

Network Biology: Mapping pathways to understand and diagnose disease

Many kinds of interaction technologies

	PHYSICAL	GENETIC			
ORDERED	Protein-gene (transcriptional, ChIP-Chip ^{22, 39}) Protein-RNA (RIP-chip ⁸⁰)	Epistatic orderings a < b OR b < a (EMAP ^{28, 83})			
Cause and effect Signal transducing	Protein-protein (kinase-substrate arrays ²¹ , <i>LUMIER</i> ⁸¹) Protein-compound ⁸²	Knock-down expression profiles (RNAi ³² , deletion mutants ^{36, 37}) Expression QTLs ^{41, 42}			
UNORDERED	Protein-protein (co-IP/MS/MS ¹⁸⁻²⁰ , Y2H ^{15, 84-86})	Synthetic lethality ab << a, b, wt			
Ambiguous directionality	Gene-gene (co-regulon ⁸⁷)	(SGA ⁸⁸ , dSLAM ^{31, 71} , EMAP ^{28, 83} , chemogenomic profiling ⁸⁹)			

Beyer, Bandyopadhyay, and Ideker Nat. Rev. Genetics (2007)

Two overriding aims: g...

1) Assemble many interactions and types into unified models 2) Get rid of false and nonfunctional interactions

These aims lead to many subproblems

- ✓ Mapping transcriptional networks
- ✓ Networks to interpret genetic variations
- ✓ Networks to interpret combinatorial perturbations (e.g. synthetic lethals)
- ✓ Network evolution
- ✓ Network-based diagnosis

Assembly of physical and genetic interactions to map transcriptional circuits

> (Chris Workman, Craig Mak with Leona Samson, Richard Kolodner)

Transcriptional response of *Saccharomyces cerevisiae* to DNA-damaging agents does not identify the genes that protect against these agents

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Contributed by Ronald W. Davis, May 8, 2002

The recent genes in bi determining to cytotoxic esis that gei are importa 4,627 diploi sential gen survival of y UV radiation addition we

Can this apparent paradox be explained by a physical model of the DNA damage response? informafact, sugdamaging hence in (17–19). increases at protect formally yeast, *S.* rectly test

type parentar strain to the same DNA-damaging agents. We round no relationship between the genes necessary for survival to the DNA-damaging agents and those genes whose transcription is increased after exposure. These data show that few, if any, of the genes involved in repairing the DNA lesions produced in this study, including double-strand breaks, pyrimidine dimers, single-strand breaks, base damage, and DNA cross-links, are induced in response to toxic doses of the agents that produce these lesions. This finding suggests that the enzymes necessary for the repair of these lesions are at sufficient levels within the cell. The data also suggest that the nature of the lesions produced by DNA-damaging agents cannot easily be deduced from gene expression profiling. Deletion of the genes has been accomplished by an international consortium, the *Saccharomyces* Genome Deletion Project, that has replaced all of the \approx 6,200 known open reading frames (ORFs) of yeast by using a PCR-mediated gene deletion strategy (20). In addition to a selectable marker, two molecular bar codes or "'tags," unique 20-base oligonucleotide sequences, are in the replacement cassette. These tags, after PCR amplification, can be detected by hybridization to the corresponding complementary sequence in a high-density oligonucleotide array, thus enabling the relative abundances of each tag, and hence the abundances of each deletion strain, to be determined (20). We have recently shown that this system can detect essentially all of



From Figure 1 of Simon et al. Cell 2001

Mapping DNA Damage Response Networks

Numbers of promoters bound by each of 30 transcription factors (TFs) before and after exposure to methyl-methane sulfonate (MMS)



Workman, Mak, et al. Science 2006



Integration of cause-and-effect interactions with physical networks



Yeang, Mak et al. Genome Biology 2005



Such methods can yield large regulatory networks

Transcriptional response of *Saccharomyces cerevisiae* to DNA-damaging agents does not identify the genes that protect against these agents

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The recent completion of the deletion of all of the nonessential genes in budding yeast has provided a powerful new way of determining those genes that affect the sensitivity of this organism to cytotoxic agents. We have used this system to test the hypothesis that genes whose transcription is increased after DNA damage are important for the survival to that damage. We used a pool of 4,627 diploid strains each with homozygous deletion of a nonessential gene to identify those genes that are important for the survival of yeast to four DNA-damaging agents: ionizing radiation, UV radiation, and exposure to cisplatin or to hydrogen peroxide. In addition we measured the transcriptional response of the wildtype parental strain to the same DNA-damaging agents. We found no relationship between the genes necessary for survival to the DNA-damaging agents and those genes whose transcription is increased after exposure. These data show that few, if any, of the genes involved in repairing the DNA lesions produced in this study, including double-strand breaks, pyrimidine dimers, single-strand breaks, base damage, and DNA cross-links, are induced in response to toxic doses of the agents that produce these lesions. This finding suggests that the enzymes necessary for the repair of these lesions are at sufficient levels within the cell. The data also suggest that the nature of the lesions produced by DNA-damaging agents cannot easily be deduced from gene expression profiling.

conferring resistance to that agent, and hence provide information on its mechanism. Recent publications have, in fact, suggested that several of the genes induced by DNA-damaging agents are involved in the repair of DNA damage and hence in the protection of the cell against such treatments (17–19). However, the assumption that genes whose expression increases in response to a particular cytotoxic agent are those that protect against the damage caused by the agent has not been formally tested. Here we use a pool of strains of budding yeast, *S. cerevisiae*, with deletion of all nonessential genes to directly test this hypothesis.

Deletion of the genes has been accomplished by an international consortium, the *Saccharomyces* Genome Deletion Project, that has replaced all of the \approx 6,200 known open reading frames (ORFs) of yeast by using a PCR-mediated gene deletion strategy (20). In addition to a selectable marker, two molecular bar codes or "'tags," unique 20-base oligonucleotide sequences, are in the replacement cassette. These tags, after PCR amplification, can be detected by hybridization to the corresponding complementary sequence in a high-density oligonucleotide array, thus enabling the relative abundances of each tag, and hence the abundances of each deletion strain, to be determined (20). We have recently shown that this system can detect essentially all of Sensitivity of the TF knockout phenotype correlates with its number of regulated targets



Mapping Pathways in Synthetic Lethal Networks

(Ryan Kelley, Sourav Bandyopadhyay with Nevan Krogan)

Finding physical pathways to explain genetic interactions

Genetic Interactions:

- Classical method used to map pathways in model species
- Highly analogous to multi-genic interaction in human disease and combination therapy
- Thousands are being uncovered through systematic studies
- Thus as with other types, the number of known genetic interactions is *exponentially increasing...*



Adapted from Tong et al., Science 2001

Integration of genetic and physical interactions



Kelley and Ideker Nature Biotechnology (2005)



Global organization of genetic linkages between physical modules (A-Z)



Towards a generative module map

• Use generative model of cell which considers *k* modules simultaneously along with their inter-module functional relationships.

• Consider both positive and negative quantitative genetic interactions (alleviating and aggravating)



Functional maps of protein complexes



Using protein networks to understand molecular evolution

(with Roded Sharan, Richard Karp, and others)

Cross-comparison of networks:

(1) Conserved regions in the presence vs. absence of stimulus(2) Conserved regions across different species



Kelley et al. *PNAS* 2003 Ideker & Sharan *Gen Res* 2008 Suthram et al. *Nature* 2005 Sharan & Ideker *Nat. Biotech.* 2006 Sharan et al. RECOMB 2004 Scott et al. RECOMB 2005

Plasmodium: a network apart?

[a] Endocytosis



[b] Unfolded protein response



Conserved *Plasmodium / Saccharomyces* protein complexes



Plasmodium-specific protein complexes

Suthram et al. *Nature* 2005 La Count et al. *Nature* 2005 Pairwise alignment of the *E. coli* protein network versus the indicated species; Sensitivity comparison of different methods



Flannick et al. Genome Research (2006) [Batzoglou Lab]

1960		1970	19	80				199	90
BIOLOGICAL SEQUENCE COMPARISON									
First protein sequences by Sanger, others	Dayhoff, Jukes/ Cantor	Needleman/ Wunsch	PAM, LOSUM ^{matrix} Sn Wate	Swis Gen hith/ erman	¦ s-Prot, Bank, L-Bank ¦ Doo	Stor	mo Tay Lipr othe	Hauss Borodov Churc Ior, nan,	ler, vsky, hill BLAST
A new type of data becomes routinely Mathematical models of		Scoring via transition probabilities		l Pu genom data	Public Mini nome-scale moti databases don		ng for Hidden is and Markov nains Models		
available	evolution	Automated pairwise alignment	Fast dy progra alignm	/namic mming ent	Analy glo prope inforn con	sis of bal erties; nation tent	Mult aligni	iple ment	Database queries are staple of molecular biology
Interaction detection with 2-hybrid, mass. spec.	Interologs; evolutionary models	Ogata/ Kanehisa	NaWish Path	BIND, MINT, BLAST	DIP, GRID prop	Ale -free me erty; tness	on's work otifs Shar Karp/lo	???? an/ deker	????
BIOLOGICAL NETWORK COMPARISON									
1990	2001	2002	2003	2004	2005	5		2010)?
	Sharan and Ideker Nat. Biotech (2006								

Using protein networks for disease classification

(Han Yu Chuang)



Such methods can yield large regulatory networks

Using protein networks to diagnose breast cancer metastasis





Network markers are more reproducible and increase classification accuracy of breast cancer metastasis



www.cytoscape.org

OPEN SOURCE Java platform for integration of systems biology data

•Layout and query of interaction networks (physical and genetic)

•Visual and programmatic integration of molecular state data (attributes)

•The ultimate goal is to provide the tools to facilitate all aspects of pathway assembly and annotation.

RECENT NEWS

•Version 2.5 released Summer 2007; Scalability+efficiency now equivalent to best commercial packages

•The Cytoscape Consortium is a 501(c)3 non-for-profit in the State of California

•The Cytoscape ® Registered Trademark awarded

JOINTLY CODED with Agilent, ISB, Pasteur, Sloan-Kettering, UCSF, Unilever, U Toronto



DNA Damage Networks

Chris Workman Craig Mak Leona Samson (MIT) Tom Begley (U Albany)

Network Evolution:

Silpa Suthram Roded Sharan (Tel Aviv) Richard Karp (Berkeley)

Interpretation of eQTLs:

Silpa Suthram Andreas Beyer Yonina Eldar (Technion) Richard Karp (Berkeley)

Genetic Interactions:

Ryan Kelley, Sourav Bandyopadhyay, Nev Krogan (UCSF)

Cancer Diagnosis:

Han Yu Chuang, Steve Briggs, Tom Kipps, Eunjun Lee (KAIST), Doheon Lee (KAIST)

Funding: NIEHS, NIGMS, NSF, Packard, Agilent, Unilever

Websites: www.pathblast.org; www.cytoscape.org

Networks perturbed by individual genetic variations

(Silpa Suthram)



Cause and effect interactions



Knock-down expression profiles (RNAi, deletion mutants)

OR

Expression QTLs



Knockout causes up-regulation Knockout causes down-regulation

Integration of cause-and-effect interactions with physical networks



Yeang, Mak et al. Genome Biology 2005

Examples

[c] RNA biosynthesis [d] Membrane-bound organelle DCP2 HSP82 DIG1 NDD1 SWI4 ARO80 UME6MET3: VMA6 RTG3 LEU3 SHO1 GAL4 OPI1 LEGEND LEU3 RTG3 TYE7 GLN3 RLR1 THI2 PHO4 RCS1 YAP1801 1.3 0 REB1 PHD1 GAT1 CLC1 ARG81 HMO1 HAP4 CIN5 PPI FHL1 MET32 MAL33 GCN4 AD1 RG5,

3 5 -Log(p-value) TF-DNA Target Gene

Silpa Suthram

Suthram et al. Nature/EMBO MSB 2008