...](#)

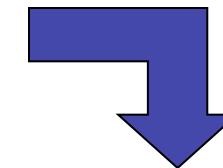
and/or

IDs to translate:

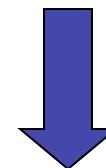
```
YIL062C  
YLR370C  
YKL013C  
YNR035C  
YBR234C
```

Output as spreadsheet:

[Submit](#)

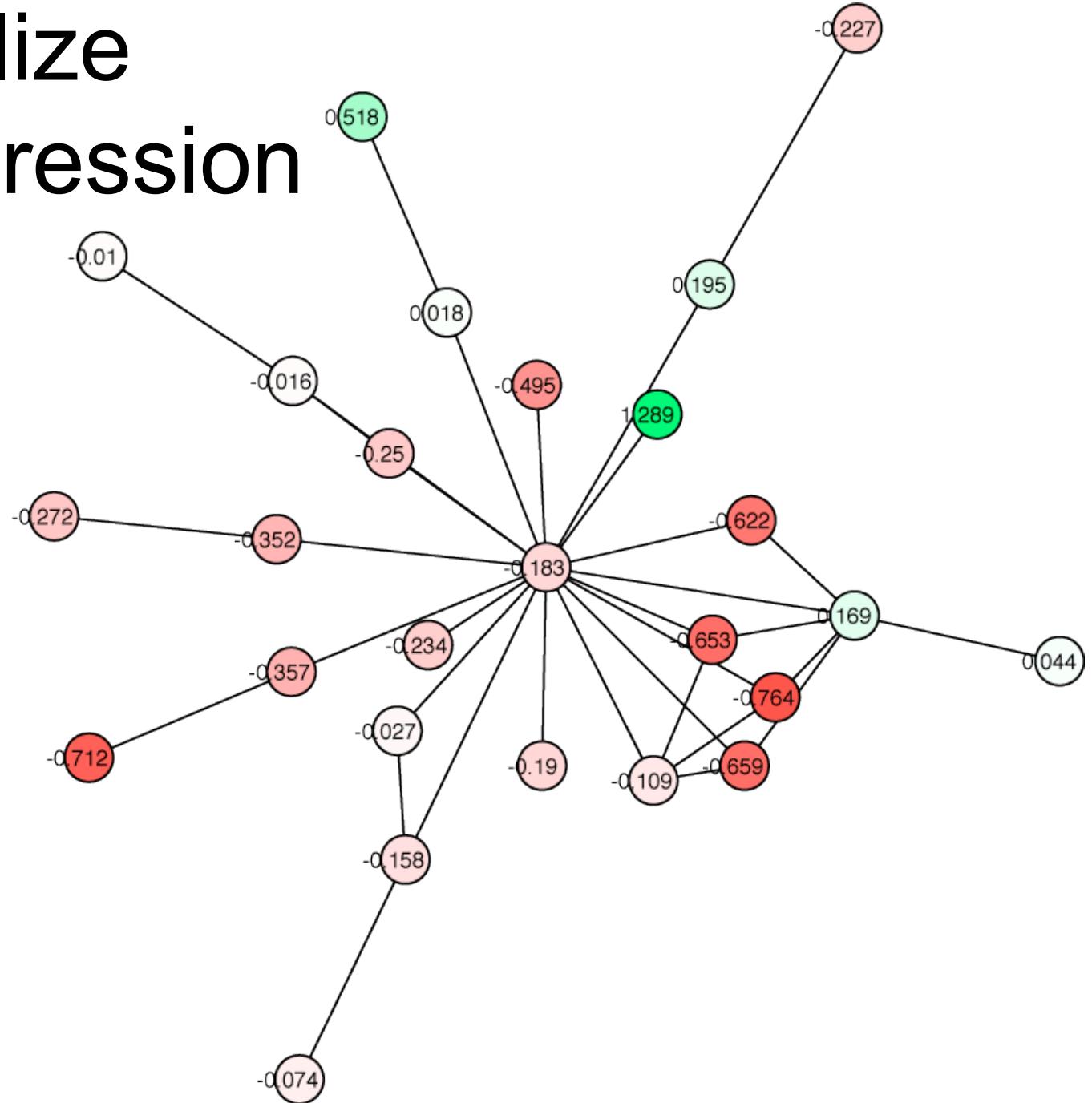


*	entrezgene
YIL062C	854748
YLR370C	851085
YKL013C	853856
YNR035C	855771
YBR234C	852536



1. Load as attributes in Cytoscape
2. Assign expression values to nodes using this attribute set

Visualize Gene Expression



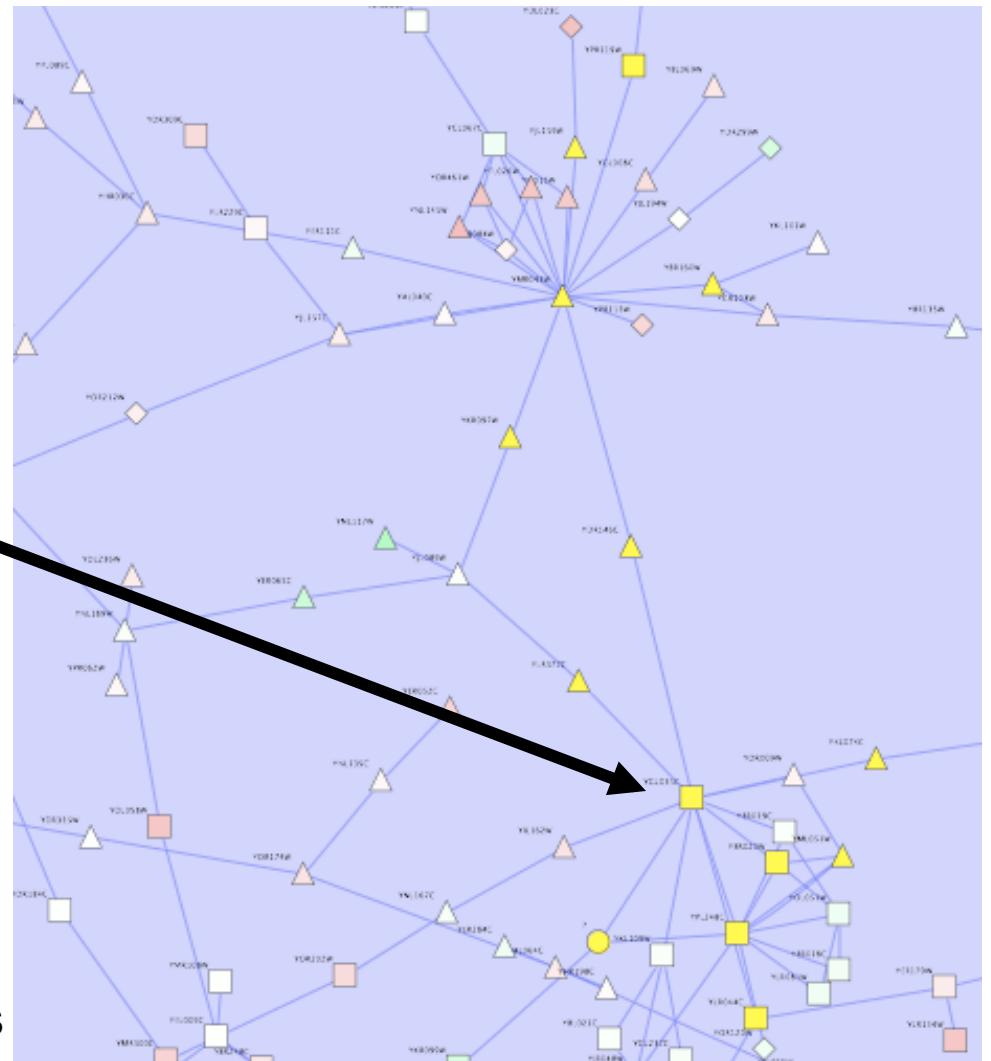
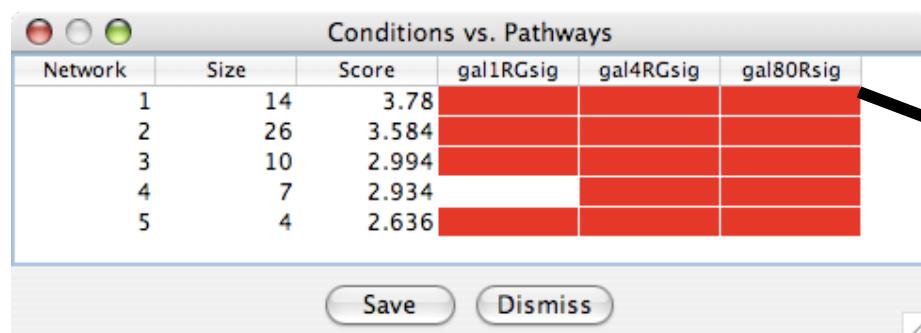
Find Active Subnetworks

- Active modules
 - Input: network + p-values for gene expression values e.g. from GCRMA
 - Output: significantly differentially expressed subgraphs
- Method
 - Calculate z-score/node, Z_A score/subgraph, correct vs. random expression data sampling
 - Score over multiple experimental conditions
 - Simulated annealing used to find high scoring networks

Active Module Results

Network: yeast protein-protein and protein-DNA network

Expression data: 3 gene knock out conditions (enzyme, TF activator, TF repressor)



Note: non-deterministic, multiple runs required for confidence of result robustness

Ideker T et al. Science. 2001 May 4;292(5518):929-34.

Bonus Slides

Gene and Protein Identifiers

- Identifiers (IDs) are names or numbers that help track database records
 - E.g. Social Insurance Number, Entrez Gene ID 41232
- Gene and protein information stored in many databases
 - → Genes have many IDs
- Records for: Gene, DNA, RNA, Protein
 - Important to use the correct record type
 - E.g. Entrez Gene records don't store sequence. They link to DNA regions, RNA transcripts and proteins.

Common Identifiers

Gene

Ensembl [ENSG00000139618](#)

Entrez Gene [675](#)

Unigene [Hs.34012](#)

RNA transcript

GenBank [BC026160.1](#)

RefSeq [NM_000059](#)

Ensembl [ENST00000380152](#)

Protein

Ensembl [ENSP00000369497](#)

RefSeq [NP_000050.2](#)

UniProt [BRCA2_HUMAN](#) or

[A1YBP1_HUMAN](#)

IPI [IPI00412408.1](#)

EMBL [AF309413](#)

PDB [1MIU](#)

Species-specific

HUGO HGNC [BRCA2](#)

MGI [MGI:109337](#)

RGD [2219](#)

ZFIN [ZDB-GENE-060510-3](#)

FlyBase [CG9097](#)

WormBase [WBGene00002299](#) or [ZK1067.1](#)

SGD [S000002187](#) or [YDL029W](#)

Annotations

InterPro [IPR015252](#)

OMIM [600185](#)

Pfam [PF09104](#)

Gene Ontology [GO:0000724](#)

SNPs [rs28897757](#)

Experimental Platform

Affymetrix [208368_3p_s_at](#)

Agilent [A_23_P99452](#)

CodeLink [GE60169](#)

Illumina [GI_4502450-S](#)

Red = Recommended

ID Mapping Services

THE SYNERGIZER

The Synergizer database is a growing repository of gene and protein identifier synonym relationships. This tool facilitates the conversion of identifiers from one naming scheme (a.k.a "namespace") to another.

[load sample inputs](#)

Select species:

Select authority:

Select "FROM" namespace:

Select "TO" namespace: [854192]

(NB: The strings in [brackets] are representative IDs in the corresponding namespaces.)

File containing IDs to translate: [Browse...](#)

and/or

IDs to translate:

```
YIL062C  
YLR370C  
YKL013C  
YNR035C  
YBR234C
```

Output as spreadsheet:

[Submit](#)



*	entrezgene
YIL062C	854748
YLR370C	851085
YKL013C	853856
YNR035C	855771
YBR234C	852536

- **Synergizer**
 - <http://llama.med.harvard.edu/cgi/synergizer/translate>
- **Ensembl BioMart**
 - <http://www.ensembl.org>
- **PIR**
 - <http://pir.georgetown.edu/pirwww/search/idmapping.shtml>

ID Mapping Challenges

- Gene name ambiguity
 - Not a good ID, but official gene symbol is ok e.g. HGNC/HUGO gene symbol
- Excel error-introduction
 - OCT4 is changed to October-4
- Problems reaching 100% coverage
 - E.g. due to version issues
 - Use multiple sources to increase coverage

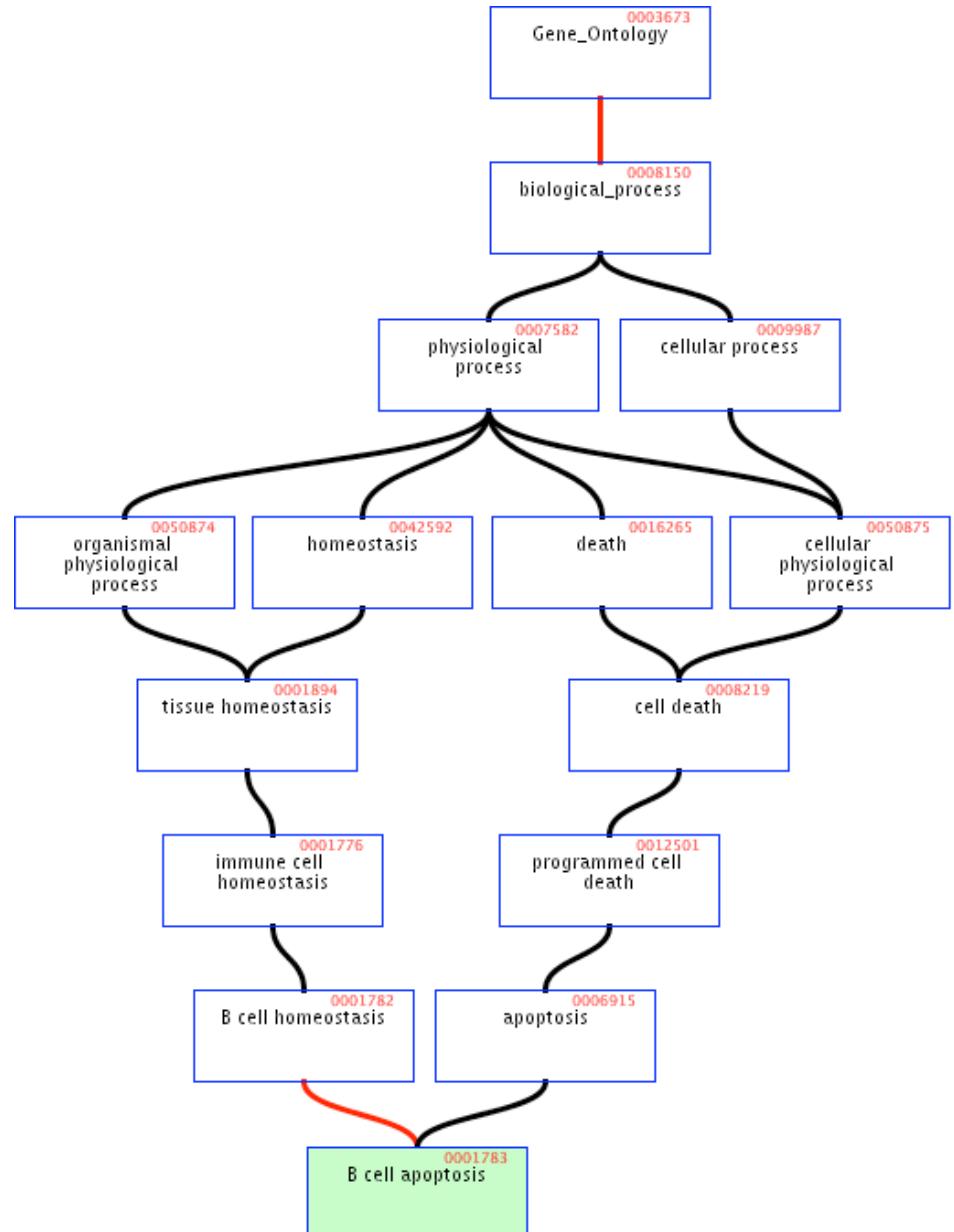
Zeeberg BR et al. Mistaken identifiers: gene name errors can be introduced inadvertently when using Excel in bioinformatics BMC Bioinformatics. 2004 Jun 23;5:80

Additional Plugins

- Bingo: over-representation analysis
- ClusterMaker: clusters networks, includes MCL
- NetworkAnalyzer: calculates statistics about a network
- (You may have to use an earlier version of Cytoscape to get some plugins to run)

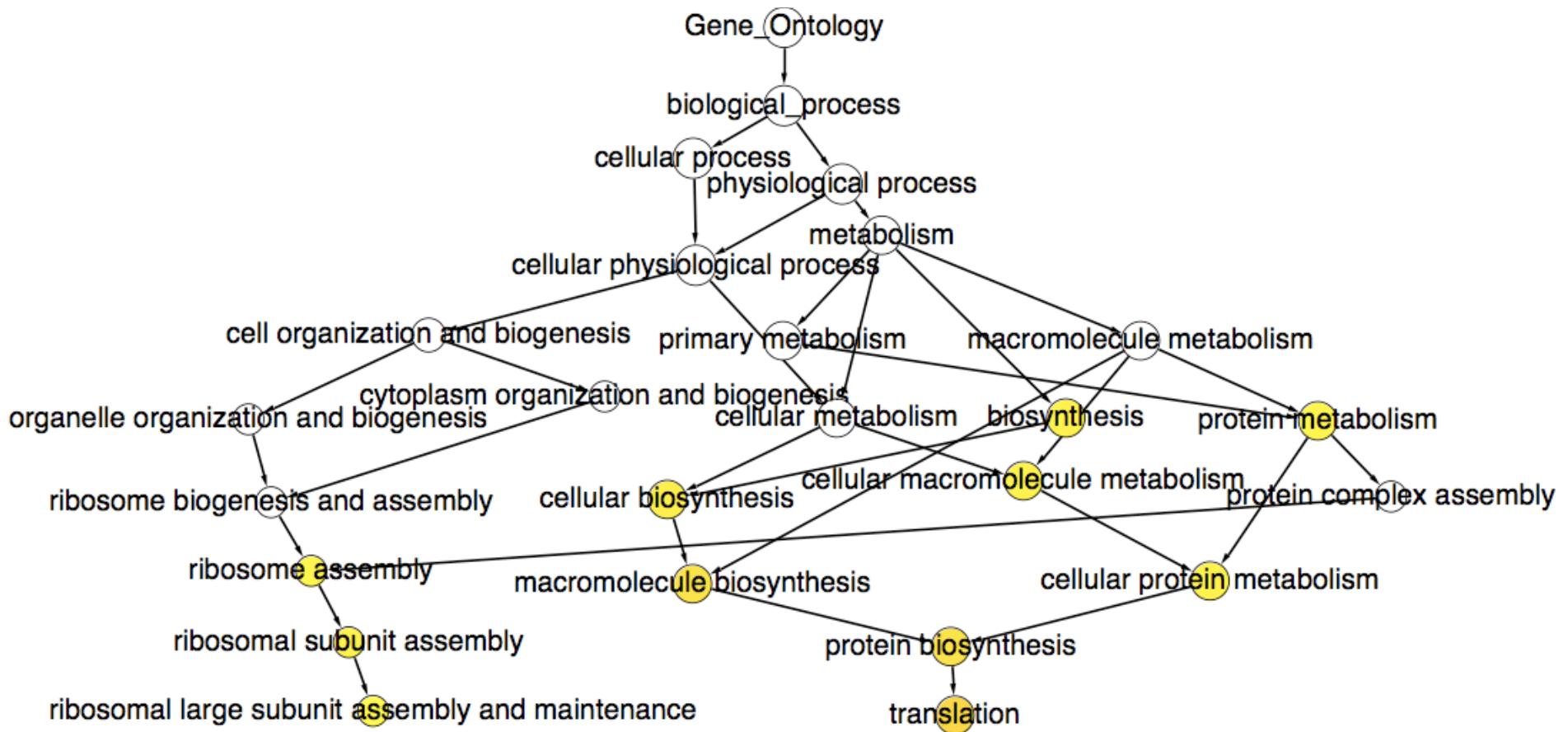
The Gene Ontology (GO)

- Describes gene function
1. Agreed upon terms (controlled vocabulary)
 - Biological process
 - Cellular component
 - Molecular function
 2. Genome annotation



BiNGO

Hypergeometric p-value
Multiple testing correction
(Benjamini-Hochberg FDR)



Caveats: Gene identifiers must match;
low GO term coverage, GO bias

Maere, S., Heymans, K. and Kuiper, M
Bioinformatics 21, 3448-3449, 2005

NetMatch

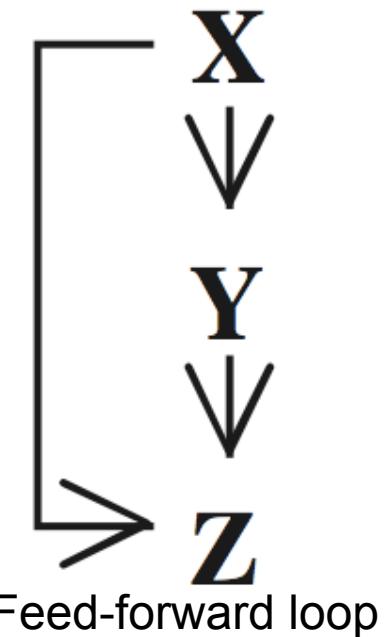
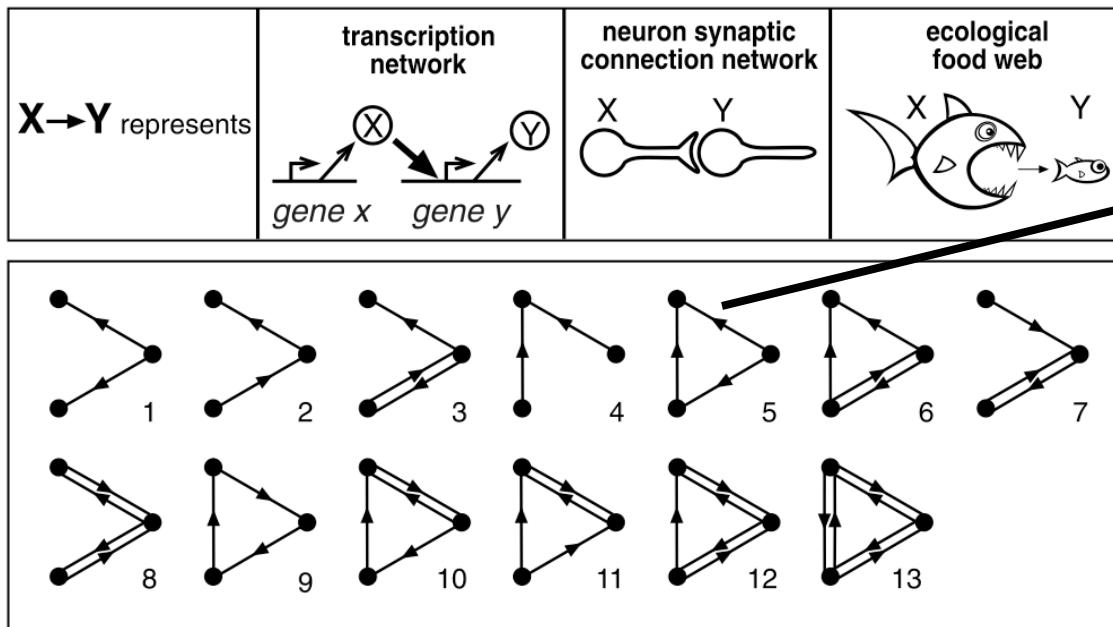
- Query a network for topological matches
- Input: query and target networks, optional node/edge labels
- Output: Topological query matches as subgraphs of target network
- Supports: subgraph matching, node/edge labels, label wildcards, approximate paths
- <http://alpha.dmi.unict.it/~ctnyu/netmatch.html>

Ferro A, Giugno R, Pigola G, Pulvirenti A, Skripin D, Bader GD, Shasha D
Bioinformatics 2007 Feb 3

Extends state space representation based search from Cordella et al. IEEE
Transactions on Pattern Analysis and Machine Intelligence, 2004, 26, 10, 1367--1372

Find Feed-Forward Motifs

- Graph motifs over-represented in many network types



Gene regulation
Neurons
Electronic circuits

Milo et al. Science 298, 2002

Find Feed-Forward Motifs

NetMatch Query Editor – new query*

Query Edit

Palette Motifs

Feed Forward Loop

Info:

Pass Query to NetMatch

Nodes: 6 Edges: 6 Paths: 0 Loops: 0

Query

The screenshot shows the NetMatch Query Editor interface. In the central workspace, there is a diagram of a feed-forward loop motif, which consists of three nodes arranged in a triangle with directed edges between them. The nodes are represented by circles with question marks. On the left, the 'Motifs' tab of the palette is selected, showing various motifs including a feed-forward loop. Below the workspace, a large 'Query' button is centered. At the bottom, status information is displayed: 'Nodes: 6 Edges: 6 Paths: 0 Loops: 0'.

NetMatch V1.0.1

File Query Wizard Help

Graph Properties:

Labeled
 Directed

Query Properties:

Query: Draw a query...
QE-FFL

Query Node Attributes: QE-FFL - Nodes Attributes

Query Edge Attributes: QE-FFL - Edges Attributes

Network Properties:

Network: 1-galFiltered.sif

Network Node Attributes: annotation.GO BIOLOGIC...

Network Edge Attributes: TextSourceInfo

Options:

Acquire Data
Go
Reset

Match Number Nodes Image

1	YMR309C, YOR361C, YPR041W	
2	YOR310C, YDL014W, YLR197W	
3	YDR100W, YGL161C, YOR036W	
4	YIL015W, YMR043W, YCL067C	

Create a new child network.

1 matches YBR020W
2 matches YGL035C
***** Match 21
0 matches YPL248C
1 matches
2 matches

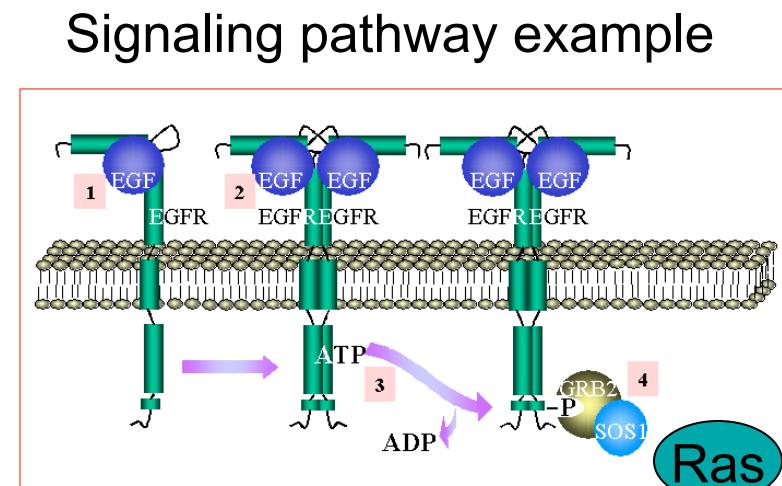
0 matches
1 matches YDR103W
2 matches YLR362W

Results

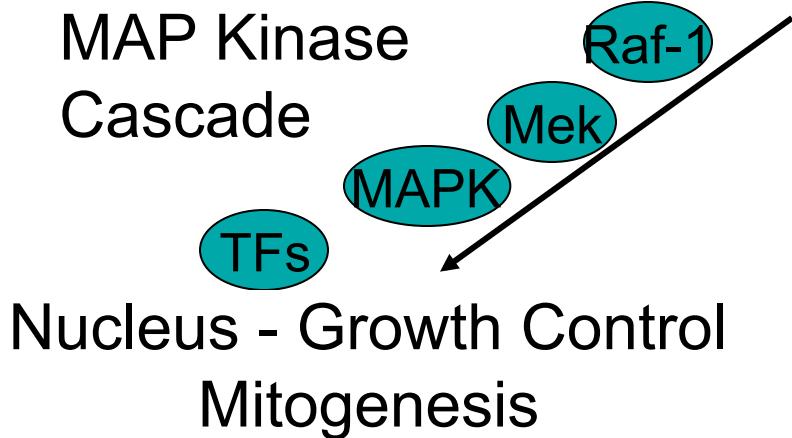
The screenshot shows the NetMatch V1.0.1 results window. On the left, a list of matches is shown with their respective node lists and small diagrams. On the right, a 'Results' box contains a summary of the matches found, including the number of matches for each node set and the total number of matches found. The 'Results' box also contains a '***** Match 21' entry, likely indicating the current match being viewed.

Find Signaling Pathways

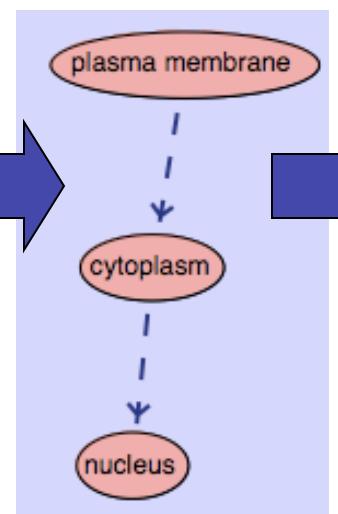
- Potential signaling pathways from plasma membrane to nucleus via cytoplasm



MAP Kinase
Cascade



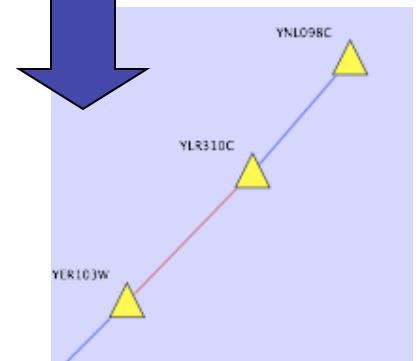
NetMatch query



NetMatch Results

Match Number	Nodes	Image
YGL008C		
4	YJL157C, YMR043W, YLR229C	
5	YJL157C, YAL040C, YLR229C	

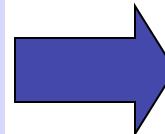
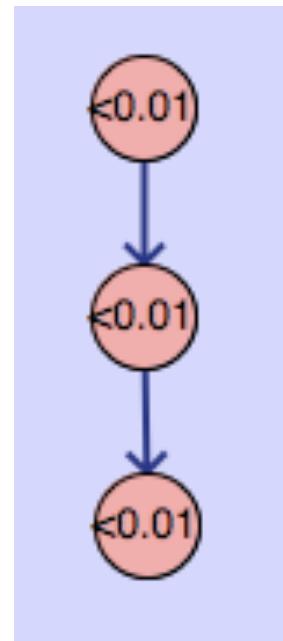
Shortest path between
subgraph matches



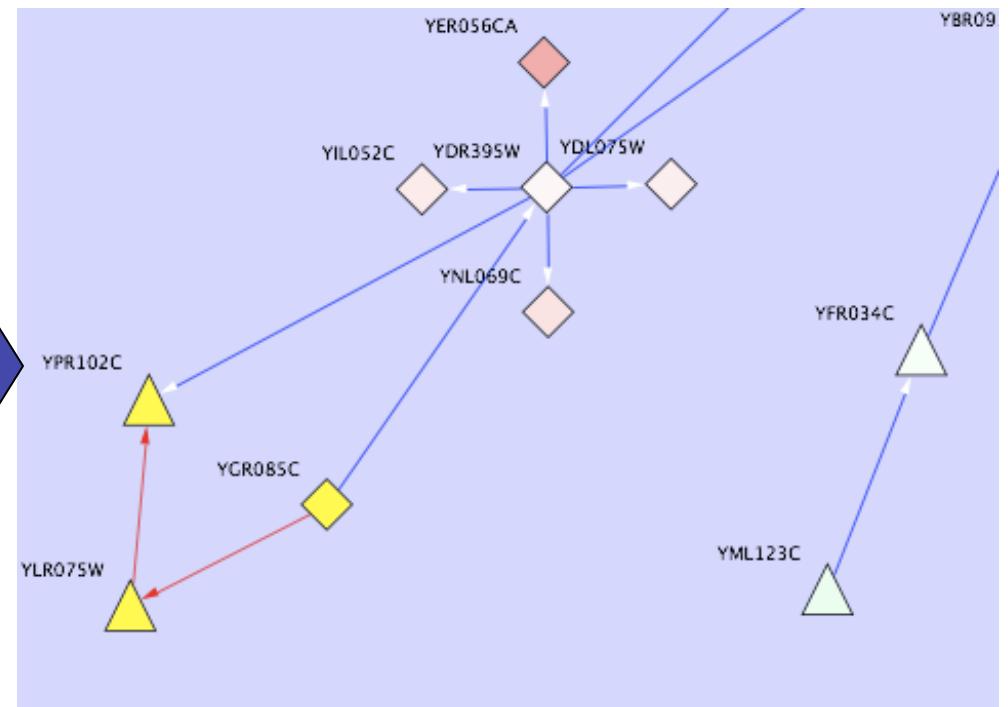
Find Expressed Motifs

Find specific subgraphs where certain nodes are significantly differentially expressed

NetMatch query



NetMatch Results



Protein
YLR075W
YGR085C
YPR102C

Differential Expression Significance
1.7255E-4
2.639E-4
3.7183E-4

Systems Biology Graphical Notation

<http://sbgn.org>

