



Modeling molecular diversity in cancer



*Integrating “omics”, mathematical models
and functional cancer biology*

Integrative Cancer
Biology Program

$$\frac{\partial n}{\partial t} = D_n \nabla^2 n - \chi \nabla \cdot (n \nabla f)$$



Modeling molecular diversity in cancer

Identifying and understanding “omic” determinants of therapeutic response in breast cancer

- A collection of cell lines as a model of molecular and biological diversity
- Three integrative biology examples
 - Associating pathways and markers with response
 - Modeling MEK signaling diversity using pathway logic
 - Bayesian network models of AKT signaling



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Model requirements

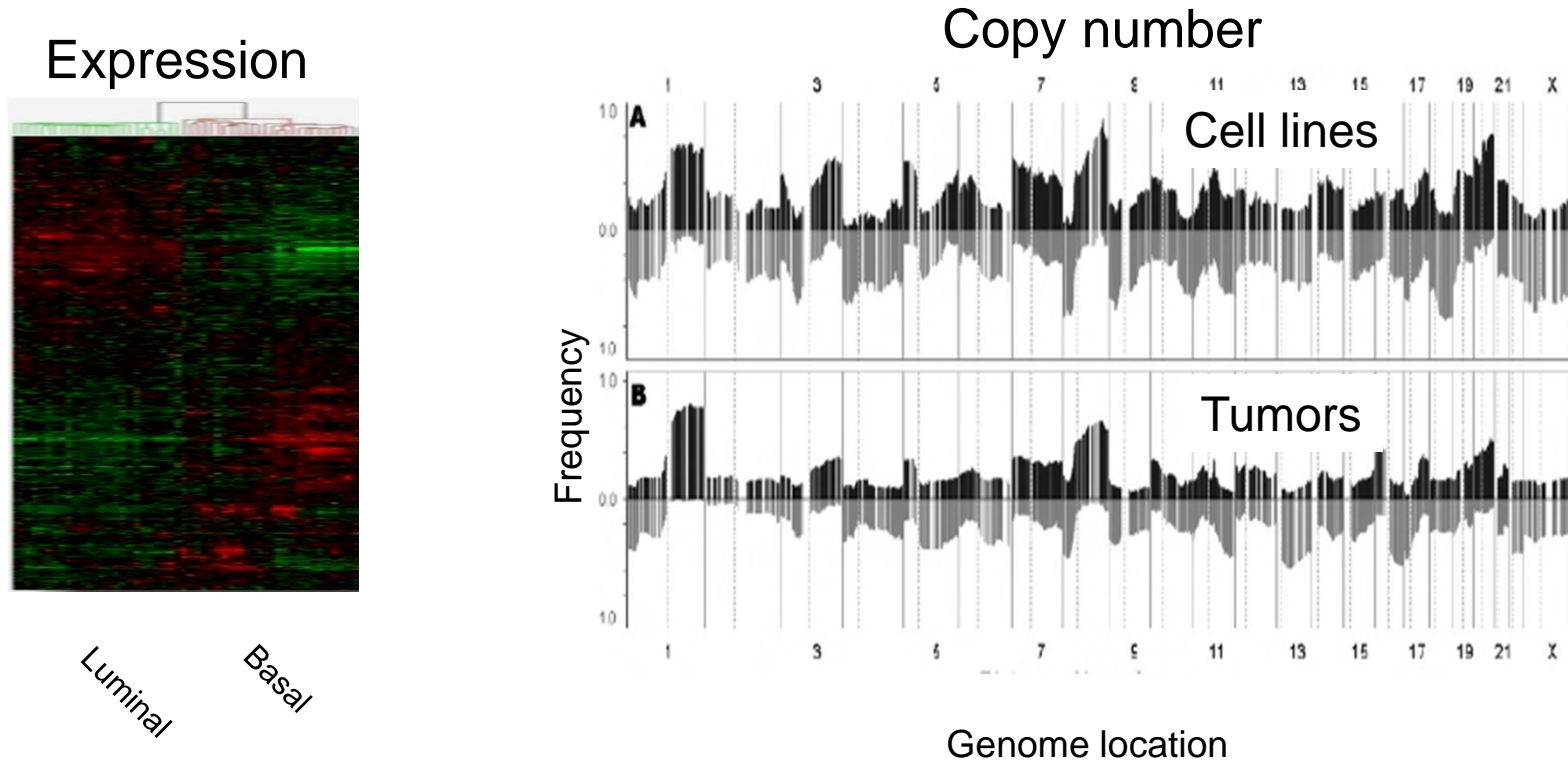
Identifying and understanding “omic” determinants of therapeutic response

- The molecular abnormalities that influence drug response in primary tumors must be functioning in the model
- The panel must have sufficient molecular diversity so that statistical analyses will have the power to identify molecular features associated with response



Cell lines as models of primary breast tumors

A collection of 50 cell lines retain important transcriptional and genomic features of primary tumors



Neve et al, *Cancer Cell* 2006
Chin et al, *Cancer Cell*, 2006



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Associating molecular markers with response to lapatinib

Univariate predictors

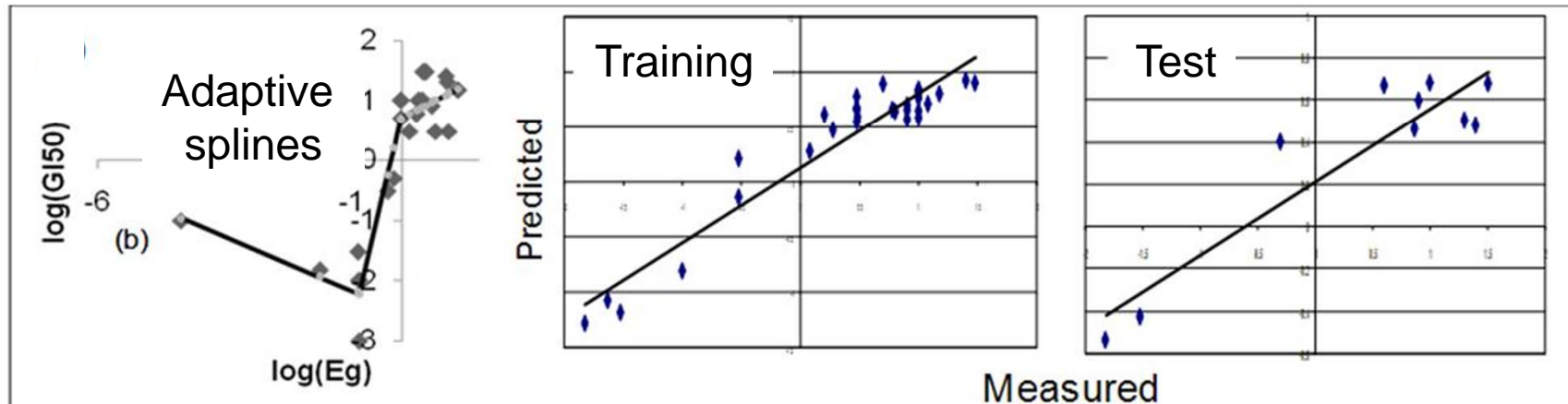
Array CGH

mRNA Expression

Protein Expression

Weighted Voting Methods

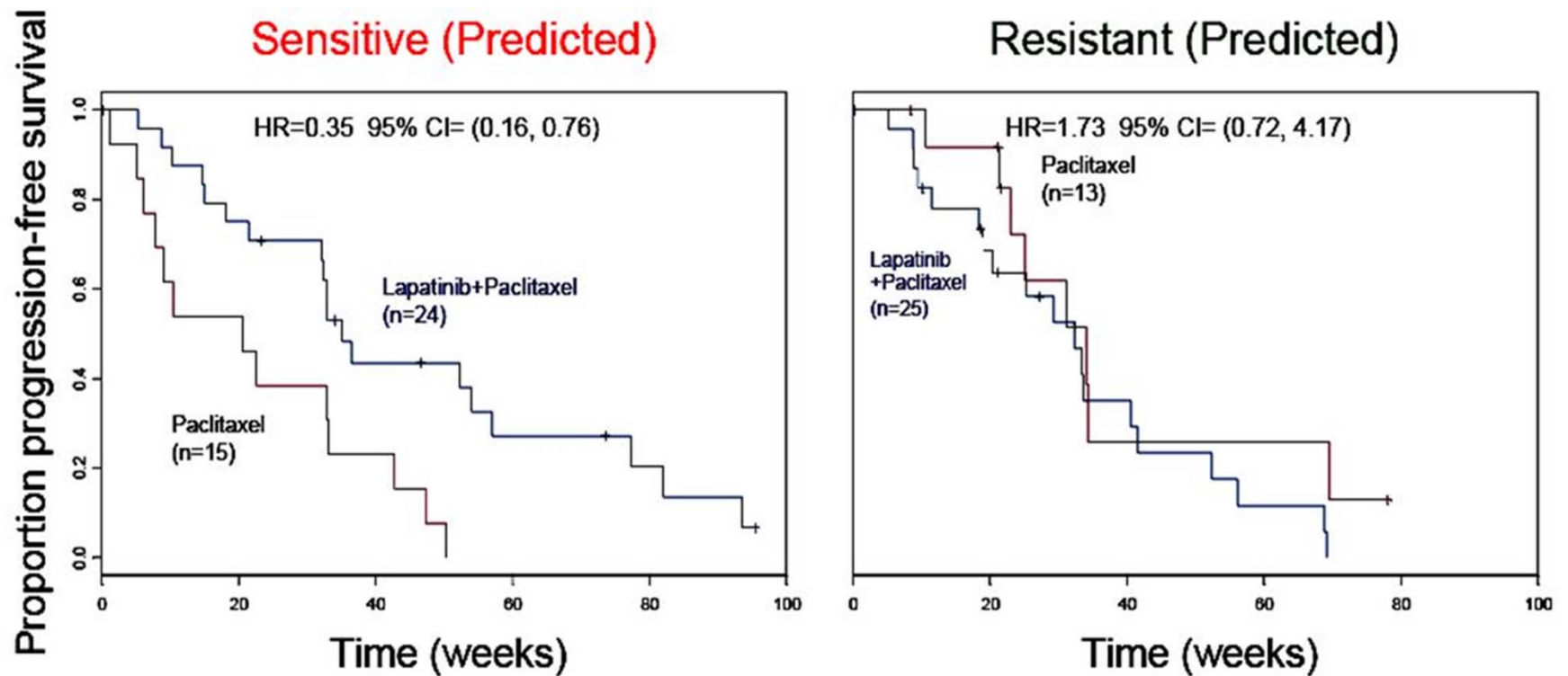
Prediction: Molecular markers and networks associated with sensitivity and resistance will predict clinical response





Test: Cell line markers predict response in HER2 positive patients

EGF30001: A randomized, Phase III study of Paclitaxel + Lapatinib vs. Paclitaxel + Placebo
HER2, GRB7, CRK, ACOT9, LJ31079, DDX5



GSK-LBNL collaboration



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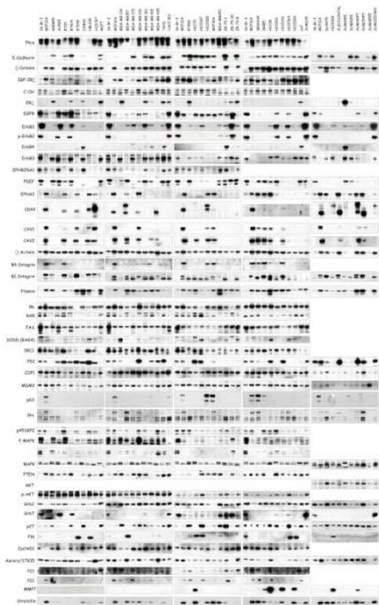
Hierarchical analysis of Pathway Logic states and rules

Curated network model

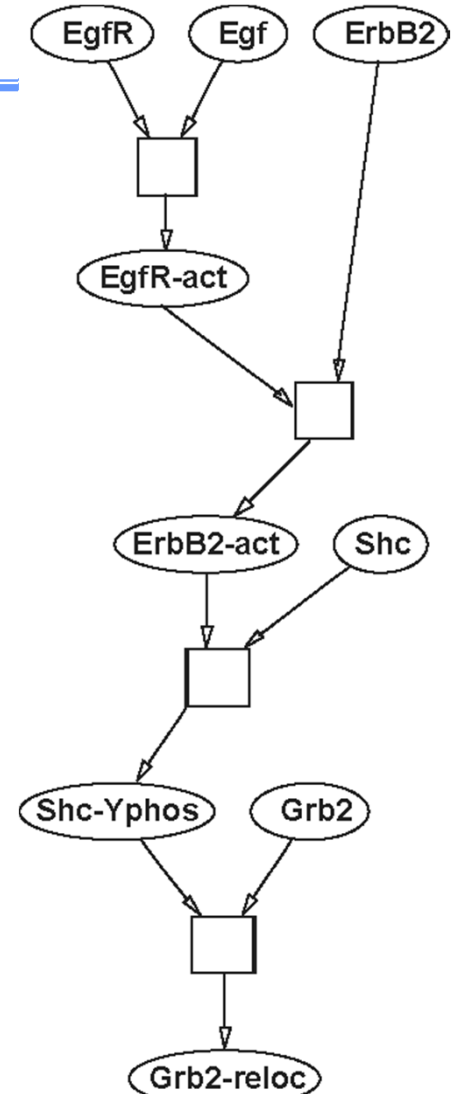
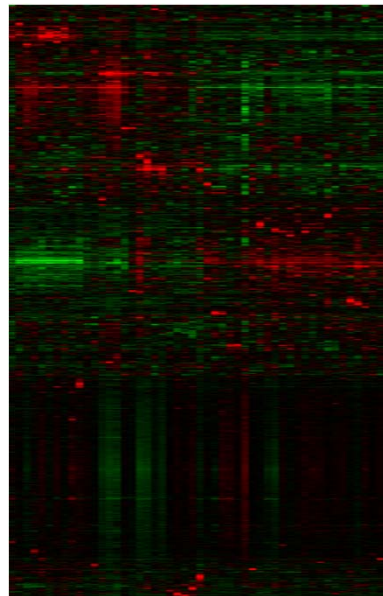
Baseline levels populate PL model states
Rules define predicted pathway activity

Protein abundances

Transcript levels



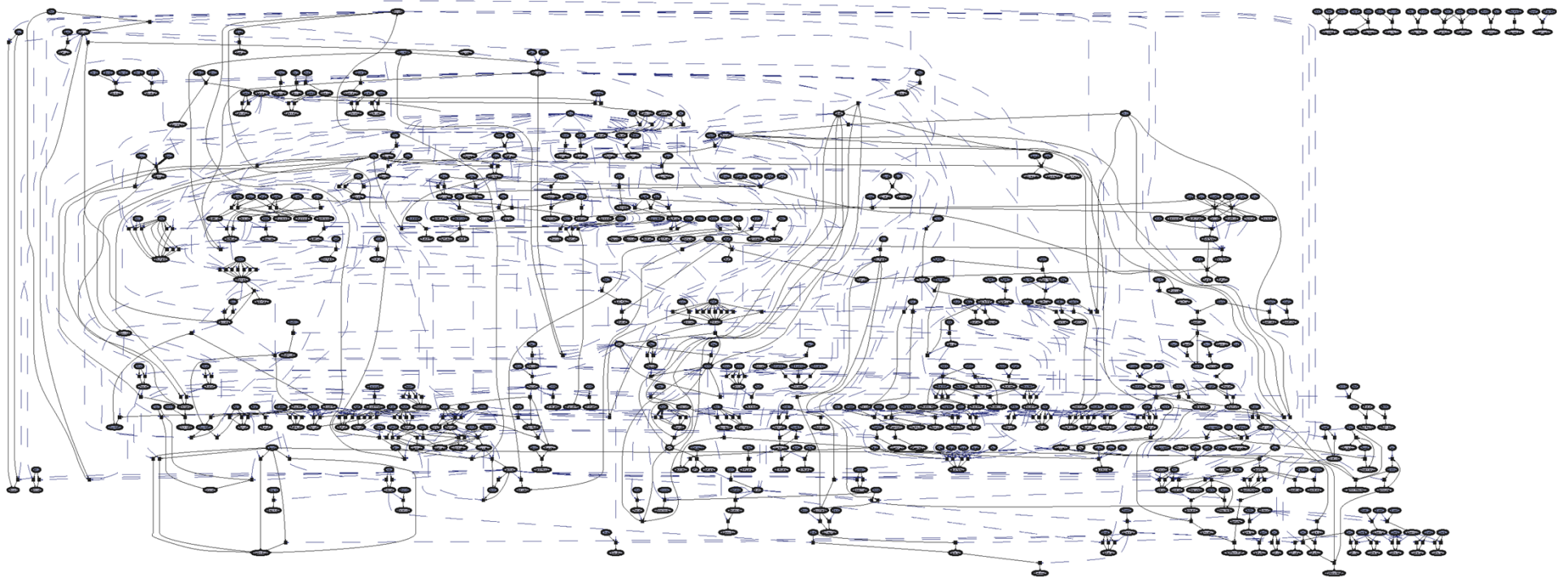
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Heiser, Spellman, Talcott, Knapp, Lauderote

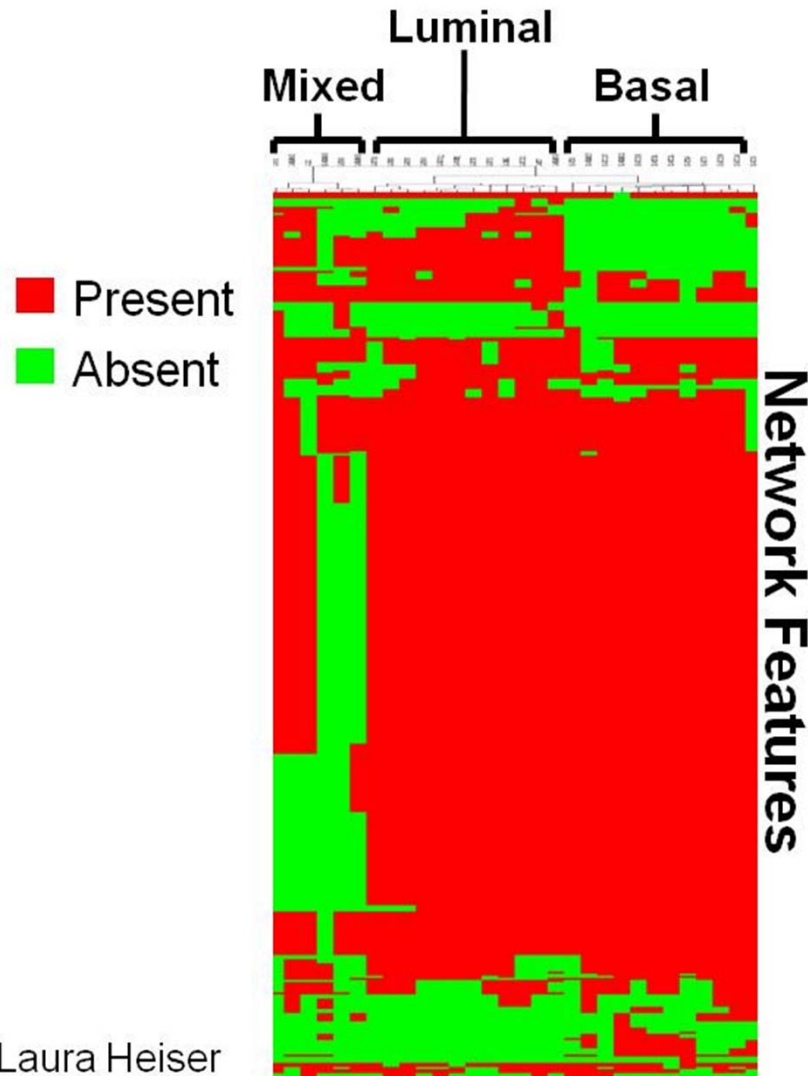


Example network of one cell line

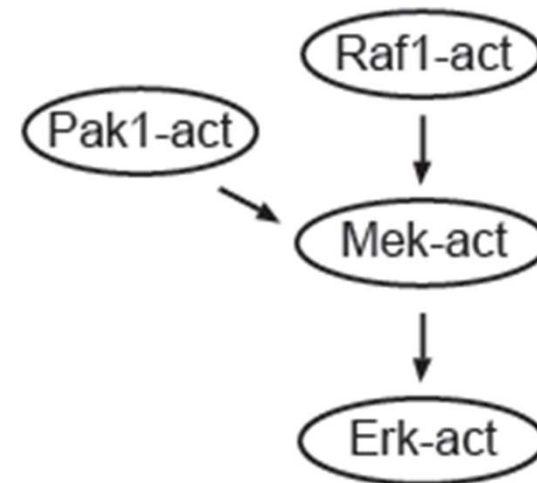




Hierarchical analysis of network features



Prediction: PAK1 is required for network activation of MEK/ERK cascade in luminal cell lines

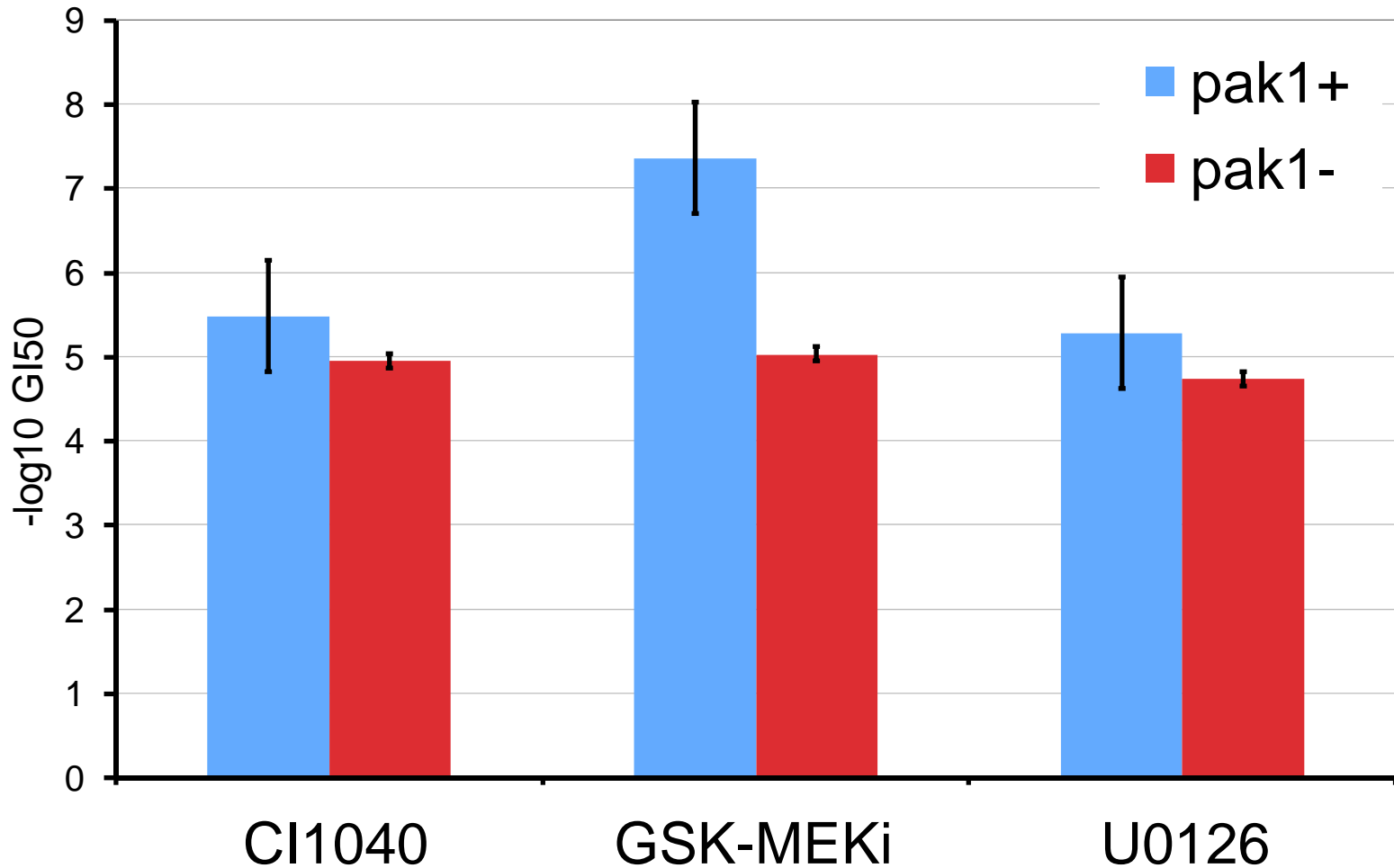


Laura Heiser

$$\frac{\partial n}{\partial t} = D_n \nabla^2 n - \chi \nabla \cdot (n \nabla f)$$



Test: PAK1⁺ luminal cell lines are more sensitive to MEK inhibitors





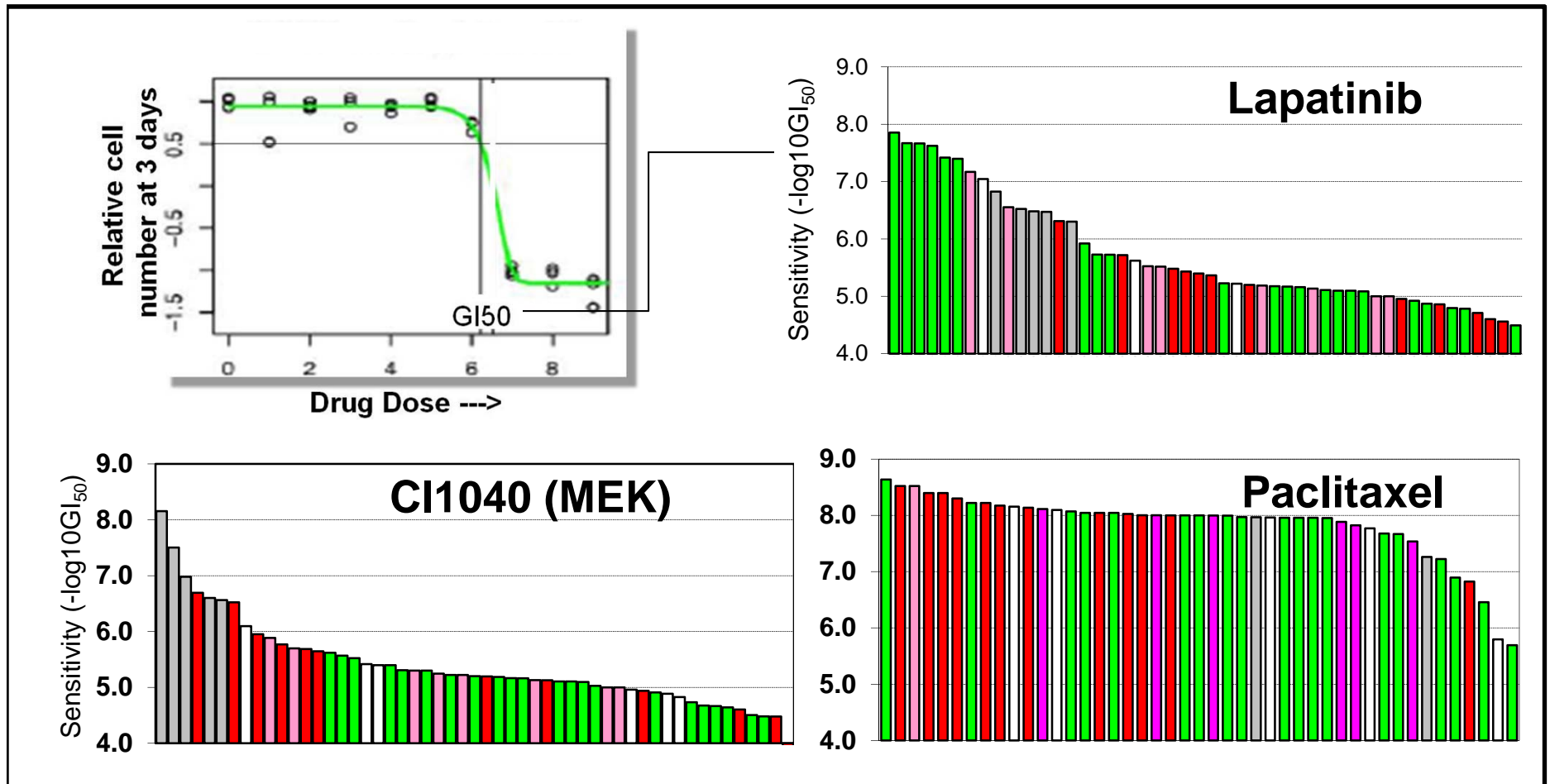
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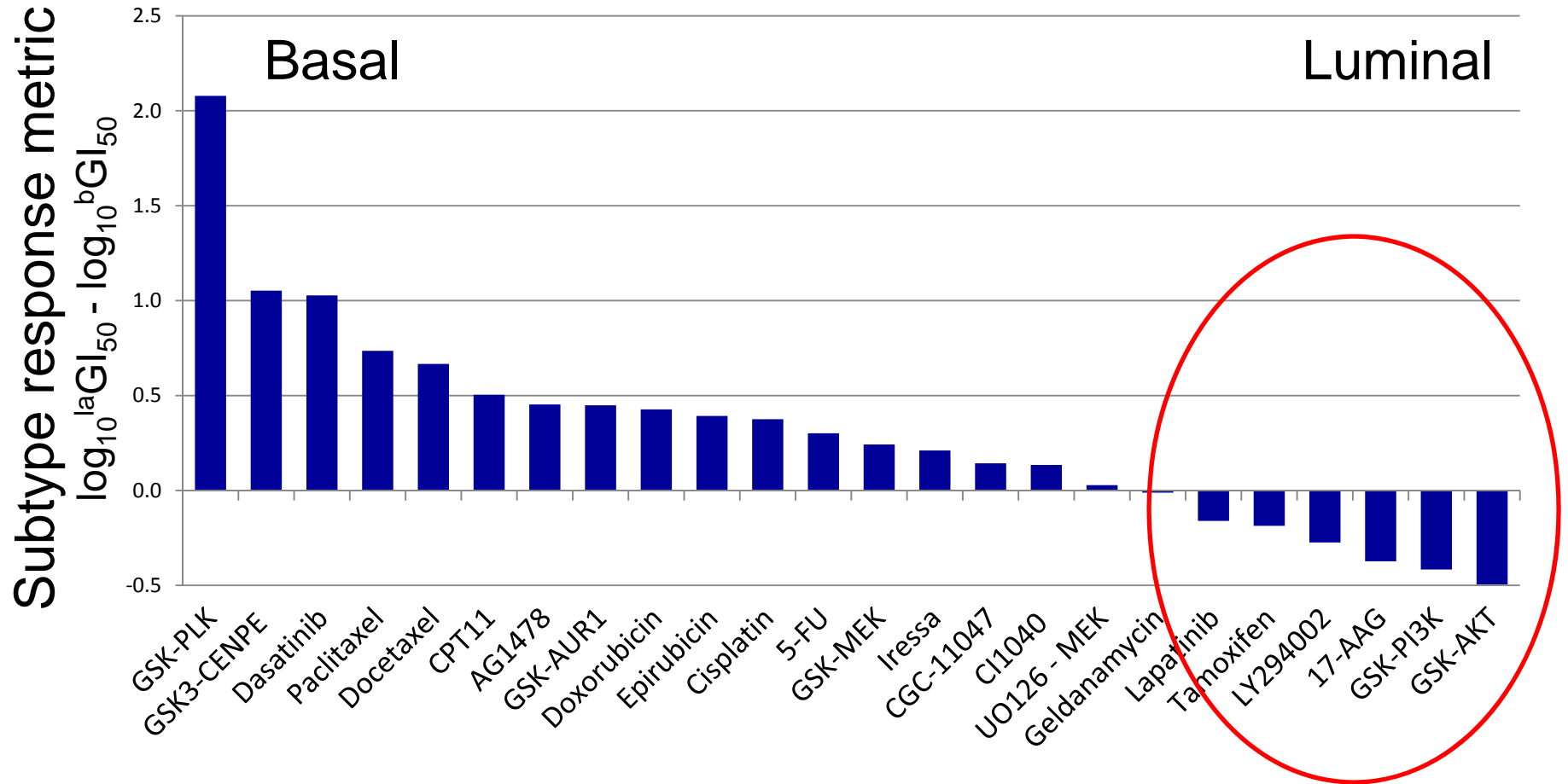
Therapeutic agents show strong luminal subtype specificity



Kuo, Guan, Hu, Bayani 2007



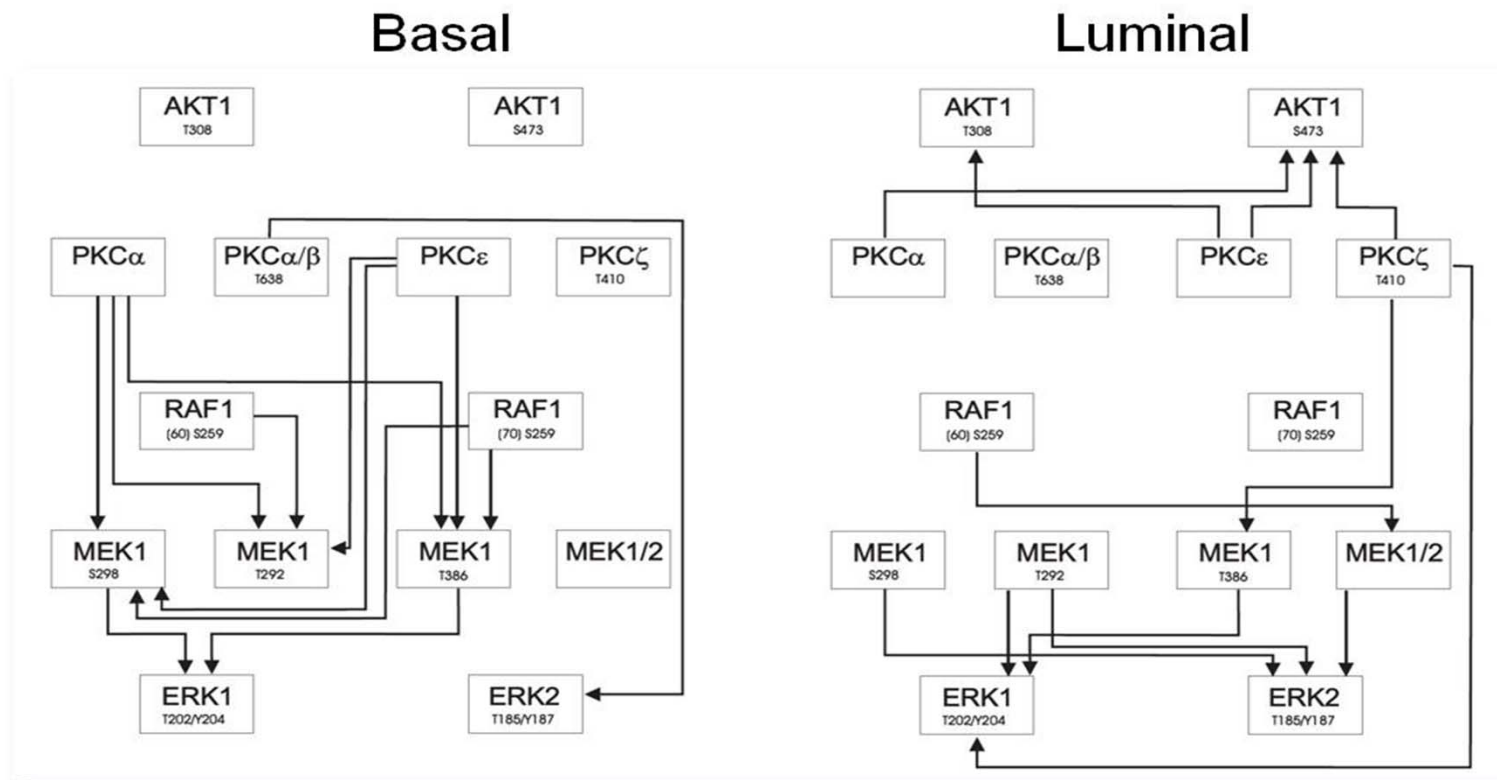
AKT pathway inhibitors show strong luminal subtype specificity





Bayesian network analysis reveals AKT dependent signaling in luminal lines

Prediction: PI3-kinase pathway mutations will occur preferentially in luminal subtype cell lines



Mukherjee , Speed, Neve, et al., 2007





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Collaborating Laboratories & Support



Engineering

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Damir Sudar

Surgery/Pathology

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Tona Gilmer
Barbara Weber
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Frank McCormick
Mary Helen
Barcellos Hoff
Rene Bernards
Gordon Mills

Comp. Biol

Paul Spellman
Laura Heiser
Keith Lauderote
Merrill Knapp
Carolyn Talcott
Sach Mukherjee
Terry Speed
Jane Fridlyand
Bahram Parvin
Lisa Williams
Steve Ashton

ICBP, SPORE, GSK, Affymetrix, Genentech, Panomics,
Cellgate, Cell Biosciences, Komen,
Avon, EGF30001 Trial Investigators

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