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Toxicogenomics: The Use of 'Omics' to Better Understand the Impact of Adverse Effects from Environmental Exposures on Human Health

Richard S. Paules, Ph.D.

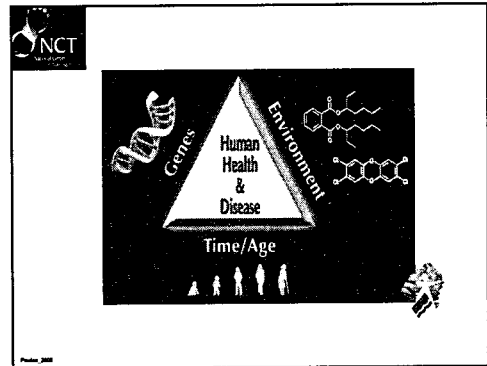
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The mission of the National Center for Toxicogenomics of the National Institute of Environmental Health Sciences of the U.S. National Institutes of Health is to utilize genomic approaches to investigate environmental effects on the etiology and progression of injury and disease processes. Thus a key goal is to design and conduct seminal studies that provide definition to and stimulate development of the field of toxicogenomics, integrating global "omics" approaches into conventional studies of toxicity and disease processes. Toxicogenomics, as defined by the NCT, combines genetics, genome-wide mRNA expression analysis (transcriptomics), cell and tissue-wide protein expression analysis (proteomics), and metabolite profiling (metabolomics) with conventional biology, physiology, pathophysiology and toxicology in an effort to understand adverse affects of gene-environment interactions on human health. Core to the NCT research strategy is the concept of phenotypic anchoring in which studies are designed to relate specific alterations in gene expression to specific adverse effects of environmental stresses defined by conventional parameters of toxicity and pathology such as clinical chemistry, histopathology, etc. To accomplish this task, studies have been designed that utilize a variety of agents that elicit a similar adverse response at a variety of doses and a variety of times of treatment that would elicit the full range of biological responses to those agents. In addition, attempts have been made to incorporate exposures to related but non-adverse agents when possible and to analyze biological responses in additional tissues that do not seem to suffer the same adverse effect (non-target tissue). When analysis of the experimental results implicates a critical role of a particular biological process or a critical role of a particular gene in the response, additional experiments are designed to test those hypotheses concerning their roles. Studies therefore are designed both to gain insight into mechanisms of injury and disease initiation and progression, and to establish signatures of adverse effects, linking gene expression alterations to specific parameters of well-defined

indices of injury and disease to develop putative biomarkers. One aspect of this is to develop biomarkers that are reflective of incipient injury or disease before the culmination of severe injury or disease in order to develop true "predictive toxicology" through the use of toxicogenomics. Our strategy to accomplish this is to build a compendium of signatures linked to environmentally important patho-biological endpoints. In order to learn more about processes involved in acute liver injury and to investigate the true power of genomics to provide insight into mechanisms of injury as well as to provide profiles of processes of injury, we selected a single agent to focus our research efforts initially. Acetaminophen (APAP) was selected as an appropriate model hepatotoxicant due to the well-defined adverse phenotypic endpoints in the liver following toxic exposures, the similarities in metabolism between rodent and human and the relevance of exposure to humans. Studies were designed to test several hypotheses including whether genomic analyses of the liver could reveal indicators of incipient injury before that injury was manifested in such a severe manner as to be detected by traditional indices of liver injury. We were able to demonstrate that gene expression analysis can yield signatures of incipient toxicity after exposure to sub-toxic doses of the toxicant of interest (Heinloth et al., *Toxicol. Sci.*, 80:193ff, 2004). In this manner, the NCT program is striving to integrate conventional biology, genetics, pathology and toxicology with emerging "omics" technologies in order to develop useful insights and potential biomarkers to aid in improving human health.

Toxicogenomics: The use of 'omics' to better understand the impact of adverse effects from environmental exposures on human health

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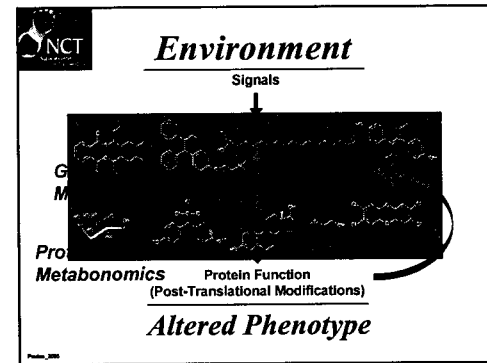
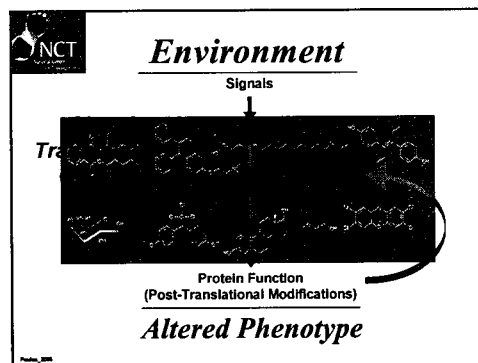


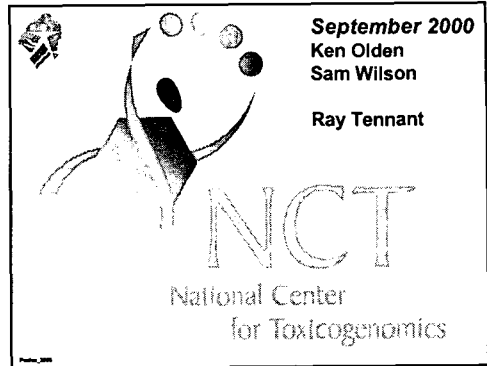
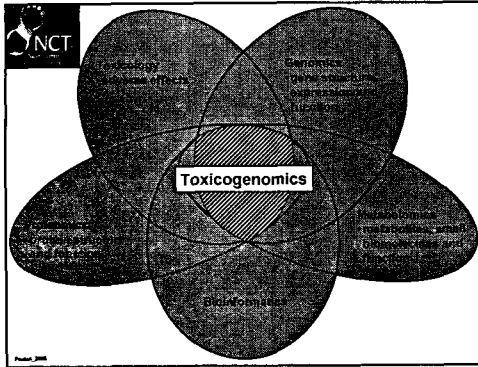
“Environment”

- Industrial chemicals
- Agricultural chemicals
- Physical agents (e.g., UV, IR)
- By-products of combustion and industrial processes (e.g., dioxin)
- Foods and nutrients
- Pharmaceuticals
- Lifestyle choices and substance abuse
- Social and economic factors
- Biological agents

Problems in Human Environmental Health Risk Assessment

- Susceptibility (one-size-fits-all)
- Extrapolation from animal models to human
- Exposure is now measured using indirect surrogates
- Intrinsic toxicity is not known for most of the chemical agents in the environment
- Paucity of knowledge of mechanisms
- Use of default assumptions





Research Opportunities for Toxicogenomics

- Genomic technologies provide an unprecedented opportunity to address highly intractable problems of toxicology and environmental health
- Assess the objective value of surrogate models for prediction of human health risk
- Identify biomarkers of incipient adverse effects
- Harness the results of diverse research efforts for the collective benefit
- Provide a rational basis for risk assessment
- Facilitate the identification of specific susceptibility polymorphisms and relate them to environmental diseases

Research Challenges of Toxicogenomics

- Difficulty in analysis of high density data
- Difficulty in integration of data obtained by different technologies
- Difficulty in linking "omics" data to specific adverse effects (phenotypic anchoring)
- Difficulty in translating statistical assessments into biological understanding
- Limitations of incomplete functional annotation of genome data bases
- Incomplete knowledge of functional pathways and networks, particularly trans-genome relationships
- Inadequate accessible data sources for informatic analysis

Fundamental Hypotheses

- Analyzing global gene expression changes will provide new insight into the mechanisms underlying adverse effects.
- Profiling global gene expression changes will provide signatures that will be highly correlative with and predictive of incipient adverse health effects from environmental stresses.

Main Objectives for the Development of Toxicogenomics

- Identify and understand mechanisms of toxicity - *Discovery Toxicology*
- Identification of biomarkers of toxicity - *Predictive Toxicology*
- Development of Knowledge-Base

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Predictive Toxicogenomics Goals

- Develop Signature Patterns of Exposure from Toxicological Gene Expression Data Sets for use in Classifying Compounds
- Define Gene Expression Signatures Patterns of Toxicant-Induced Adverse Effects

Successful Application of Transcriptomics for Compound / Effect Classification

- Burczynski, ME, *et al.*, (2000) *Toxicological Sciences*. 58(2):399-415.
- Bartosiewicz, M, *et al.*, (2001) *Environmental Health Perspectives*. 109(1):71-4.
- Huang, Q, *et al.*, (2001) *Toxicological Sciences*. 63(2):196-207.
- Hughes, TR, *et al.*, (2000) *Cell*. 102(1):199-28.
- Hughes, TR, *et al.*, (2001) *Nature Biotechnology*. 19(4):342-7.
- Hamadeh, HK, *et al.*, (2002) *Toxicologic Pathology*. 30(4):470-82.
- Hamadeh, HK, *et al.*, (2002) *Toxicological Sciences*. 67(2):232-40.
- Hamadeh, HK, *et al.*, (2002) *Toxicological Sciences*. 67(2):219-31.
- Steiner, G, *et al.*, (2004) *Environmental Health Perspectives*. 112(12):1236-48.
- Amin, RP, *et al.*, (2004) *Environmental Health Perspectives*. 112(4):465-79.
- Heinloth, AM, *et al.*, (2004) *Toxicological Sciences*. 80(1):193-202.
- Moggs, JG, *et al.*, (2004) *Environmental Health Perspectives*. 112(16):1589-606.
- Iconix (<http://www.iconixpharm.com/>)
- GeneLogic (<http://www.genelogic.com/>)

Hamadeh, et al., Gene expression analysis reveals chemical-specific profiles. 2002. *Toxicological Sciences*, 67, 219-231.

Peroxisome Proliferators
(Wyeth-14,643, Clofibrate, & Gemfibrozil)

Phenobarbital

d-Mannitol

NIEHS / Boehringer-Ingelheim

Hamadeh, et al., Prediction of compound signature using high density gene expression profiling. 2002. *Toxicological Sciences*, 67, 232-240.

- RNA samples derived from individual livers of chemically exposed rats were sent to NIEHS with identity blinded
- Multiple bioinformatics tools were employed to classify compounds using previously derived data

NIEHS / Boehringer-Ingelheim

Experimental Design

2 Time Points: 1 Day and 14 Days of Daily Oral Dosing

Clofibrate 250 mg/kg/day

Wyeth 14,643 250 mg/kg/day

Gemfibrozil 100 mg/kg/day

Phenobarbital 120 mg/kg/day

D-Mannitol 5000 mg/kg/day

+

+

+

+

+

95% Confidence

Gene list

Gene list

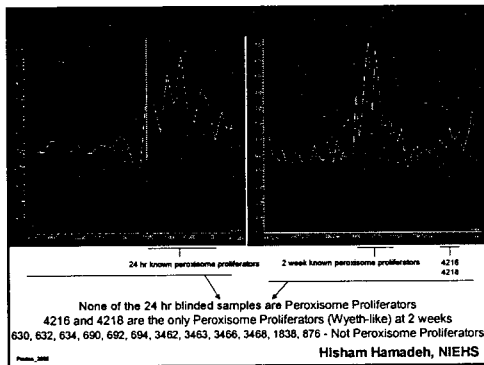
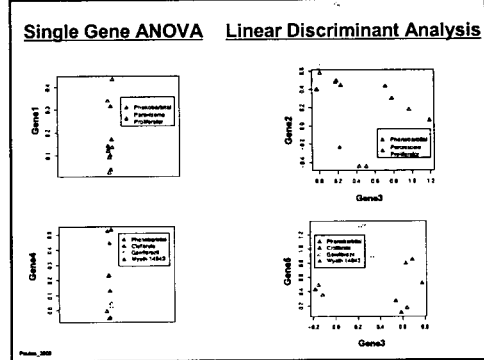
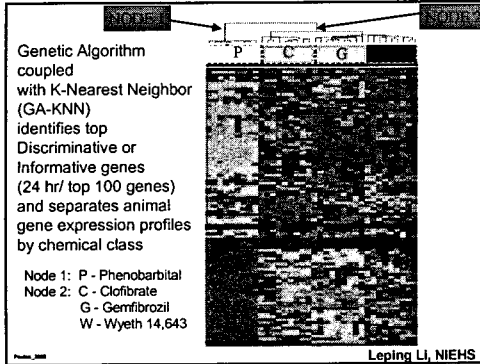
Gene list

Validated Differentially Expressed Genes (2 or 3 out of 3)

Higher order analyses

Hisham Hamadeh, NIEHS / Kerry Blanchard, BIPI

- 31 -



Blinded Samples: Predictions

Sample	Prediction	Time (days)
616	Phenobarbital-like	1
618	Phenobarbital-like	1
672	Phenobarbital-like	14
674	Phenobarbital-like	14
676	Phenobarbital-like	14
678	Phenobarbital-like, Weak	14
688	Phenobarbital-like, Weak	14
270	Clofibrate/Wyeth-like	3
272	Clofibrate/Wyeth-like	3
274	Clofibrate/Wyeth-like	3
276	Clofibrate/Wyeth-like	3
4216	Wyeth-like	14
4218	Wyeth-like	14
10 Others	Not Phenobarbital-like	1, 3, 14

Figure 2000
Hisham Hamadeh, NIEHS

Blinded Samples: 22 / 23 Correctly Identified

Sample	Prediction	Actual	Time (days)
616	Phenobarbital-like	High Dose Phenytoin	✓ 1
618	Phenobarbital-like	High Dose Phenytoin	✓ 1
672	Phenobarbital-like	High Dose Phenytoin	✓ 14
674	Phenobarbital-like	High Dose Phenytoin	✓ 14
676	Phenobarbital-like	High Dose Phenytoin	✓ 14
678	Phenobarbital-like, Weak	High Dose Phenytoin	✓ 14
270	Clofibrate/Wyeth-like	High Dose DEHP	✓ 3
272	Clofibrate/Wyeth-like	High Dose DEHP	✓ 3
274	Clofibrate/Wyeth-like	High Dose DEHP	✓ 3
276	Clofibrate/Wyeth-like	High Dose DEHP	✓ 3
4216	Wyeth-like	High Dose DEHP	✓ 14
4218	Wyeth-like	High Dose DEHP	✓ 14
688	Phenobarbital-like, Weak	High Dose Hexobarbital	14
10 Others	Not Phenobarbital-like	Hexobarbital, Low DEHP, Controls	✓ 1, 3, 14

Figure 2000
Hisham Hamadeh, NIEHS

Limitations of Compound Classification

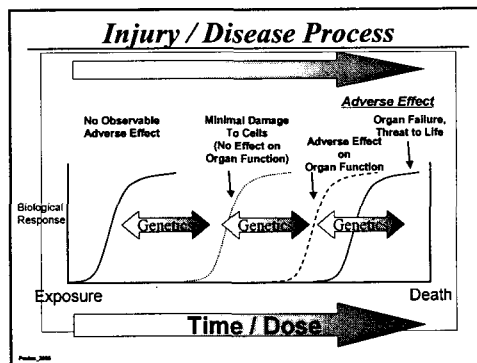
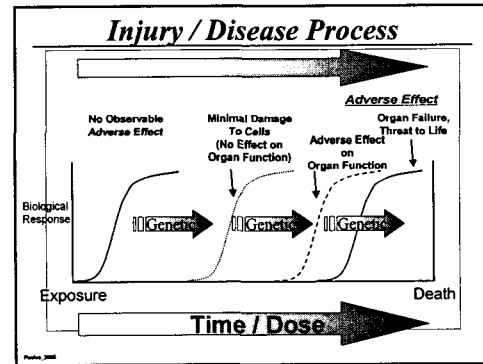
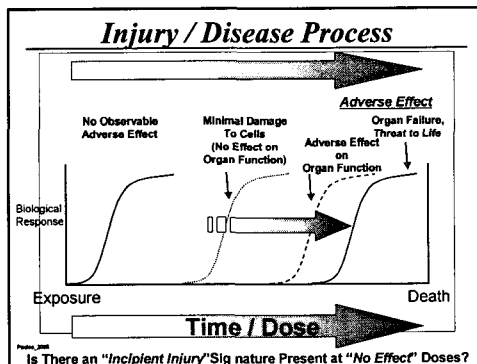
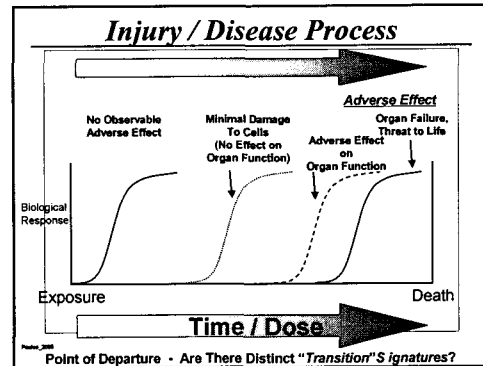
- Are all members of a "class" equivalent hazards?
- What to do with compounds that belong to more than one "class"?
- What to do with novel compounds?
- What to do with mixtures of compounds?

Figure 2000

NCT *Predictive Toxicogenomics Goals*

- Develop Signature Patterns of Exposure from Toxicological Gene Expression Data Sets for use in Classifying Compounds
- Define Gene Expression Signature Patterns of Toxicant-Induced Adverse Effects

Poster_2005



Phenotypic Anchoring of Expression Profiles

Hamadeh, et al., Methapyrilene toxicity: anchorage of pathologic observations to gene expression alterations. 2002. Toxicologic Pathology 30, 470-482.

NIEHS / Boehringer-Ingelheim

Heinloth, et al., Gene expression profiling of rat livers reveals indicators of potential adverse effects. 2004. Toxicological Sciences, 80, 193-202.

NIEHS ToxPath Team

Poster_2005

Prototype: ACETAMINOPHEN

- Human and rodent hepatotoxicant well-studied mechanism
- Significant human exposure
- Human dose range from pharmacological to lethal dose
- 56,000 Emergency room admissions (FDA; 2002)
- > 500 life-threatening exposures with 100 fatalities per year
- Clinical chemistry poor prognosticator of survival
- Rodents and humans metabolize acetaminophen similarly

Prototype: ACETAMINOPHEN

Hypotheses

1. A gene expression signature can be identified that will allow for discrimination between mild and severe liver injury from APAP exposure in rats.
 - 1.1. A gene expression signature can be identified that reveals incipient liver injury that manifests itself only at higher doses and/or at later times.
2. A gene expression signature can be identified in rat blood (similar or dissimilar to that in liver) that will allow for discrimination between mild and severe liver injury from APAP exposure.
3. A gene expression signature can be identified in human blood that will be similar to that identified in rat blood that will allow discrimination between mild injury, severe injury and irreversible liver failure following APAP intoxication.

Experimental Design

3 Time Points: 6, 24, 48 Hrs following Single Oral Dosing

Low
50 mg/kg/day
PreToxic
150 mg/kg/day
Toxic, Recoverable
1500 mg/kg/day

NIEHS ToxPath Team

Alanine Aminotransferase

Histopathological Lesions

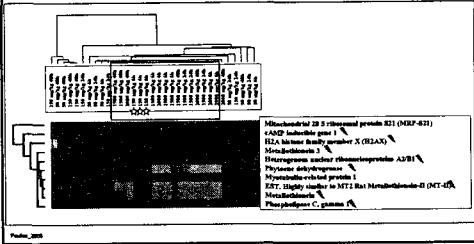
Control
 Centrilobular region, normal hepatic vesicles
 1500 mg/kg 24h
 Swollen, necrotic, w/ inflammatory infiltrates
 1500 mg/kg 48h
 Extensive necrosis with hemorrhage
 1500 mg/kg 48h
 Loss of hepatocyte lines, w/ mononuclear infiltrates

Differentially Expressed Genes

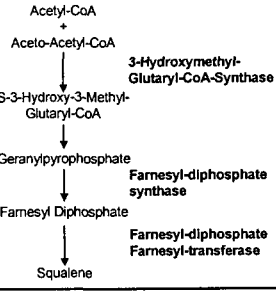
Genes and EST's Differentially Expressed by Dose and Time

Early Indicators of Adverse Effects

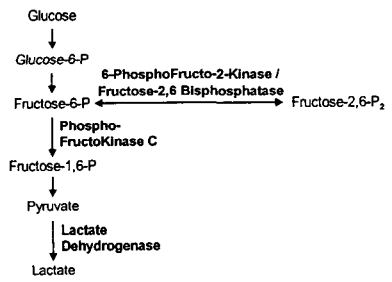
Implicated in Oxidative Stress or DNA Damage Responses:



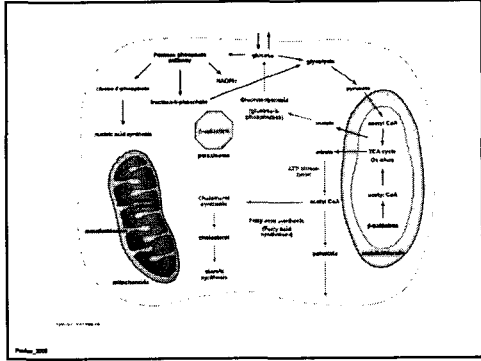
Cholesterol Synthesis



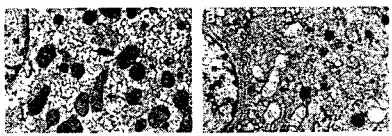
Glycolysis




Enzyme ID	Gene Description	50 mg/kg APAP		100 mg/kg APAP		150 mg/kg APAP	
		5 hr	24 hr	5 hr	24 hr	5 hr	24 hr
AC02297	Adiponectin	0.98	1.02	0.94	1.02	0.97	0.98
AK00402	Chitinase 3-like 1	0.98	1.02	0.94	1.02	0.97	0.98
AK00478	Glutathione S-transferase gamma 1	0.98	1.02	0.94	1.02	0.97	0.98
AK00511	Glutathione S-transferase mu 1	0.98	1.02	0.94	1.02	0.97	0.98
AK00528	Glutathione S-transferase pi 1	0.98	1.02	0.94	1.02	0.97	0.98
AK00538	Glutathione S-transferase theta 1	0.98	1.02	0.94	1.02	0.97	0.98
AK00548	Glutathione S-transferase zeta 1	0.98	1.02	0.94	1.02	0.97	0.98
AK00558	Glutathione S-transferase chi 1	0.98	1.02	0.94	1.02	0.97	0.98
AK00568	Glutathione S-transferase kappa 1	0.98	1.02	0.94	1.02	0.97	0.98
AK00578	Glutathione S-transferase lambda 1	0.98	1.02	0.94	1.02	0.97	0.98
AK00588	Glutathione S-transferase nu 1	0.98	1.02	0.94	1.02	0.97	0.98
AK00598	Glutathione S-transferase xi 1	0.98	1.02	0.94	1.02	0.97	0.98
AK00608	Glutathione S-transferase eta 1	0.98	1.02	0.94	1.02	0.97	0.98
AK00618	Glutathione S-transferase delta 1	0.98	1.02	0.94	1.02	0.97	0.98
AK00628	Glutathione S-transferase mu 2	0.98	1.02	0.94	1.02	0.97	0.98
AK00638	Glutathione S-transferase mu 3	0.98	1.02	0.94	1.02	0.97	0.98
AK00648	Glutathione S-transferase mu 4	0.98	1.02	0.94	1.02	0.97	0.98
AK00658	Glutathione S-transferase mu 5	0.98	1.02	0.94	1.02	0.97	0.98
AK00668	Glutathione S-transferase mu 6	0.98	1.02	0.94	1.02	0.97	0.98
AK00678	Glutathione S-transferase mu 7	0.98	1.02	0.94	1.02	0.97	0.98
AK00688	Glutathione S-transferase mu 8	0.98	1.02	0.94	1.02	0.97	0.98
AK00698	Glutathione S-transferase mu 9	0.98	1.02	0.94	1.02	0.97	0.98
AK00708	Glutathione S-transferase mu 10	0.98	1.02	0.94	1.02	0.97	0.98
AK00718	Glutathione S-transferase mu 11	0.98	1.02	0.94	1.02	0.97	0.98
AK00728	Glutathione S-transferase mu 12	0.98	1.02	0.94	1.02	0.97	0.98
AK00738	Glutathione S-transferase mu 13	0.98	1.02	0.94	1.02	0.97	0.98
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AK00758	Glutathione S-transferase mu 15	0.98	1.02	0.94	1.02	0.97	0.98
AK00768	Glutathione S-transferase mu 16	0.98	1.02	0.94	1.02	0.97	0.98
AK00778	Glutathione S-transferase mu 17	0.98	1.02	0.94	1.02	0.97	0.98
AK00788	Glutathione S-transferase mu 18	0.98	1.02	0.94	1.02	0.97	0.98
AK00798	Glutathione S-transferase mu 19	0.98	1.02	0.94	1.02	0.97	0.98
AK00808	Glutathione S-transferase mu 20	0.98	1.02	0.94	1.02	0.97	0.98
AK00818	Glutathione S-transferase mu 21	0.98	1.02	0.94	1.02	0.97	0.98
AK00828	Glutathione S-transferase mu 22	0.98	1.02	0.94	1.02	0.97	0.98
AK00838	Glutathione S-transferase mu 23	0.98	1.02	0.94	1.02	0.97	0.98
AK00848	Glutathione S-transferase mu 24	0.98	1.02	0.94	1.02	0.97	0.98
AK00858	Glutathione S-transferase mu 25	0.98	1.02	0.94	1.02	0.97	0.98
AK00868	Glutathione S-transferase mu 26	0.98	1.02	0.94	1.02	0.97	0.98
AK00878	Glutathione S-transferase mu 27	0.98	1.02	0.94	1.02	0.97	0.98
AK00888	Glutathione S-transferase mu 28	0.98	1.02	0.94	1.02	0.97	0.98
AK00898	Glutathione S-transferase mu 29	0.98	1.02	0.94	1.02	0.97	0.98
AK00908	Glutathione S-transferase mu 30	0.98	1.02	0.94	1.02	0.97	0.98
AK00918	Glutathione S-transferase mu 31	0.98	1.02	0.94	1.02	0.97	0.98
AK00928	Glutathione S-transferase mu 32	0.98	1.02	0.94	1.02	0.97	0.98
AK00938	Glutathione S-transferase mu 33	0.98	1.02	0.94	1.02	0.97	0.98
AK00948	Glutathione S-transferase mu 34	0.98	1.02	0.94	1.02	0.97	0.98
AK00958	Glutathione S-transferase mu 35	0.98	1.02	0.94	1.02	0.97	0.98
AK00968	Glutathione S-transferase mu 36	0.98	1.02	0.94	1.02	0.97	0.98
AK00978	Glutathione S-transferase mu 37	0.98	1.02	0.94	1.02	0.97	0.98
AK00988	Glutathione S-transferase mu 38	0.98	1.02	0.94	1.02	0.97	0.98
AK00998	Glutathione S-transferase mu 39	0.98	1.02	0.94	1.02	0.97	0.98
AK01008	Glutathione S-transferase mu 40	0.98	1.02	0.94	1.02	0.97	0.98




Mitochondrial Damage after Exposure to 150 mg/kg APAP





Conclusions

- > Genome-based technologies provide an unprecedented opportunity to explore the biological complexity underlying adverse effects.
- > While the opportunities are great, so are the challenges of being able to explore high density data.
- > Progress will occur slowly and incrementally; a solid foundation for high quality data assimilation and analysis must be established.
- > The opportunity now exists for creating an information base which compiles the results of diverse individual studies into a compendium of chemical effects in biological systems that can be a resource for the scientific community.



NCT

- Ray Tennant, NCT Director
- Mike Waters, Assist. Dir. CEBS

<p><i>Current</i></p> <ul style="list-style-type: none"> • Jeff Tucker • Jennifer Collins • Danica Ducharme • Rick Fannin • Sherry Grissom • Stella Sieber • Todd Auman • Kevin Gerrish • Alexandra Heinloth • Pierre Bushel • Jeff Chou • Jianying Li • Jonathan Miller 	<p><i>NMG</i></p> <p><i>Past</i></p> <ul style="list-style-type: none"> • Cindy Afshari • Hisham Hamadeh • Emile Nuwaysir • Rupesh Amin • Lee Bennett • J. Carl Barrett <p><i>Boehringer-Ingelheim</i></p> <ul style="list-style-type: none"> • Ray Stoll • Kerry Blanchard • Supriya Jayadev 	<p><i>NCT ToxPath</i></p> <ul style="list-style-type: none"> Alexandra Heinloth Gary Boorman Mike Cunningham Julie Foley Rick Irwin Leping Li Alex Merrick Paul Nettesheim Nigel Walker <p><i>SAS Institute</i></p> <ul style="list-style-type: none"> • Russ Woffinger <p><i>Molecular Mining</i></p> <ul style="list-style-type: none"> • Steve Milsener
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