# Computational Systems Biology: Biology X

### Bud Mishra

Room 1002, 715 Broadway, Courant Institute, NYU, New York, USA

L#1:(September-13-2010) Cancer and Signals

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Cancer's a Funny Thing: I wish I had the voice of Homer To sing of rectal carcinoma, This kills a lot more chaps, in fact, Than were bumped off when Troy was sacked....

-JBS Haldane, The Enchantress of Florence, 2008.









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# Administrivia

- Instructor: Bud Mishra
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- Office Hours: Mondays 1:30 pm 2:30 pm (and by appt)



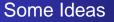
# Administrivia

- BIOLOGY X
- Course Details: G22.3033-002 || Bioinformatics
- Time and Place: Monday, 5:00 PM 6:50 PM EST
  || Room 1221, 719 Broadway
- Number of Credits: 3 credits
- **Course Work:** Software Project, Analyzing Genetics Data, Review Articles
- Languages of Choice: R (May be Python, Matlab, Mathematica — But no Perl please)

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- Signaling is a well-studied phenomenon both in evolutionary game theory and in cell biology.
- In game theory, signaling frameworks have been used to study the evolution of such fundamental phenomena as conventions and cooperation, while in biology, signal transduction has been extensively studied as a basic ingredient to multicellularity, enabling cells to communicate and coordinate.
- However, approaches that span both fields are scarce.



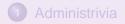
- In this course, we explore the idea of viewing multicellular organisms as signaling systems in the game-theoretic sense, attempting to unify these two perspectives on signaling.
- A multicellular organism corresponds to a population of cells in a cooperative state, with a working signaling system in place.
- We will discuss how the evolution of such a system may be modeled. Then, we will in particular be interested in the breakdown of cooperation, leading to an interpretation of cancer as a disease of multicellularity.
- The course will be as self-contained as possible and include introductions to evolutionary game theory and signaling systems, signal transduction in cell biology, and the biology of cancer.

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- Required Textbook (1): Signals: Evolution, Learning, and Information || Brian Skyrms || Oxford University Press || 2010.
- Required Textbook (2): *Biology of Cancer* || Robert A. Weinberg || Garland Science, 2006.
  - Recommended textbook (1): Game Theory Evolving || Herbert Gintis || Princeton University Press || 2000
  - Recommended textbook (2): Computational Biology of Cancer || Dominik Wodarz and Natalia L. Komarova || World Scientific Publishing Company || 2005
  - Recommended textbook (3): Information Theory, Inference & Learning Algorithms || David J. C. MacKay || Cambridge University Press || 2002







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### **Cancer and Signals**

#### Main Thesis

• Cancer is a complex disease.



- It is associated with the natural somatic evolution.
- 2 It has its origin in multicellularity.
- It makes use of every mechanism in biology.

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- Bioinformatics to study Mechanisms in Biology: Genomics, Transcriptomics (c- and ncRNA), Proteomics, Metabolomics and Signaling
- *Structure of Multicellularity*: Signaling and Signaling Games; Multicellularity, Information Theory and Rate Distortion Theory
- Evolutionary Structure of Somatic Modifications: Repeated Games, Signaling Games and Neologism

Let us think about these inter-connected questions from a single global perspective...

## A Tentative Syllabus

I would like to focus this course on three basic questions...

- Biology of cancer
- Signaling game models of multicellular biology
- Information theoretic utility functions in signaling games

### Possible Sets of Lectures

- Lecture 1: Causes of Cancer
- Lecture 2: A Very Brief Introduction to Game Theory
- Lecture 3: Introduction to Biology
- Lecture 4: The Nature of Cancer
- Lecture 5: Growth Factors, Receptors and Cancer
- Lecture 6: Cytoplasmic Signaling Circuitry
- Lecture 7: Game Theory: Signaling Games

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### Possible Sets of Lectures (Contd.)

- Lecture 8: Repeated Games and Evolutionary Dynamics
- Lecture 9: Probability and Information Theory
- Lecture 10: Rate-Distortion Theory and Game Utility
- Lecture 11: Cancer Models: Genomics
- Lecture 12: Cancer Models: Signaling
- Lecture 13: Cancer Models: Genetics
- Lecture 14: Cancer Models: Therapeutics
- Lecture 15: The Future Challenges

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#### What "causes" cancer?

- Cancer as a disease of the genome;
- Cancer as a somatic evolutionary process;
- Cancer as a price of symbiosis (mitochondrial);
- Cancer as a response to multi-cellularity;
- Sancer as a price of repair/regeneration (stem cells);
- Cancer as a consequence of energy consumption (Anaerobic Glycolysis); Warburg Effect.
- Cancer as a response to external stress; and
- Cancer as a response to the micro-environment (hyperand hypo-methylation).

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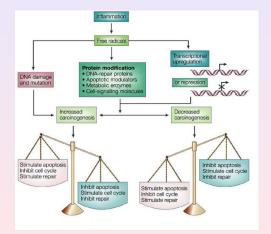
### Processes involved in Cancer

#### Coordination of processes in cancer progression

- Inflammation
- Autophagy and mitophagy
- Apoptosis
- Hypoxia
- Anaerobic glycolysis
- Fibrosis
- Signaling

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- The computational principles to create a Kripke model (with states and state transitions)
- It uses Information Bottleneck Theory (a special case of RDT)



*D<sub>i</sub>* = The random variable; Samples are the rows in submatrix of *D<sub>i</sub>*



• **States** Cluster  $D_i$  =; effectively identifying the state variable  $X_i$  such that the mutual information between  $D_i$  and  $X_i$  is minimized

min  $I(D_i; X_i)$  subj DISTORTION

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• **State-Transitions** Preserve "ontologies"... Relate the conditional distribution *P*(*O*|*X<sub>i</sub>*) with

$$\begin{array}{ll} P(O|X_{i+1}) & (1 \le i < k) \\ \& & P(O|X_{i-1}) & (1 < i \le k) \end{array} \end{array}$$

Measure distortions using Kullback-Liebler Distances.

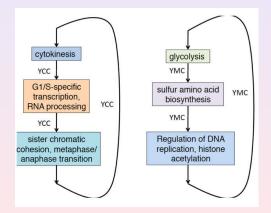
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- Cluster the data over time as well as gene sets...
- Cluster and cluster-edges optimize the mutual information terms

Measure distortions using Kullback-Liebler Distances.

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#### Combined Kripke Model related to cell cycle



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### CAN THESE ALGORITHMS BE USED TO UNDERSTAND ALL THE PROCESSES INVOLVED IN CANCER????

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### Heated Discussions on the Suggested Topics... Resulting in a New and Better Syllabus... That EVERYONE Loves!

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# Projects

- Agent-Based Models of Cancer
- Mathematical Models (ODE and PDE) of Cancer
- Copy-Number Variation Analysis of Cancer Genomics
- Systems Biology of Cancer: Temporal, transcriptomic data
- PPI Networks and Cancer
- Progression in Pancreatic Cancer, Therapeutics (Abraxane + Gem)
- Cancer Genetics Model
- Biological Lab Projects (Single Cell analysis of tumors, Tumor heterogeneity, Tumorigenesis, Dedifferentiation)

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# [End of Lecture #1]

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