From experimental data to mechanistic hypotheses: analysis of proteomic data using a very large-scale causal model

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#### Abstract

High-throughput proteomic analyses of tissue and bio-fluid samples can yield datasets comprising measured differences in hundreds - or even thousands - of proteins. In principle. this rich source of data can provide a systems-level view of the biological processes in an experiment, leading to testable hypotheses describing the mechanisms that led to the observed changes. But typically, the integration of hundreds of observations to infer the active biological networks is an unmanageable task, limiting the analysis to categorization of the changed proteins by annotations and by patterns of modulation. To identify disease mechanisms, compound mechanisms and biomarkers from proteomic and systems biology experiments requires the development of a model of biology. Using a mental model, a scientist can reason about hundreds of distinct molecules present within a cell, but reasoning over tens of thousands of molecules and their interrelationships is impossible. We describe the development and application of a very large-scale causal, computable model of biology which has been used to identify molecular cause and effect hypotheses consistent with data from proteomic experiments. Automated causal analysis can be used to define upstream networks of molecular events which could result in experimentally observed protein changes. It can be used to identify possible causal pathways linking initial experimental perturbations to observed protein or phenotypic changes. Large-scale causal analysis is a powerful new systems-based approach for the interpretation of molecular state measurements in drug discovery.

### Goal: To be relevant, proteomic data must empower a systems-level understanding of biology

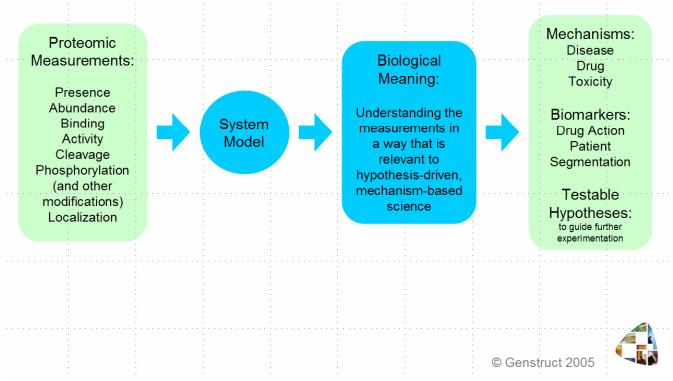


Figure 1. Proteomic data and systems level understanding

Challenge: Synthesizing scientific knowledge with proteomic data is an enormous task; Reasoning about the entire system is beyond human capabilities

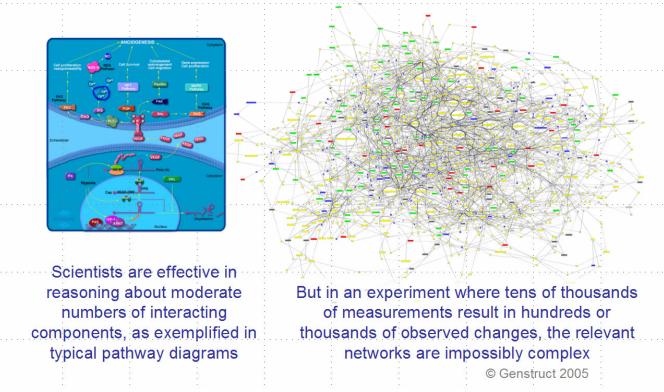


Figure 2. The conceptual challenge of high-throughput data

Our Approach: Develop a computable framework for representing biological knowledge designed to facilitate reasoning about changes in molecular and phenotypic states. Biological Hypotheses: Knowledge Mechanisms Automated System Computer Aided Biomarkers Model Modeling Reasoning Biological Points of Measurements intervention Proteomic Genomic Metabonomic Phenotypic © Genstruct 2005

Figure 3. Computable framework for reasoning about biological state changes

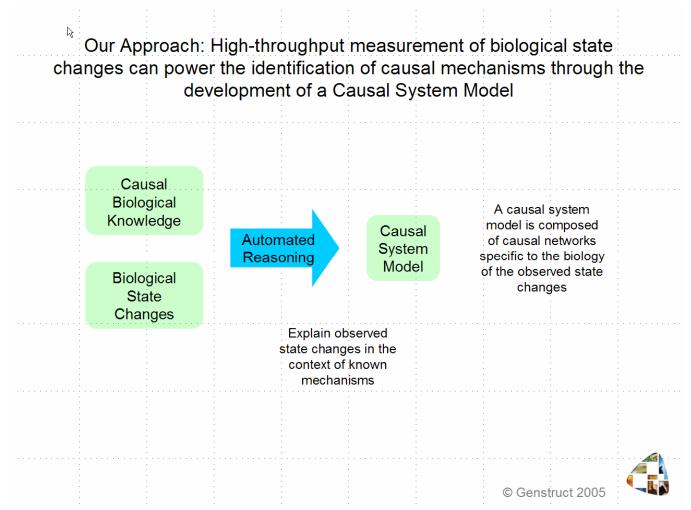


Figure 4. Causal System Models

Results: A technology platform and knowledge environment for the development of Causal System Models which explain proteomic and other panomic state changes and the generation of testable hypotheses within those models.

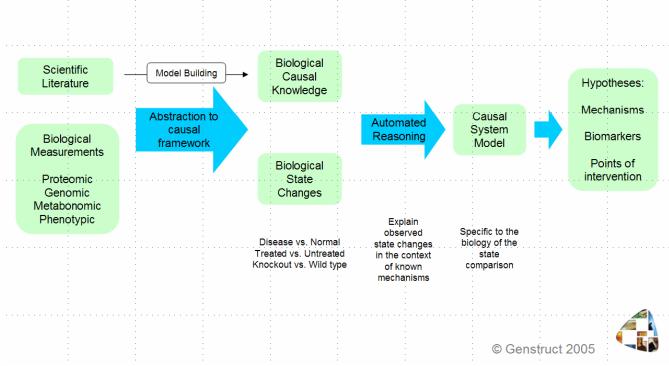


Figure 5. An integrated technology platform and knowledge environment

# Framework: A compact, precise ontology of molecular entities, their activities and modifications, biological processes and locations

Entity	Meaning	Measurement
X	Abundance of X	Proteomics, Metabonomics
exp(X)	Gene expression of X	RNA profiling
catof(X)	Catalytic activity of X	Activity assays
kaof(X)	Kinase activity of X	Kinase activity assay
X {P@Y}	X phosphorylated at tyrosine	Phosphoprotein-mapping
taof(X)	Transcriptional activity of X	Promoter-binding assays

Examples of entities used in the Genstruct knowledge representation



Figure 6. Examples of entities used in the Genstruct knowledge representation

Framework: Simple causal relationships connect the molecular entities, their activities and modifications, and biological processes.

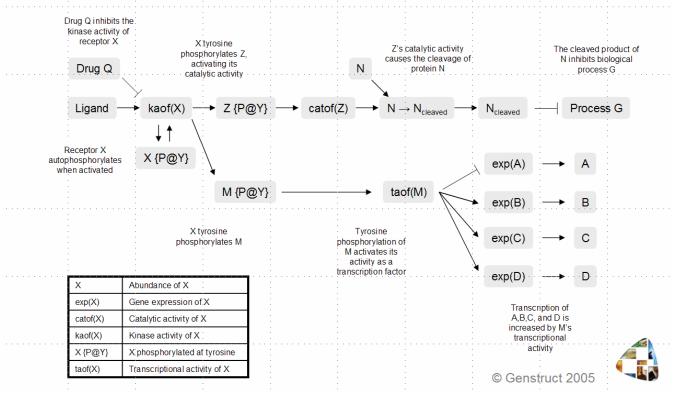


Figure 7. Example of representation of a simple causal network

Experimental Data: Significant molecular differences between well-defined biological states are abstracted as "state changes" associated with specific concepts in the framework.

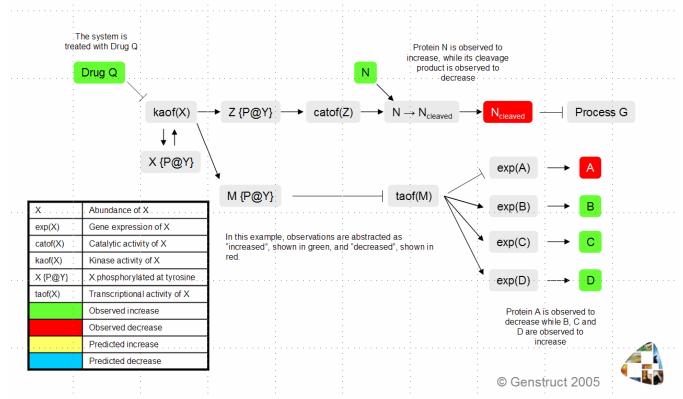


Figure 8. Biological state changes associated with specific entities

Predictions: The biological states of each step in the network can be predicted, based upon the observed experimental data and the causal relationships defined in the model

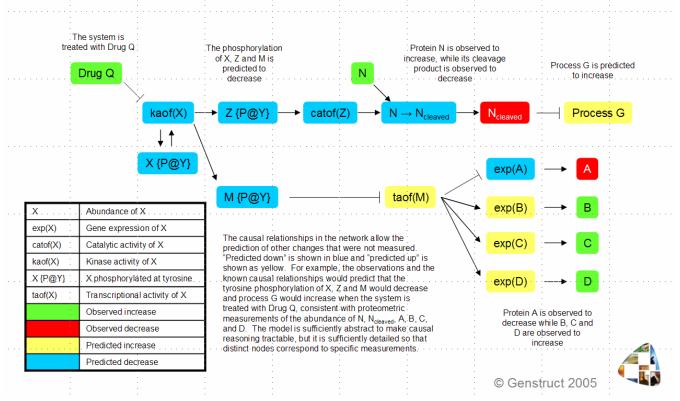


Figure 9. Predictions of state changes based on observed changes and causal relationships

Synergy: Biological States from other panomic measurements augment the predictive ability of the model, and validate the proteomic measurements The system is treated with Drug Q The phosphorylation Protein N is observed to of X, Z and M is increase, while its cleavage Process G is predicted predicted to product is observed to Drug Q to increase. decrease decrease kaof(X) Z {P@} Process G catof(Z) X {P@Y} exp(A) M {P@Y} taof(M) exp(B) Abundance of X Gene expression of X exp(X) mRNA expression profiling of the transcripts encoding exp(C) catof(X) Catalytic activity of X proteins A, B, C, and D confirm the predictions from the proteomic data in three out of 4 cases, lending greater kaof(X) Kinase activity of X confidence in the identification of the proteomic changes. .X {P.@Y} and enabling a further interrogation of changes that may X phosphorylated at tyrosine D. exp(D) be below standard noise levels. In addition, the taof(X) Transcriptional activity of X proteomic measures provide additional support to the observed gene expression measurements, and result in Observed increase Gene Expression greater confidence in the model.

Figure 10. Integration of data from multiple measurement modalities in a causal framework

Observed decrease Predicted increase Predicted decrease

Microarray Profiling

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## Reasoning: Causal System Models are created by identifying the most explanatory biological networks and merging them into a single, internally consistent model

- All possible explanations for each observed change are explored
- Each explanation is evaluated as a hypothesis:
  - How many observed changes can it's predictions explain?
  - How consistent are it's predictions with the observed changes?
  - Is the support for the hypothesis by the observed data significantly better than could occur by chance?
- The highest ranking biological networks are assessed for mutual compatibility and their relationships to relevant phenotypes and experimental perturbations.
- An internally consistent set of networks are selected and merged into a Causal System Model of the mechanisms which differ between the biological states compared in the experiment.



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Figure 11. Causal reasoning to generate and select hypotheses

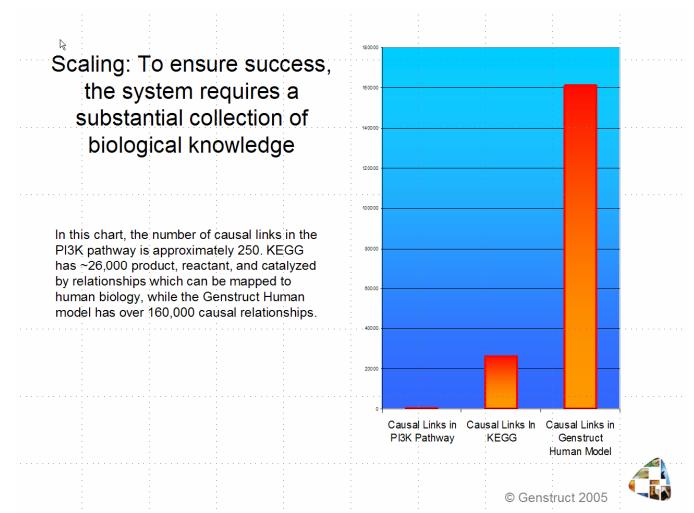


Figure 12. Scale of the system

#### Causal System Modeling shifts the hypothesis-driven research paradigm to handle systems-level measurements

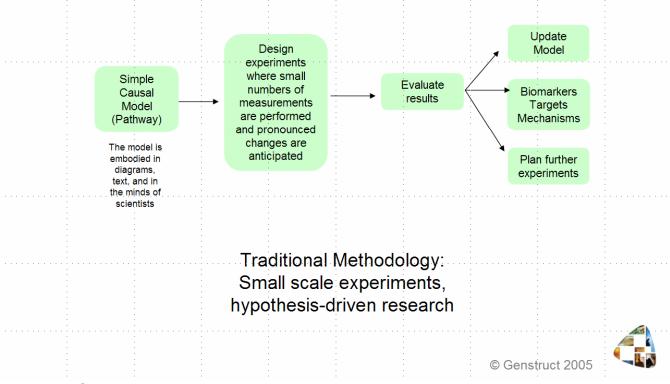


Figure 13. Small scale experiments and hypothesis-driven research

B Causal System Modeling shifts the hypothesis-driven research paradigm to handle systems-level measurements Cluster changes by Lack of patterns over time mechanistic or treatment interpretation Design makes it experiments difficult to Evaluate Categorize changes where biological update the results: by annotation and states can be model or Simple compared by look for statistical plan further Causal overrepresentation thousand or experiments. Model Most of the tens of observed thousands of Ignore changes changes distinct outside of simple involve measurements causal model **Biomarkers** proteins that found by are not part statistical of the causal Focus on a small means may model number of not be pronounced grounded in changes for in-depth mechanisms investigation High Throughput Methodology: Abundant measurements overwhelm simple causal models, limiting the relevance of the data to hypothesis-driven research © Genstruct 2005

Figure 14. High throughput methodology and hypothesis-driven research

#### Causal System Modeling shifts the hypothesis-driven research paradigm to handle systems-level measurements

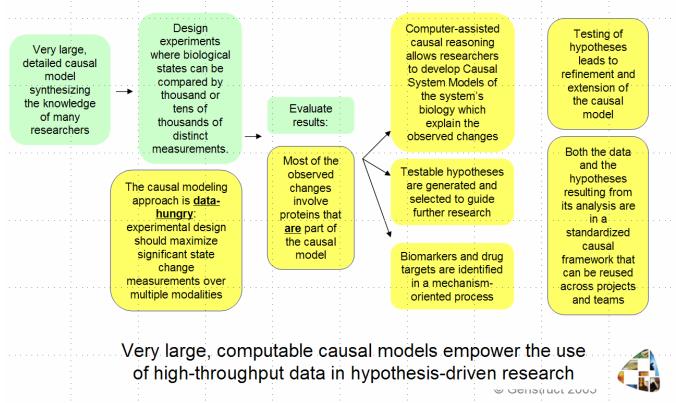


Figure 15. Causal System Modeling and hypothesis-driven research

### Causal System Models enhance the value and utility of proteomic measurements

- Because the framework integrates measurements from multiple modalities (proteomic, genomic, metabonomic), each gains value by synergy with the others.
  - RNA transcript abundance can be compared with protein abundance
  - Changes in enzyme abundance can be compared to changes in substrates
  - Ambiguous downstream abundance changes can be resolved to specific upstream signaling pathways by phosphorylation assays
- Specific identification of protein species becomes more valuable when the Causal System Model uses their changes to distinguish between competing hypotheses
  - Full-length proteins vs. cleavage products
  - Identifying changes in protein modification
  - Distinguishing isoforms and splice variants
  - Verifying species origin in xenograft models



Figure 16. Causal System Models enhance the value and utility of proteomic measurements

#### Summary

- Large scale causal models are a practical means to incorporate high-throughput proteomics data into hypothesis-driven, mechanism-based research
  - Generation of testable hypotheses
  - Identification of biomarkers
  - Selection of novel drug targets
- Large scale Causal Models can integrate and simultaneously exploit measurements from multiple 'omics technologies.
- Causal System Modeling shifts the research paradigm to make high-throughput measurements a critical part of hypothesis driven research.
- Causal system modeling has been successfully applied to a diverse range of biology and model systems
  - Breast cancer
  - Prostate cancer
  - Muscle hypertrophy and atrophy
  - Type 2 diabetes
  - Vascular inflammation
  - Dyslipidemia
- Construction and effective use of large scale causal models depends upon a compact, precise ontology focused on simple causal relationships.
- Due to the conserved nature of biology, once biological knowledge is encoded, it can be transparently reused, both within a disease area and across mammalian biology
- Our methodology generates Causal System Models that characterize disease states
  or other biological phenomena. Because they are fully supported by the literature
  references that underlie each causal connection, they are both a mechanism for
  "what if" predictions and a dynamic knowledge resource.

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Figure 17. Summary