

National Cancer Institute

State of the Science in Cancer Research

Dr. John E. Niederhuber
Director, National Cancer Institute

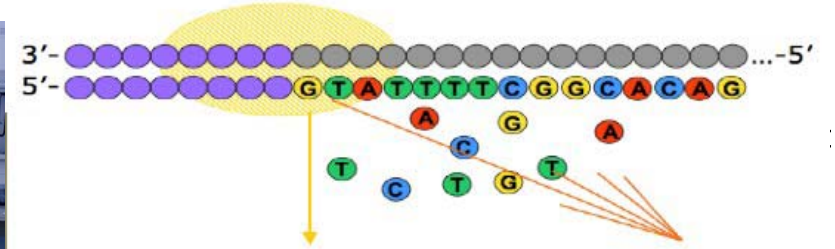
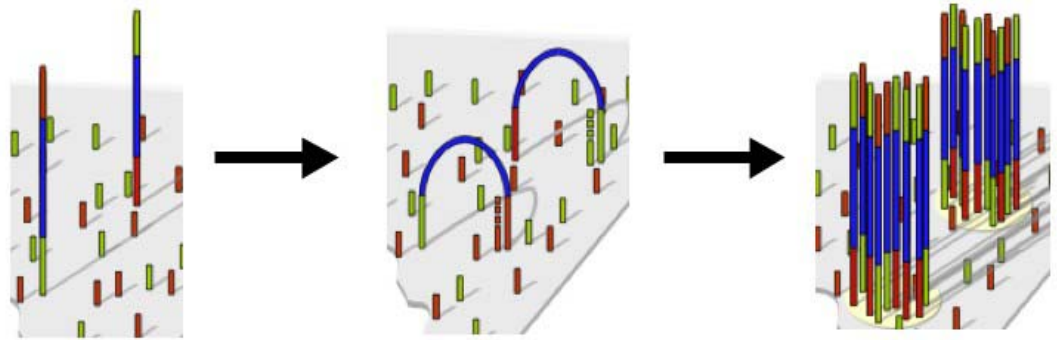
A New Frontier in Oncology: Integrating and
Leveraging the Physical Sciences

February 27, 2008

U.S. DEPARTMENT
OF HEALTH AND
HUMAN SERVICES

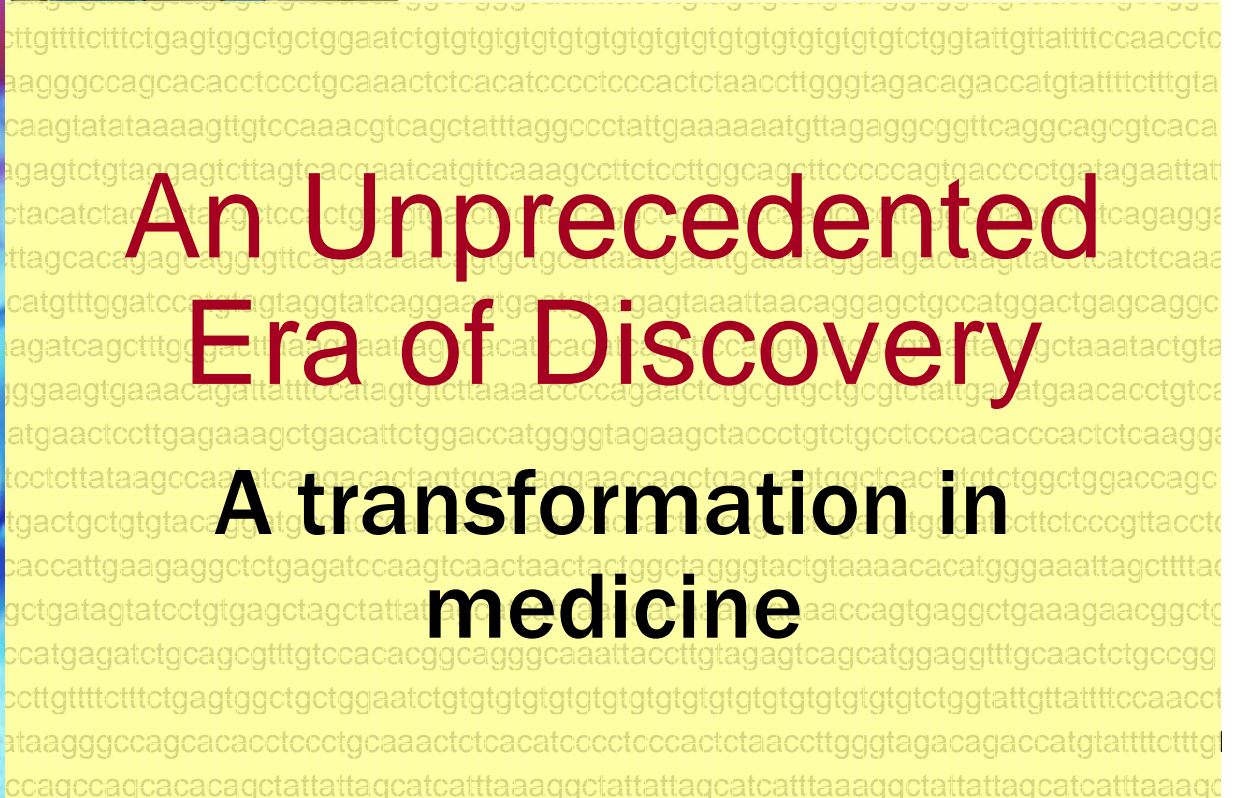
National Institutes
of Health

What can physics,
physical chemistry,
and applied
mathematics bring
to cancer biology?



An Unprecedented Era of Discovery

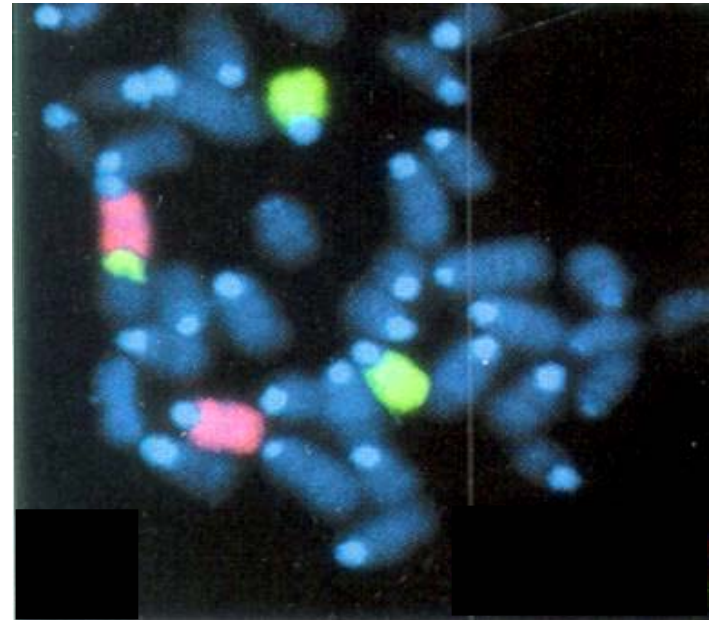
A transformation in medicine



Cancer is a Disease of the Genome

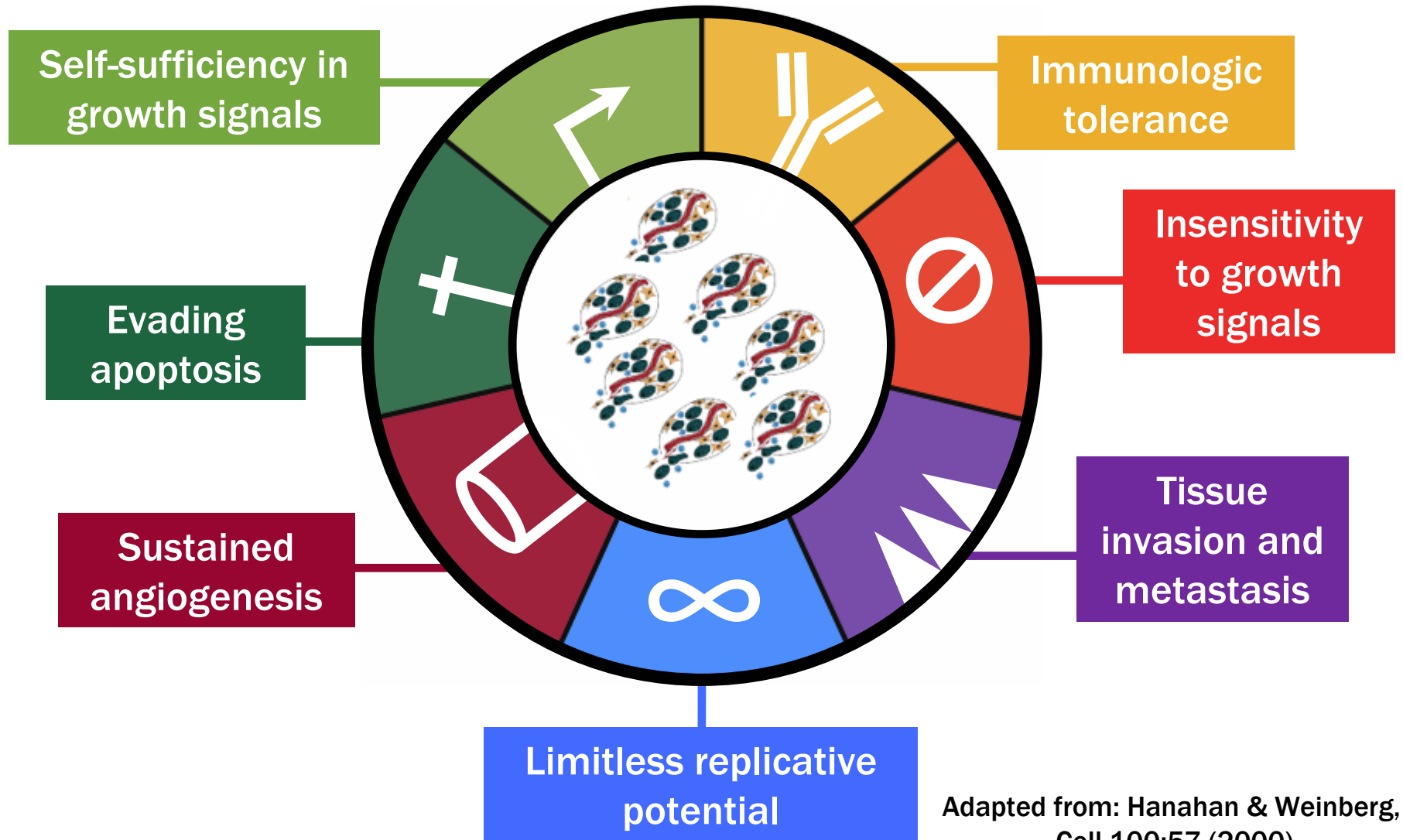
It arises from changes within the DNA of our cells during their lifespan

- Deletions
- Amplifications
- Mutations
- Translocations
- Epigenetic changes



A Complex Foe

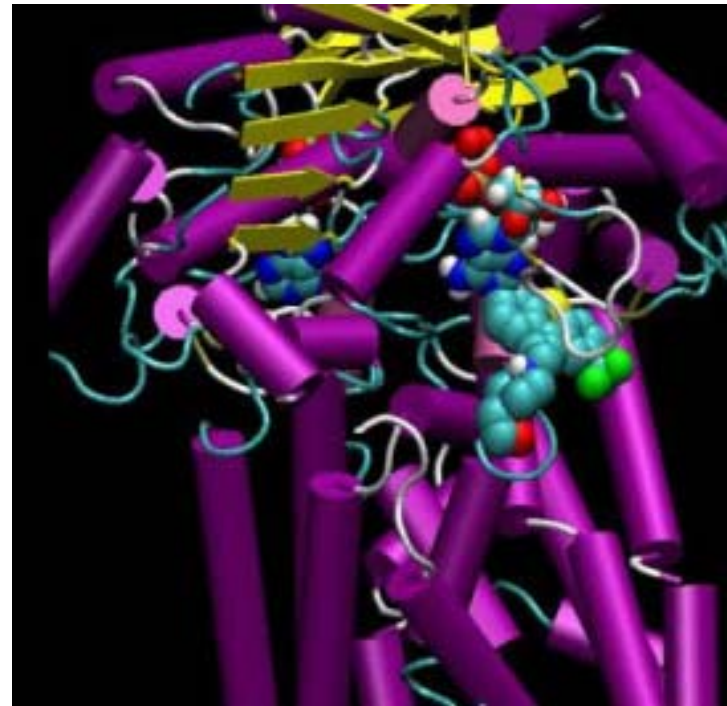
The essential aberrations of cancer



The Complexity of the “Interactome”

Molecular interaction networks

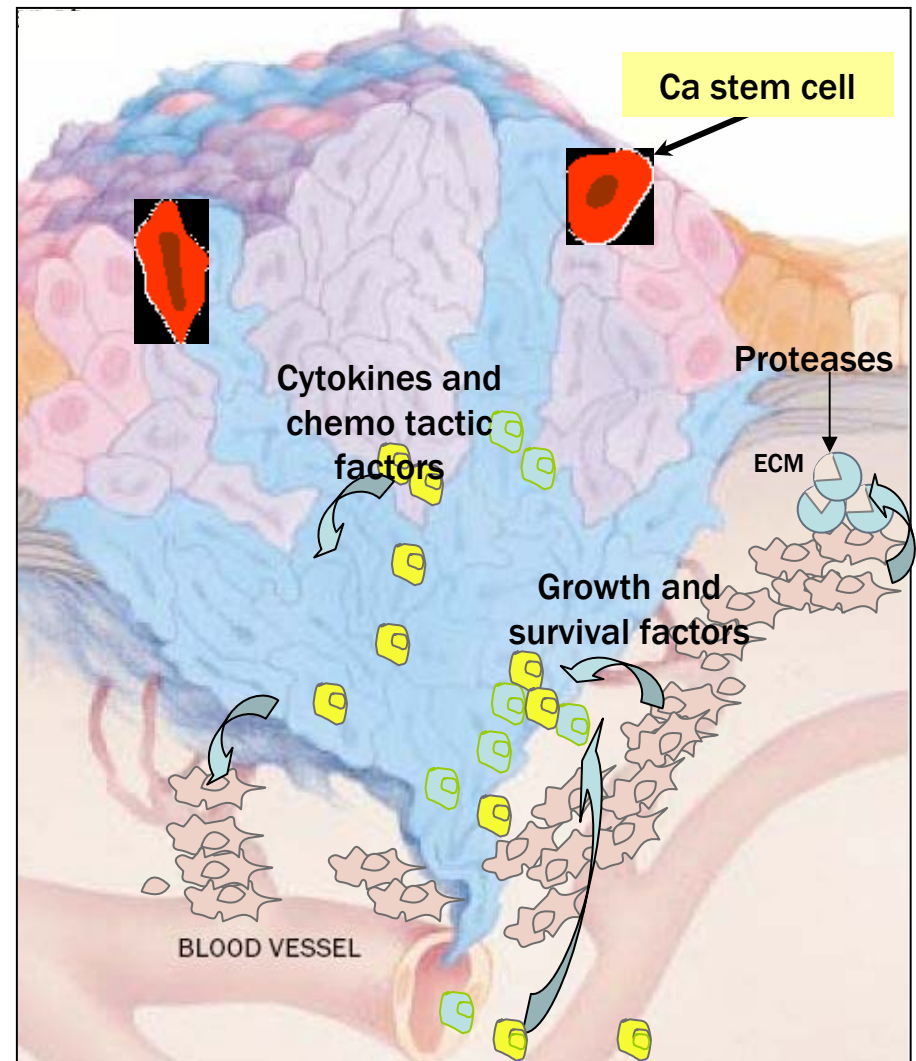
- Protein-protein interactions
- Protein-DNA interactions
- MicroRNA-mRNA interactions



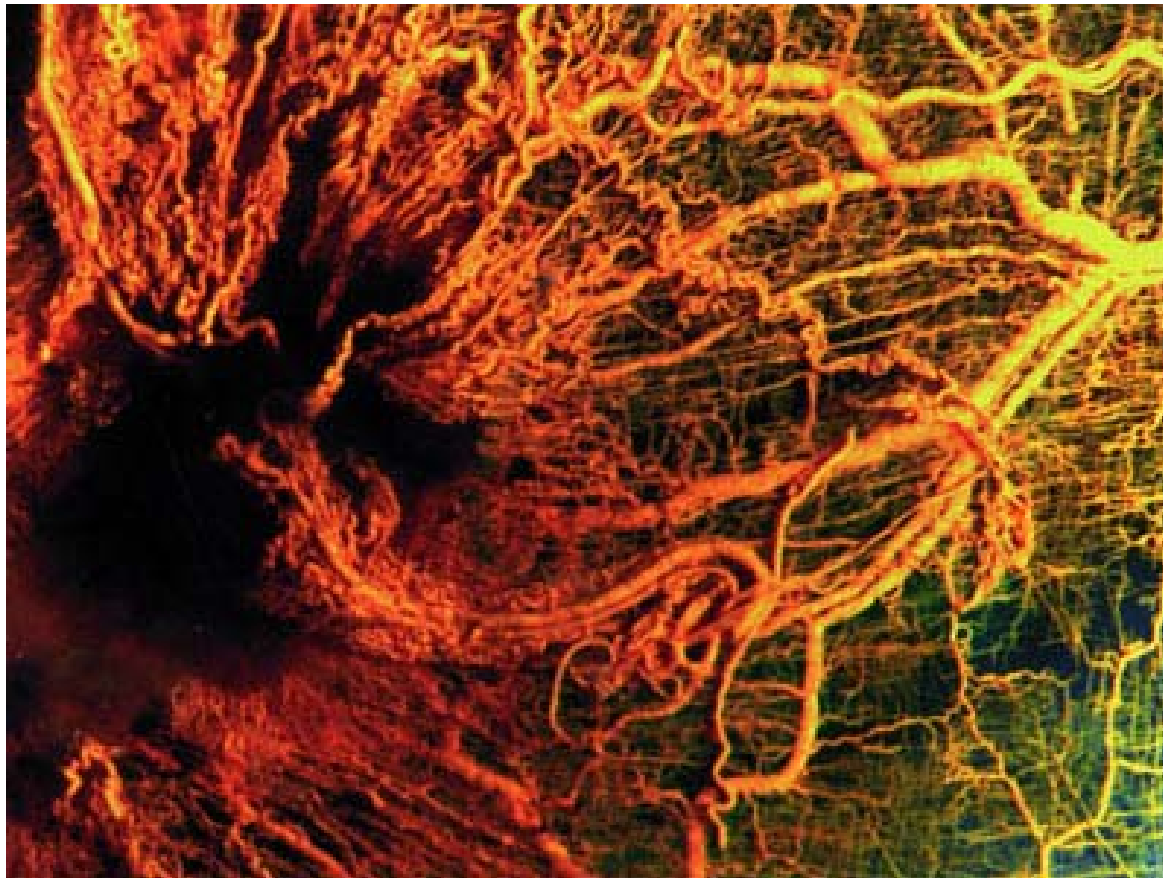
Tumors are “organs” composed of many interdependent cell types

Factors produced in cells within the tumor microenvironment can alter aspects of tumor cell behavior

- Growth factors produced in adjacent cells promote cell proliferation and survival
- Cytokines and chemo tactic factors produced by inflammatory cells and other stroma promote cell migration and invasion
- Proteases produced by the mE break down basement membrane, altering the architecture of tissue structures and migration/invasion of tumor cells



New Blood Vessel Growth Into Tumor

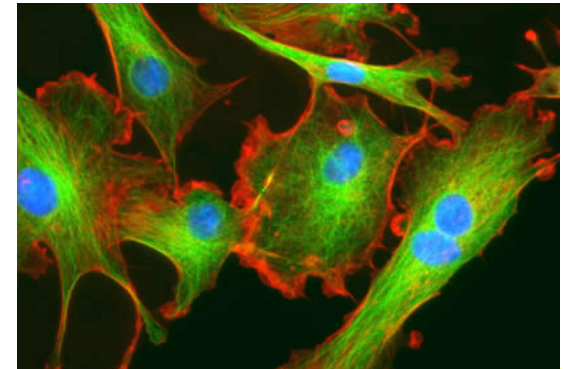


MicroRNA profiling results

Name	P-Value	Average Consistent Fold Change
hsa-miR-503	0.0072	-1.88
hsa-miR-424	0.0015	-1.81
hsa-miR-146a	0.0015	-1.80
hsa-miR-181b	0.0072	-1.80
hsa-miR-542-3p	0.0072	-1.57
hsa-miR-638	0.0072	-1.54
hsa-miR-29b	0.0266	-1.50
hsa-miR-127	0.0266	1.53
hsa-miR-148a	0.0266	1.57
hsa-miR-224	0.0002	1.61
hsa-miR-31	0.0072	3.66

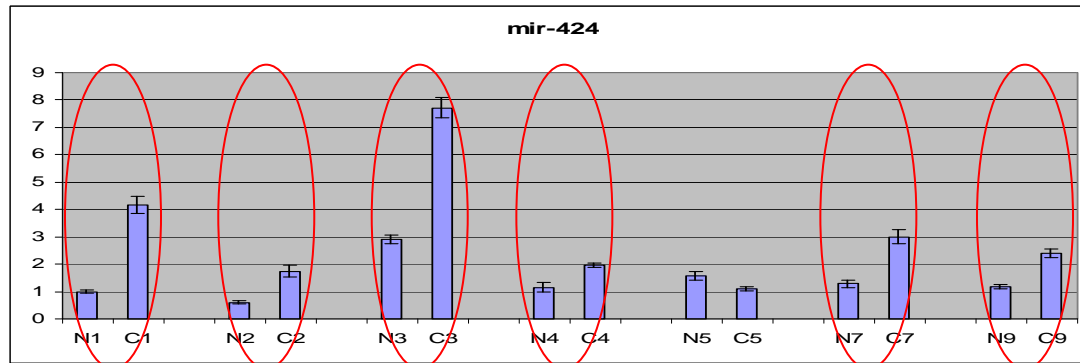
Higher in cancer fibroblasts

Higher in normal fibroblasts

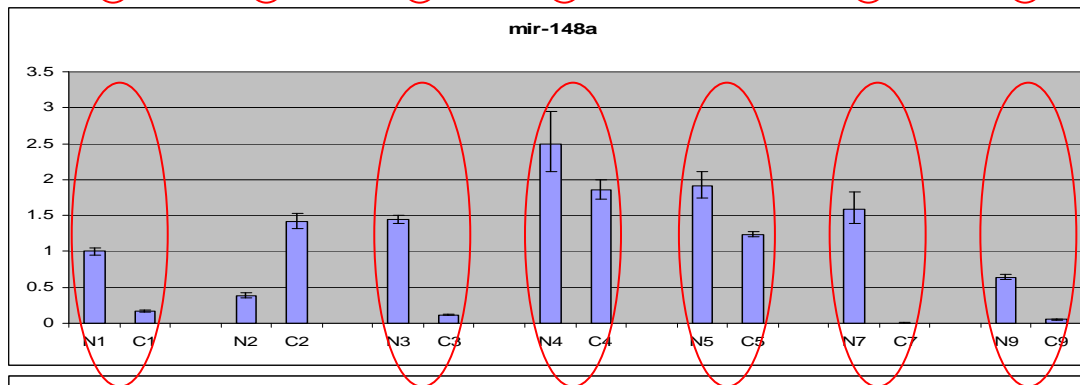


Out of 553 human microRNA spotted on the chip, 11 showed more than a 1.5-fold difference between normal and cancer fibroblasts with the P-value < 0.027

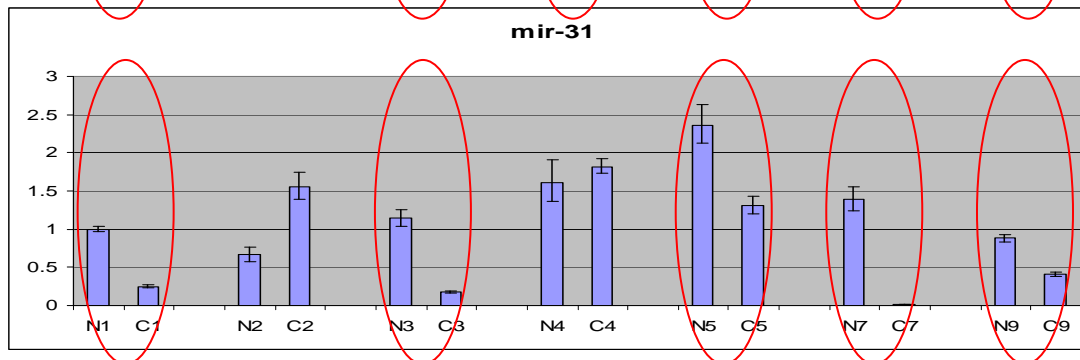
Real-time PCR validation of microRNA profiling data



**Upregulated in
cancer
fibroblasts**



**Down regulated in
cancer
fibroblasts**



**Down regulated in
cancer
fibroblasts**

27 October 2005 | www.nature.com/nature

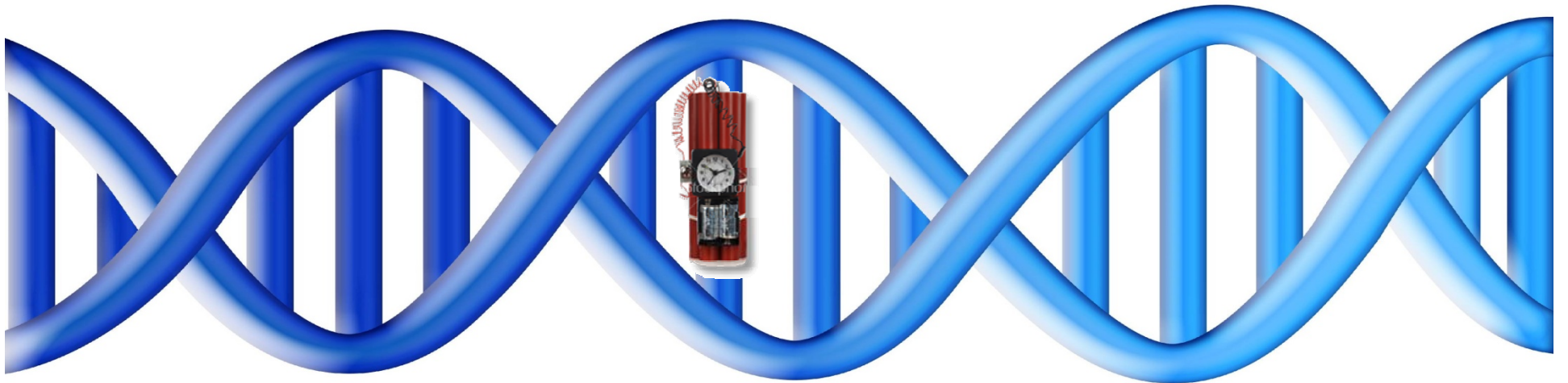
THE INTERNATIONAL WEEKLY JOURNAL OF SCIENCE

nature

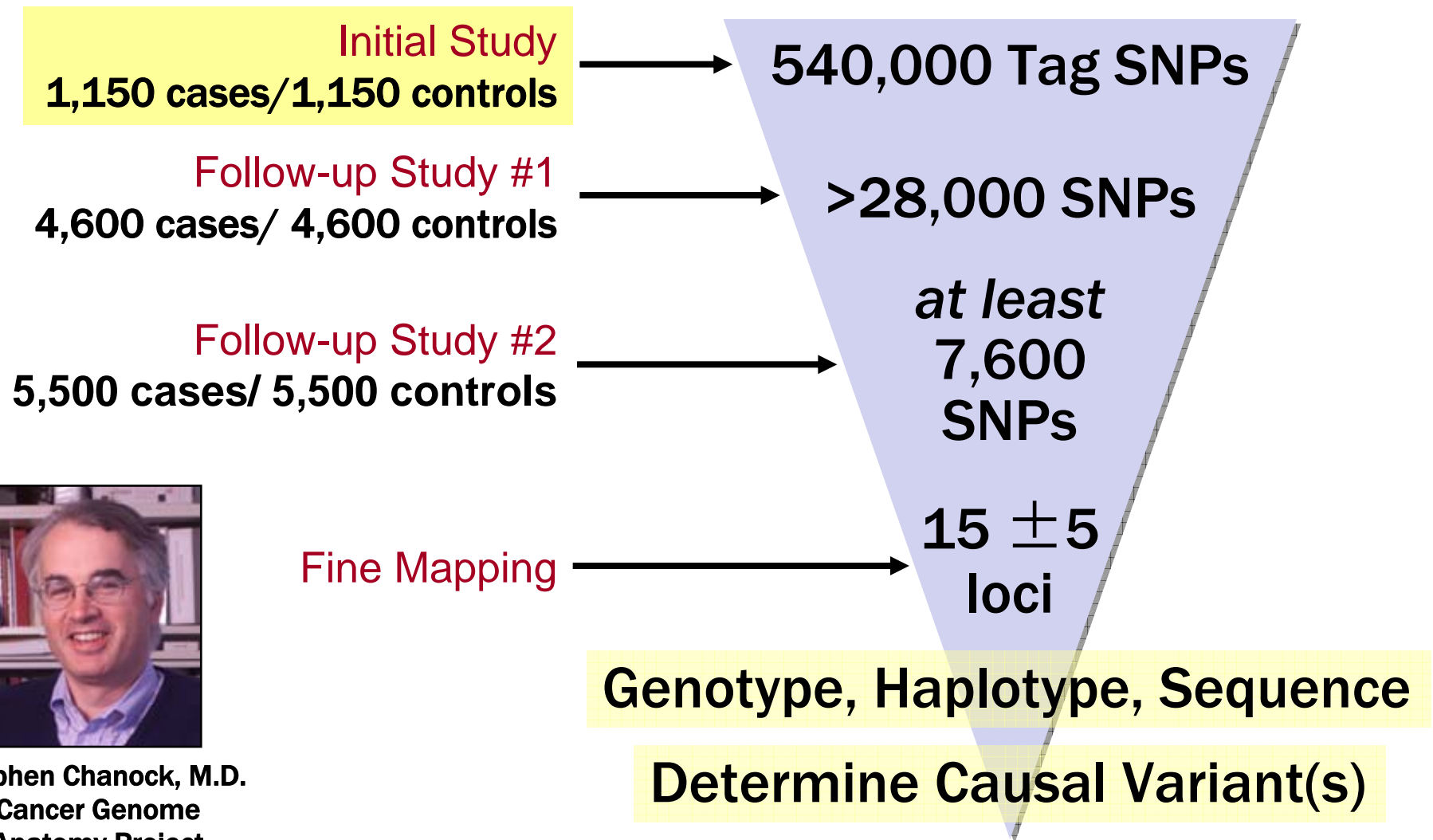


THE HAPMAP PROJECT

Chapter and verse on human genetic variation

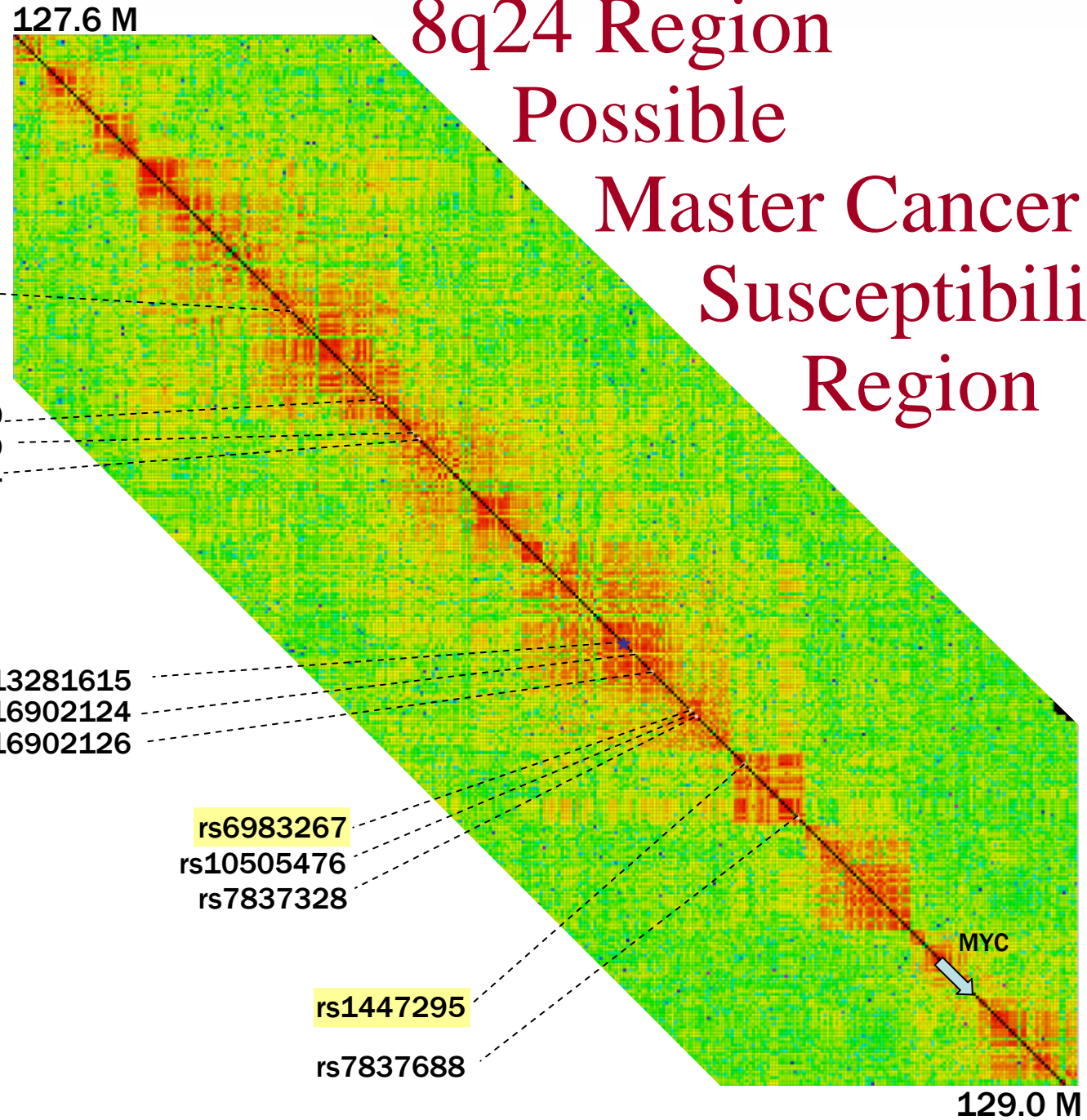


Identifying Genetic Markers for Prostate & Breast Cancer



Stephen Chanock, M.D.
Cancer Genome
Anatomy Project

8q24 Region Possible Master Cancer Susceptibility Region



Region 2
Prostate
only

Breast only

Region 3
Prostate & Colon

Region 1
Prostate only

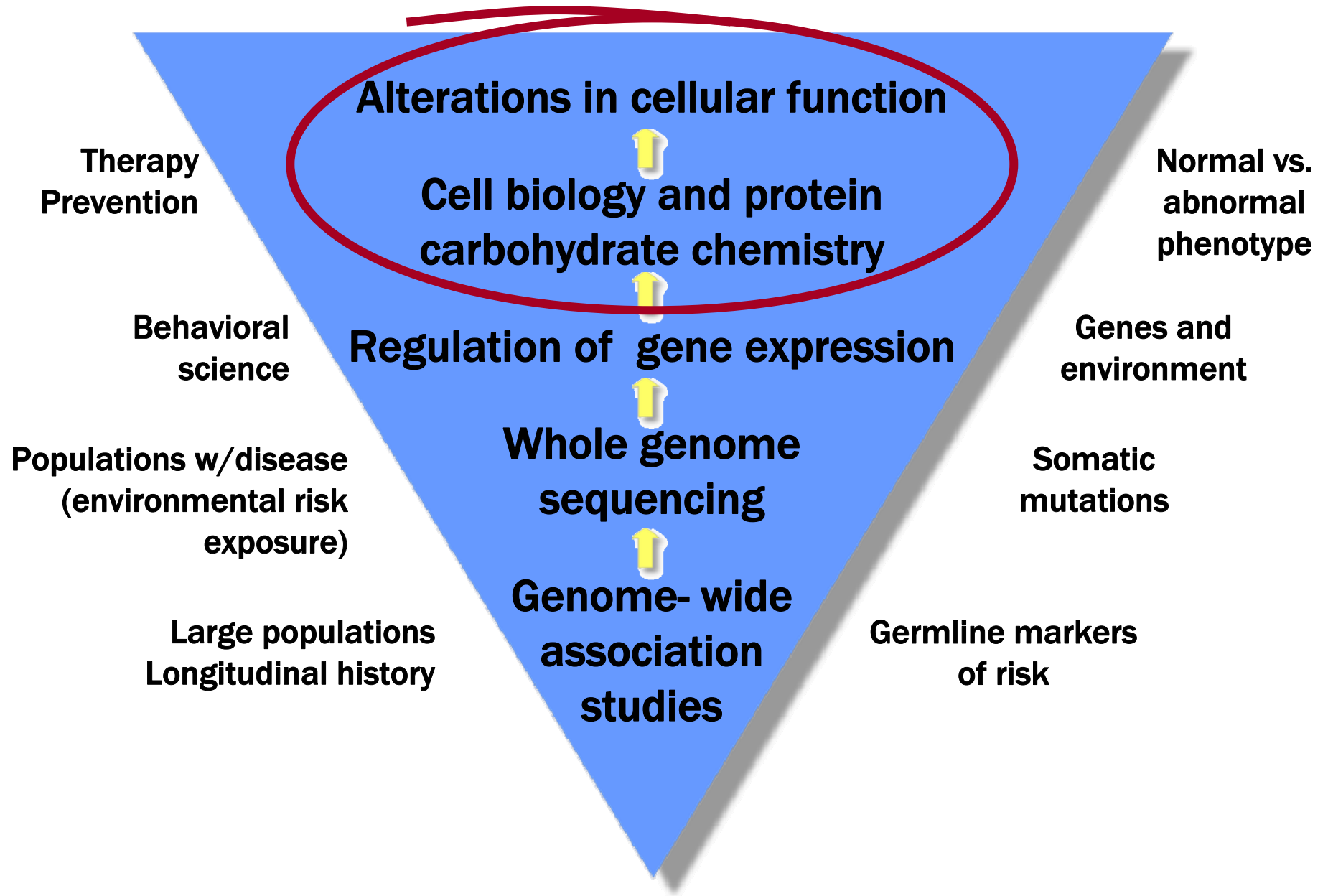
129.0 M

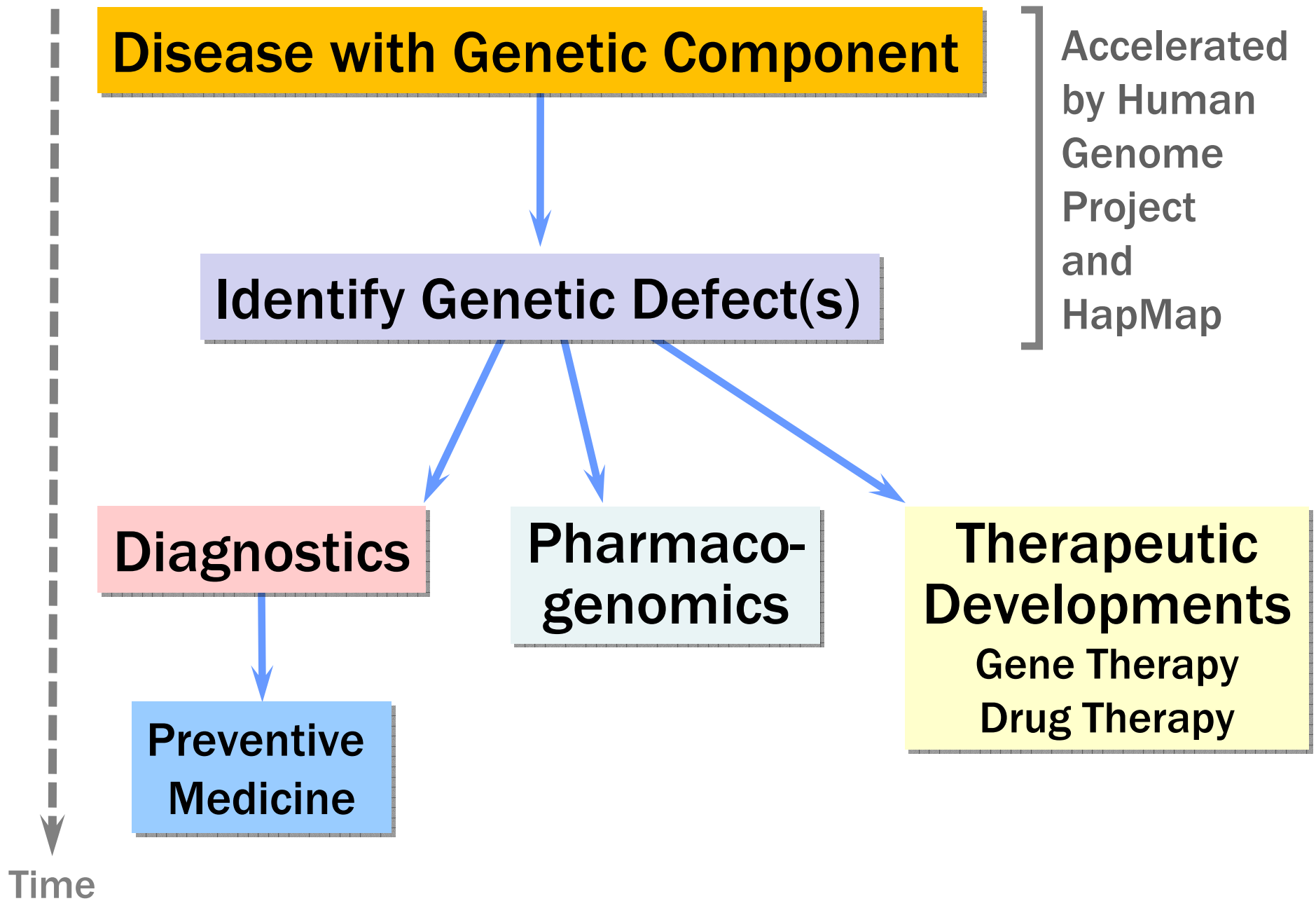
The Cancer Genome Atlas

- **Pilot Project of NCI-NHGRI**
 - Cancers: Brain, Lung and Ovarian
- Components
 - Human Cancer Biospecimen Core Resource
 - Genome Sequencing Centers
 - Targeting sequence in phase 1
 - Cancer Genome Characterization Centers
 - Data Management, Bioinformatics and Computational Analysis
- Data Access-Registered



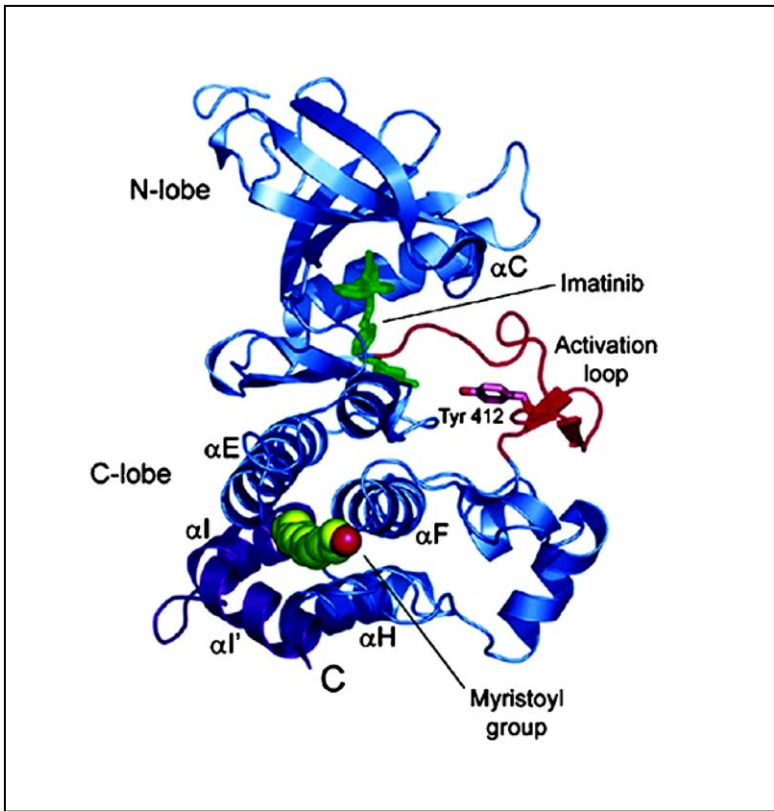
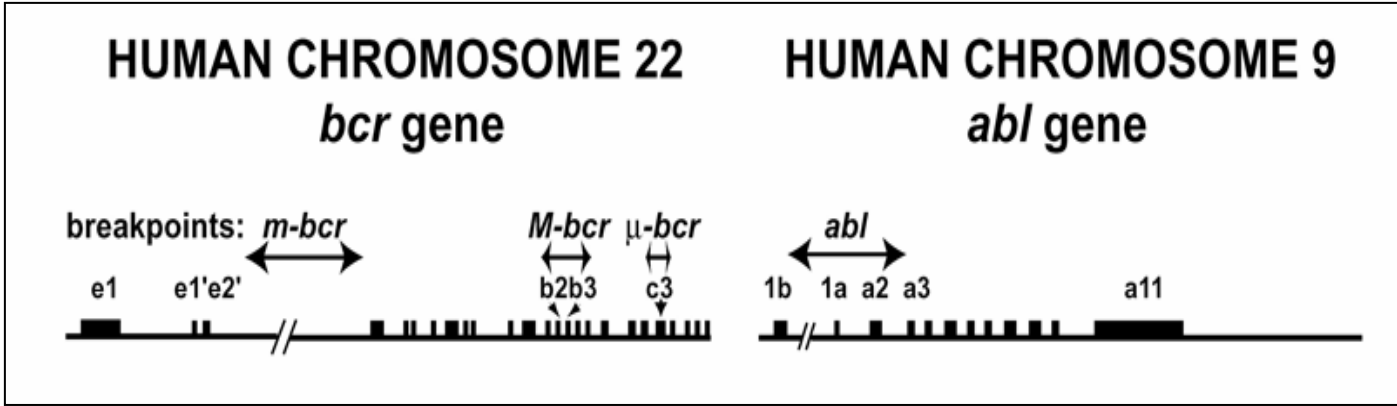
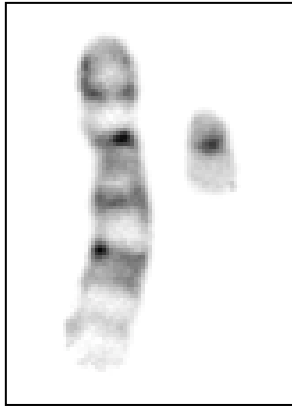
Functional Genomics



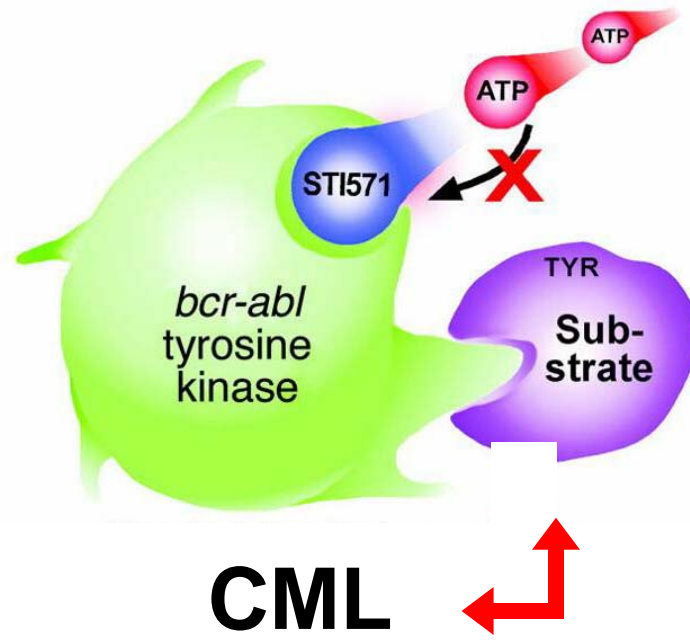


Courtesy Francis Collins, NHGRI

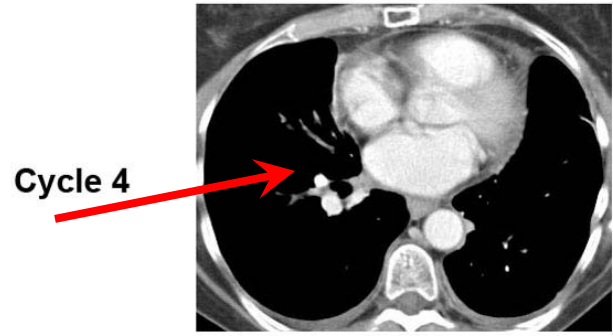
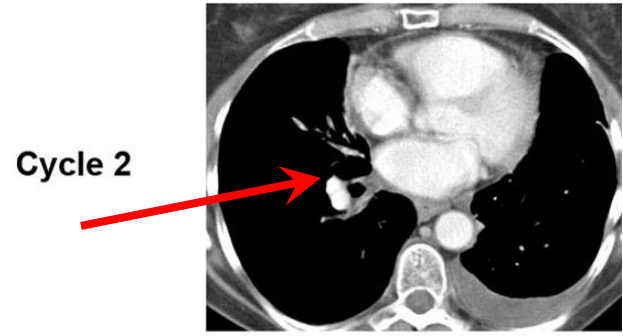
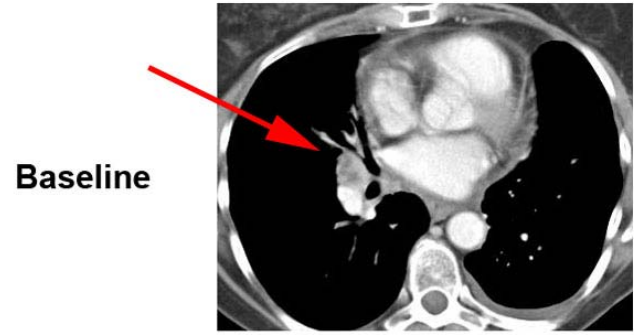
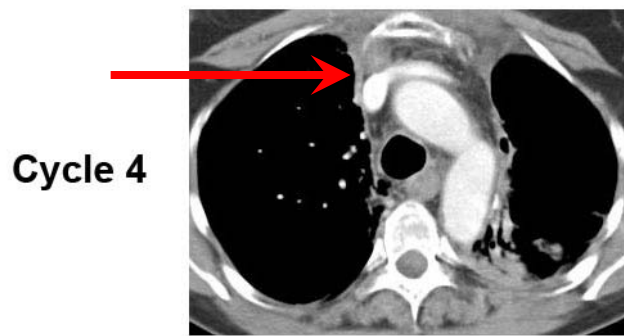
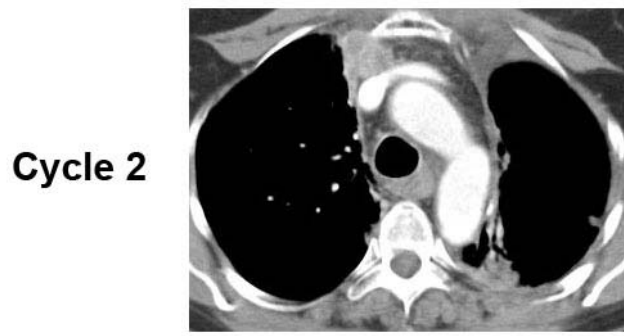
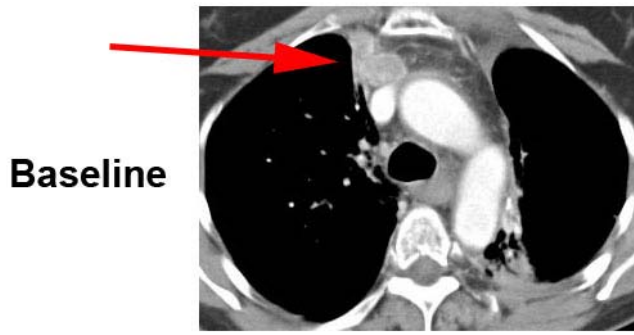
CML t(9;22)



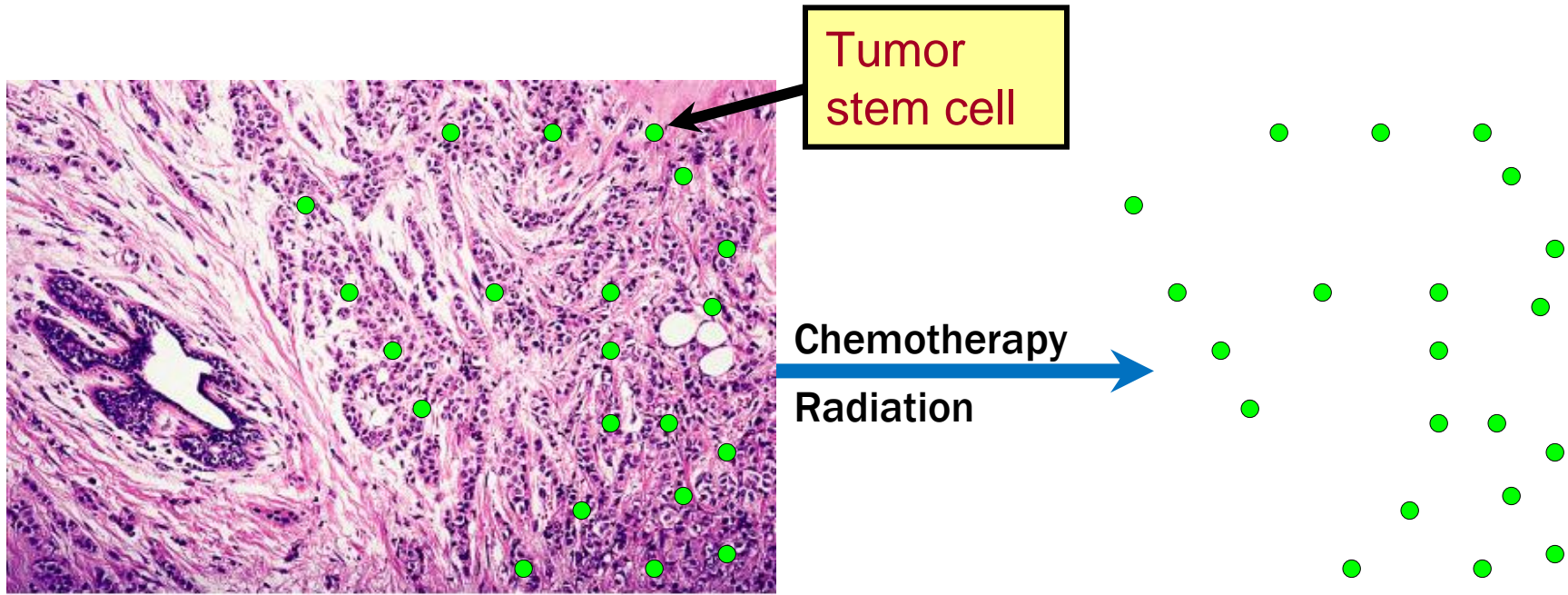
Imatinib (Gleevec)



Response in Patient with Metastatic NSCLC to Epigenetic Therapy



Cancer Stem Cells



Tumor

“Remission”

Cancer Stem Cells

- Have ability to travel to other tissues; do not need to acquire this characteristic
- Drug resistance not acquired but present
 - High levels of ABC (ATP Binding Cassette) transporters actively efflux drugs (ABCG2, ABCB1, ABCD1)
- Molecular pathways uniquely present in HSCs and CSCs (Notch, Hedgehog (Hh), Wnt, Bmi-1)

Breast Cancer Stem Cells

	2000*	500*	200*	100*
Stem-like cells (1%)				
CD44⁺ CD24⁻ ESA⁺	10/10	4/4	4/4	1/6
Non stem cells (99%)				
CD44⁺ CD24⁻ ESA⁻	0/10	0/4	0/4	0/6

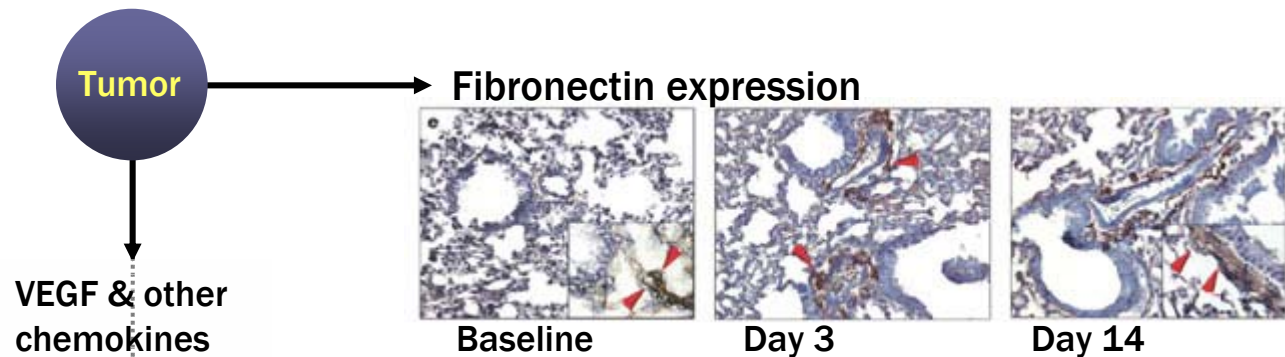
* Number of cells injected into mammary fat pads in NOD/SCID mice.

The stem cell-like cells out of a breast tumor are tumorigenic; the rest of the cells are not.

- What role do **tissue stem cells** play in the process of **metastasis**?
- What is the role of **cancer stem cells/tumor initiator cells** in the development of a **metastatic lesion**?

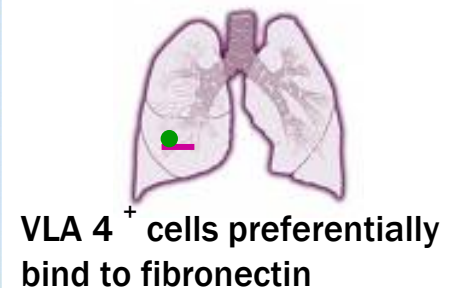
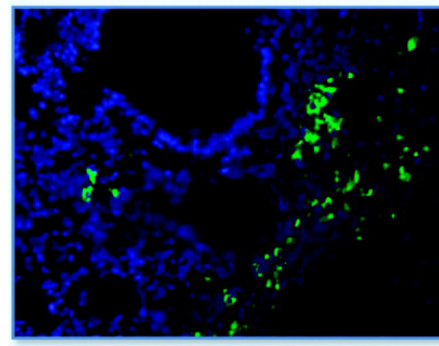
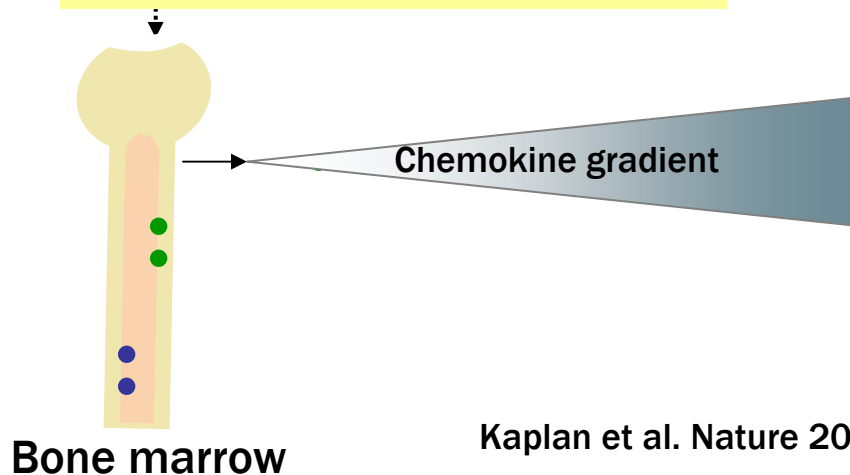
Chemokines & Role in the Premetastatic Niche

1. Fibronectin recruitment



- VLA 4⁺ VEGF R1⁺ HPCs
- VEGF R2⁺ EPCs
- Fibronectin fibers

2. Premetastatic niche

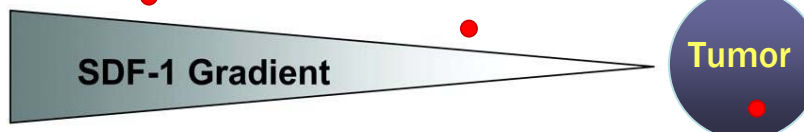
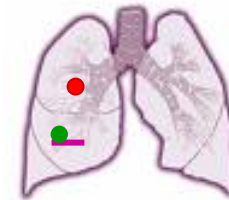
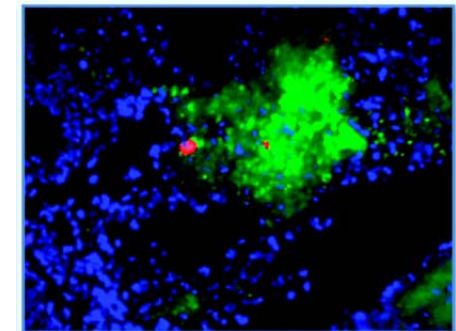
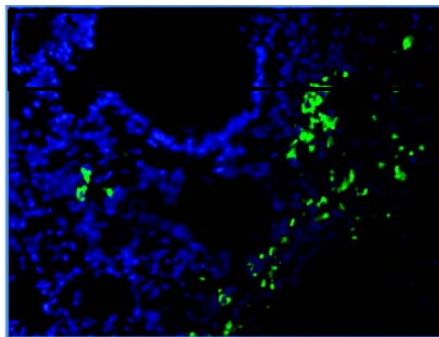


Kaplan et al. Nature 2005;438(8): 820 -7

Kaplan et al. Cancer Res 2006; 66(23):11089-93

Possible Mechanism for Recruitment of Tumor Stem Cells to Premetastatic Niche

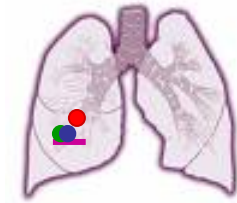
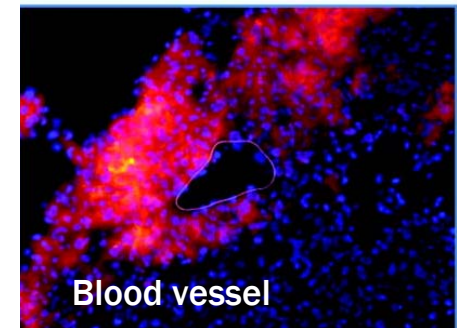
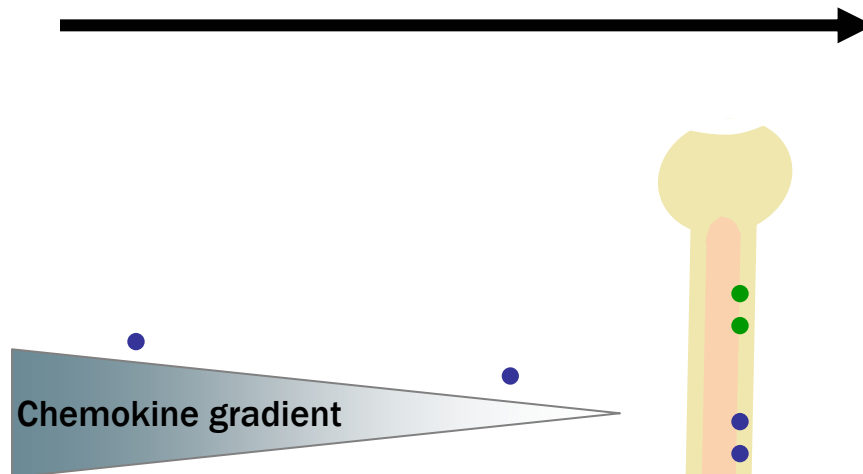
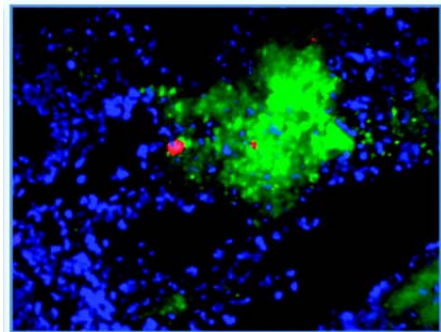
Premetastatic niche



- VLA 4⁺ VEGF R1⁺ HPCs
- VEGF R2⁺ EPCs
- Fibronectin fibers
- CRX4 R⁺ tumor cell

Kaplan et al. Cancer Res 2006;
66(23):11089-93

Vascular Organization of Metastasis



Bone marrow

- VLA 4⁺ VEGF R1⁺ HPCs
- VEGF R2⁺ EPCs
- Fibronectin fibers
- CRX4 R⁺ tumor cell

Kaplan et al. Cancer Res 2006;
66(23):11089-93

Neutrophils in a Chemoattractant (fMLP) Gradient

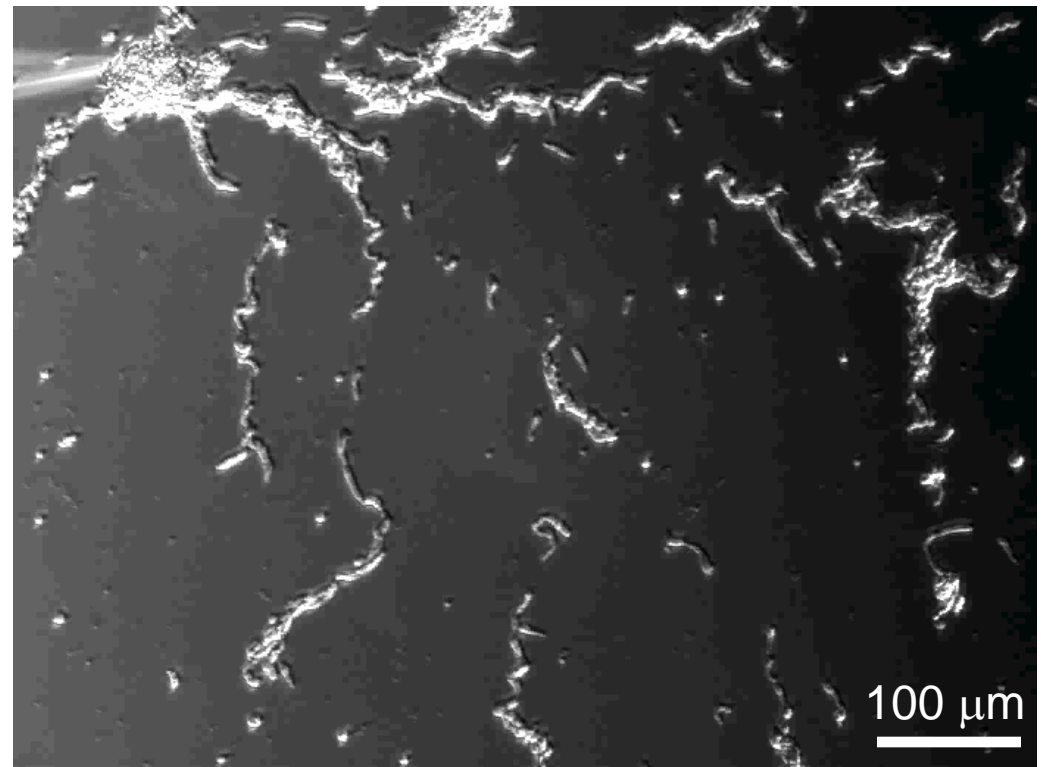


**Carole A. Parent,
Ph.D.**

One frame every 10 sec.

Dictyostelium cells Migrating in a Chemical Gradient

- Cells migrate in a chemical gradient to form aggregates
- Single-cell migration is heavily studied and well characterized
- In contrast, group cell dynamics are not well understood

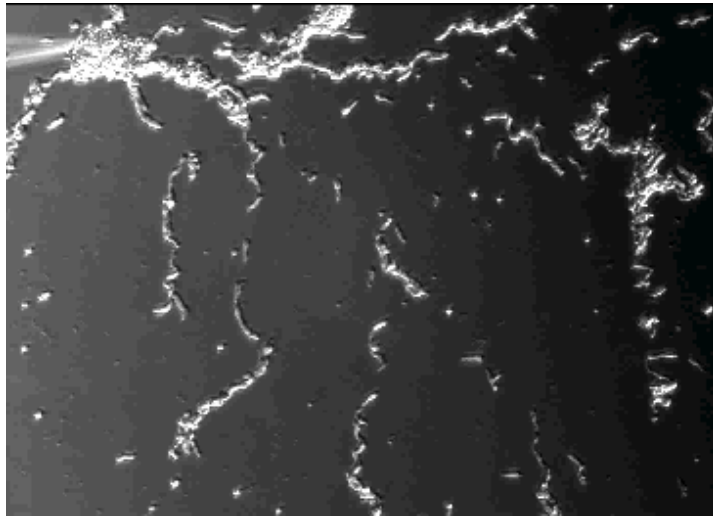


Dictyostelium cells migrating towards a signal-releasing pipette (upper left) over the course of 2 hrs

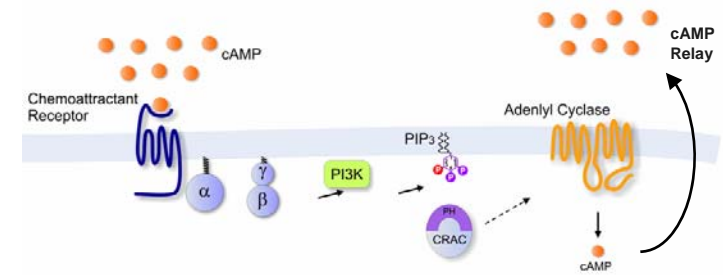
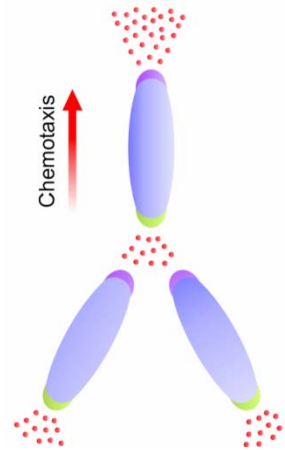
Group Cell Migration

- Group cell migration is crucial to many cellular processes, including metastasis
 - How do cells migrate in groups?
 - How do cells behave inside and outside groups?
 - Which cells migrate together?
- Physical analyses and techniques can help provide answers and – when coupled to cell biological, genetic, and biochemical analyses – they can give rise to a global understanding of the molecular mechanisms at play during group migration

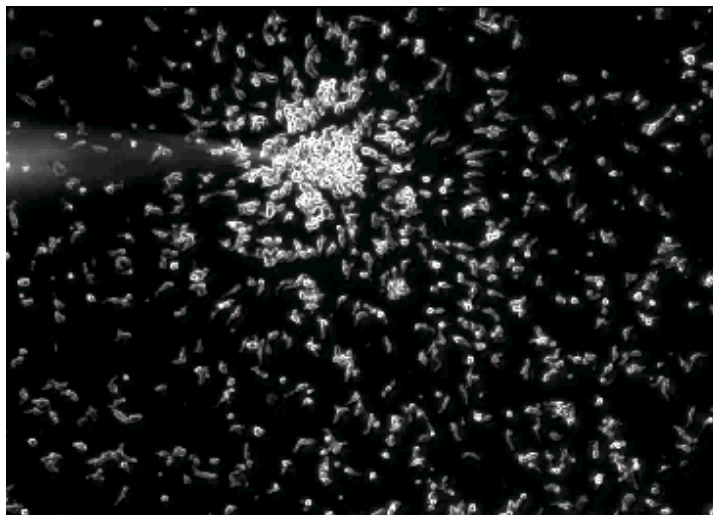
How do cells migrate in groups?



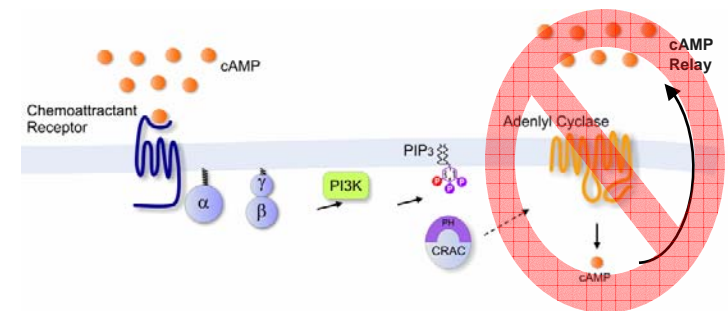
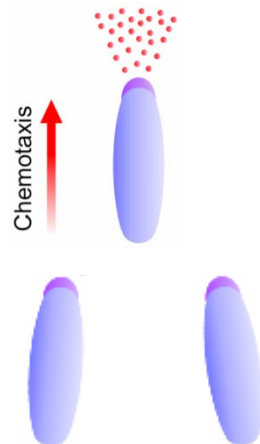
Chemical Signal



Dictyostelium cells migrating towards a signal-releasing pipette form groups by **relaying chemical signals**

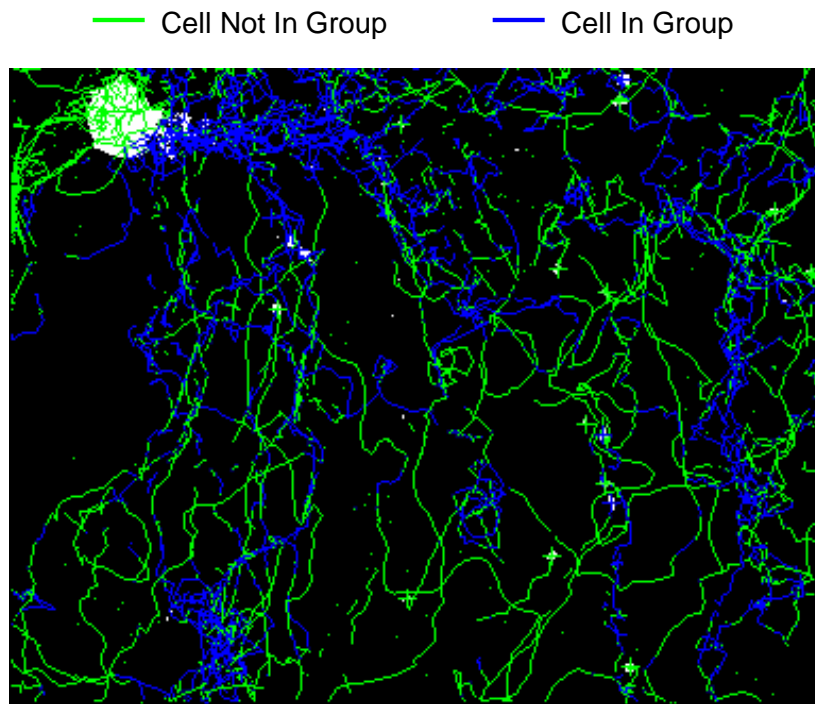


Chemical Signal

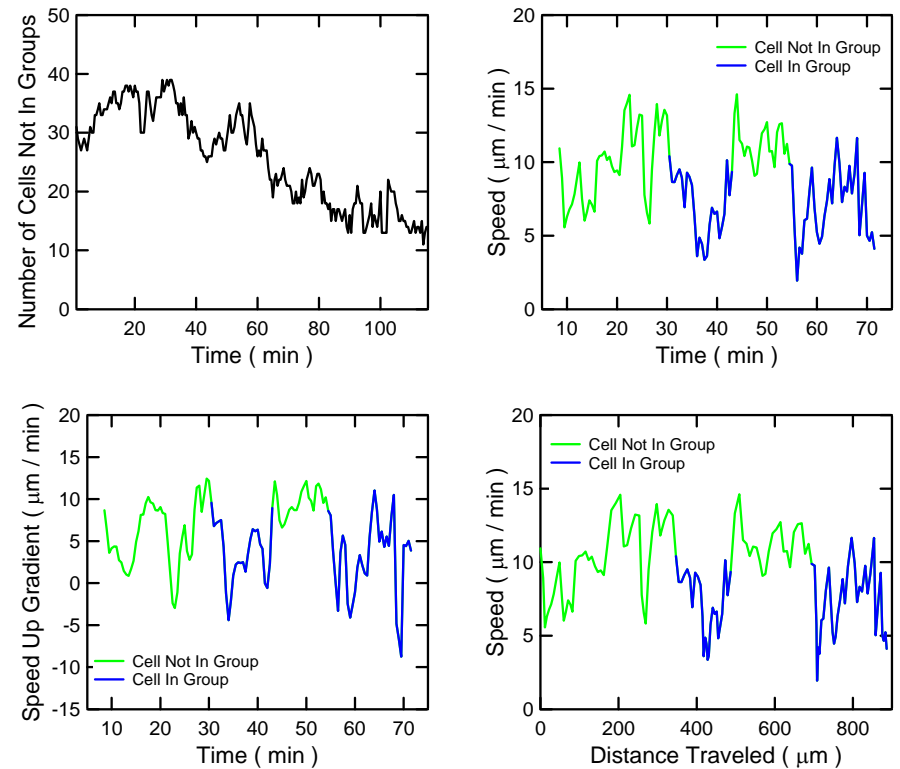


Dictyostelium cells **defective in signal relay** do not form groups and instead migrate independently

Computational Tools for Statistical Analysis of Complex Biological Behaviors



Automated software tracks the migration of cells from the previous movie

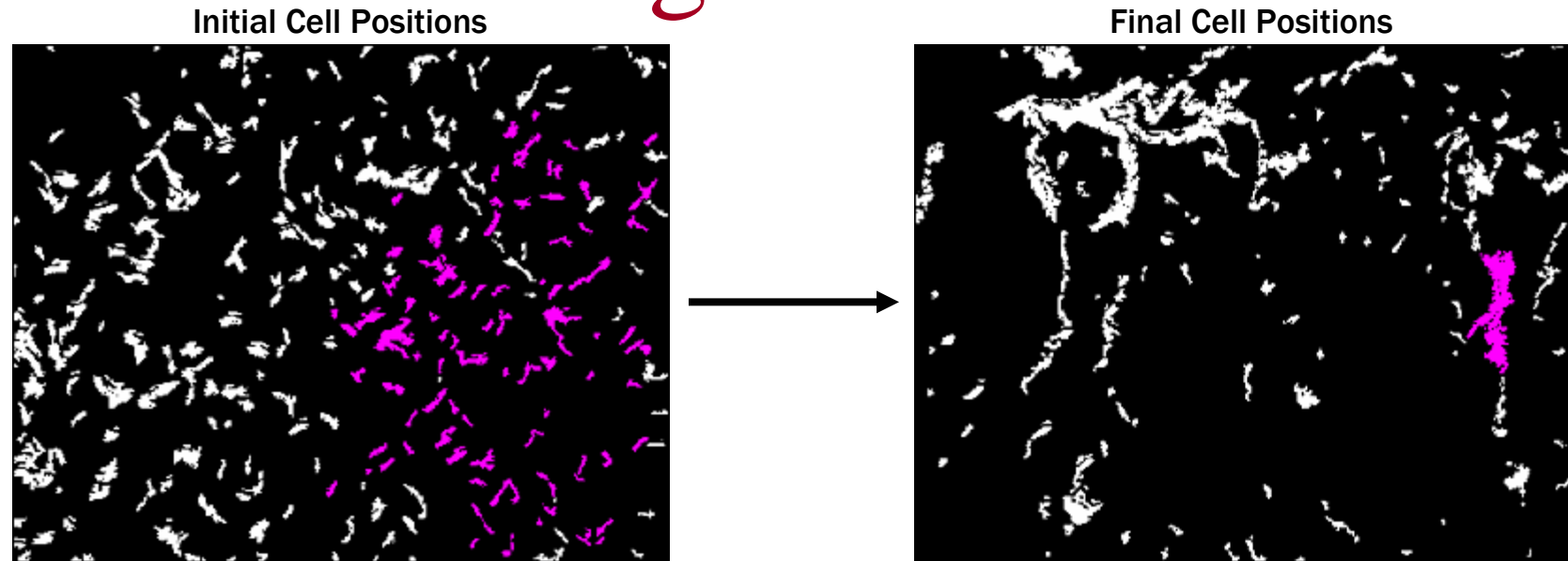


Various statistics for a representative cell from the movie on the left

How are cellular aggregates formed?

- Tracking software analyzes image stacks and provides statistics, such as velocity, velocity correlations, directionality, and shape
- By comparing the motility characteristics of wild type cells to mutants cell lines that have specific signal transduction defects, we can identify key components affecting group behavior and build pathways that control group cell migration

Which cells decide to migrate together?



Dictyostelium cells at 0 and 90 minutes. In the right panel a group of cells is highlighted in pink. In the left panel cells that entered this group are highlighted in the same color.

Software automatically determines which cells join which communities and allows quantitative analyses with respect to cell density, strength of chemical gradient, and mutant behavior

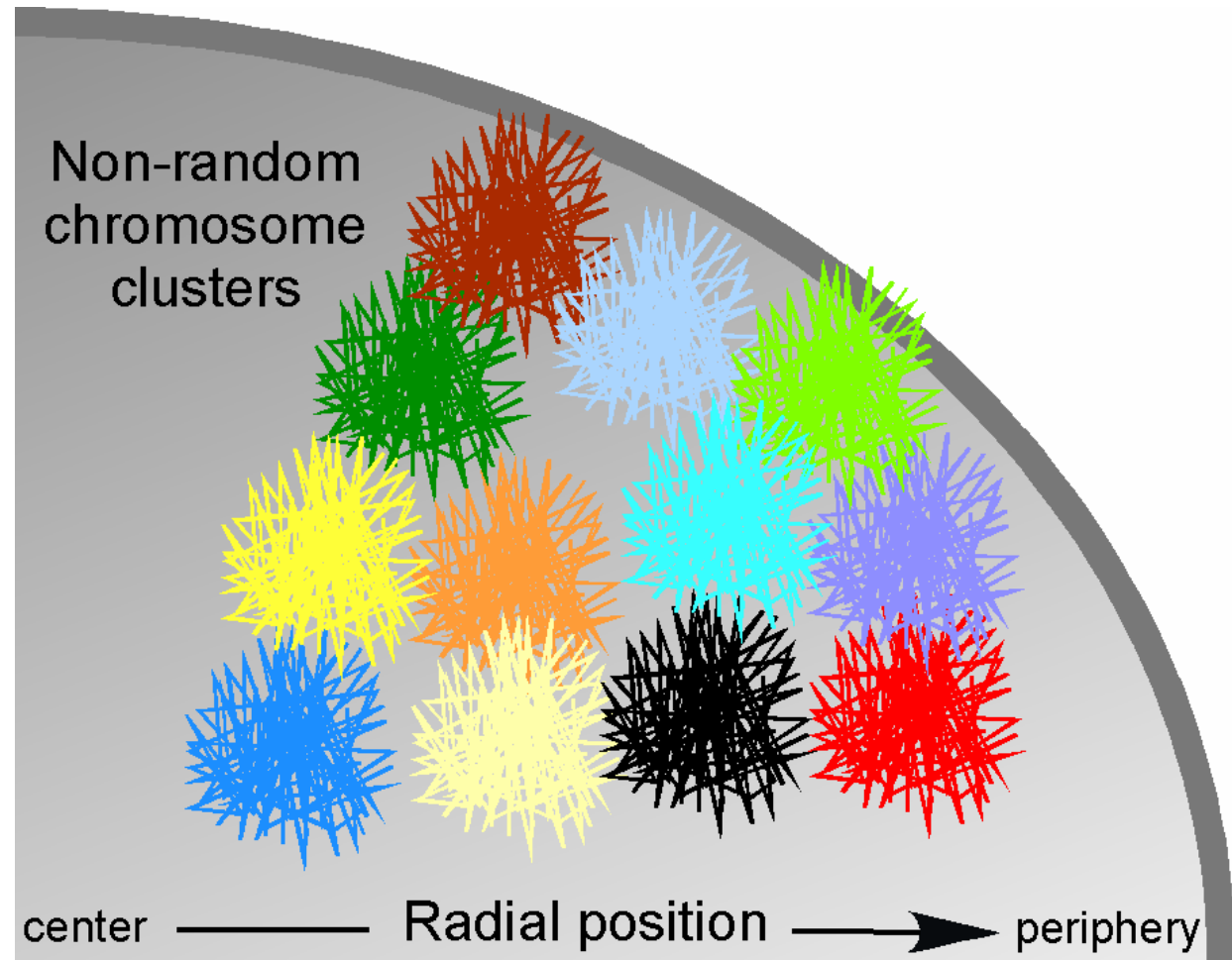
Organization of the Genome in 3D Space



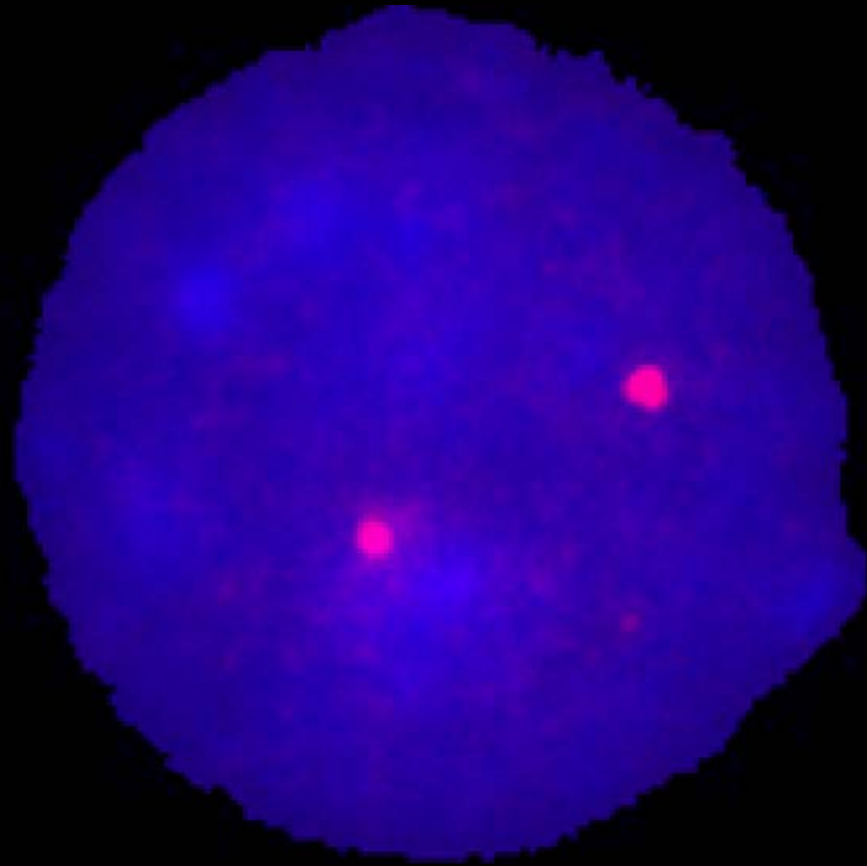
Tom Misteli,
Ph.D., NCI



Karen Meaburn,
Ph.D., NCI



Diagnostic applications

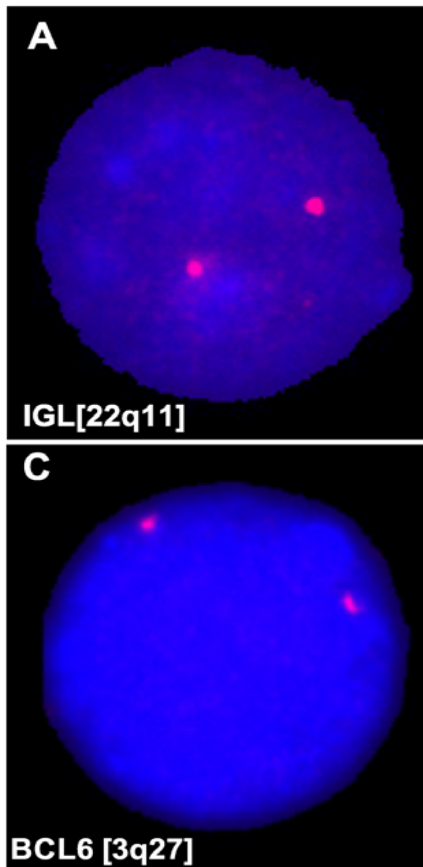


Normal



Breast tumor

Non-Random Genome Organization



Non-random chromosome position

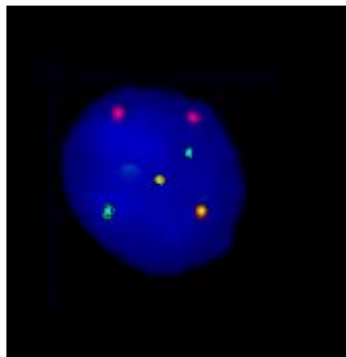
Non-random gene position

- **Cell-type**
- **Tissue-specific**
- **Evolutionarily conserved**
- **Differentiation**
- **Development**

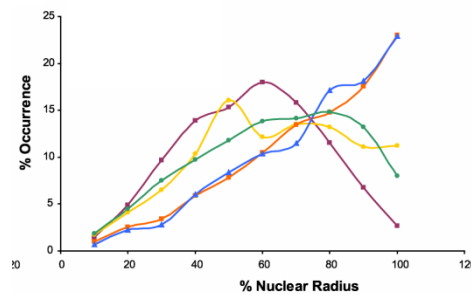
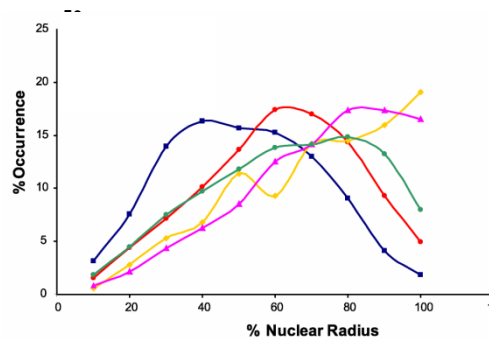
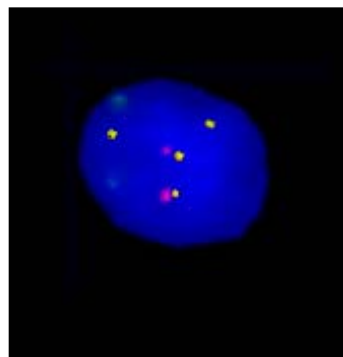
Spatial positioning as a potential diagnostic tool

Interphase Cytogenetics

normal

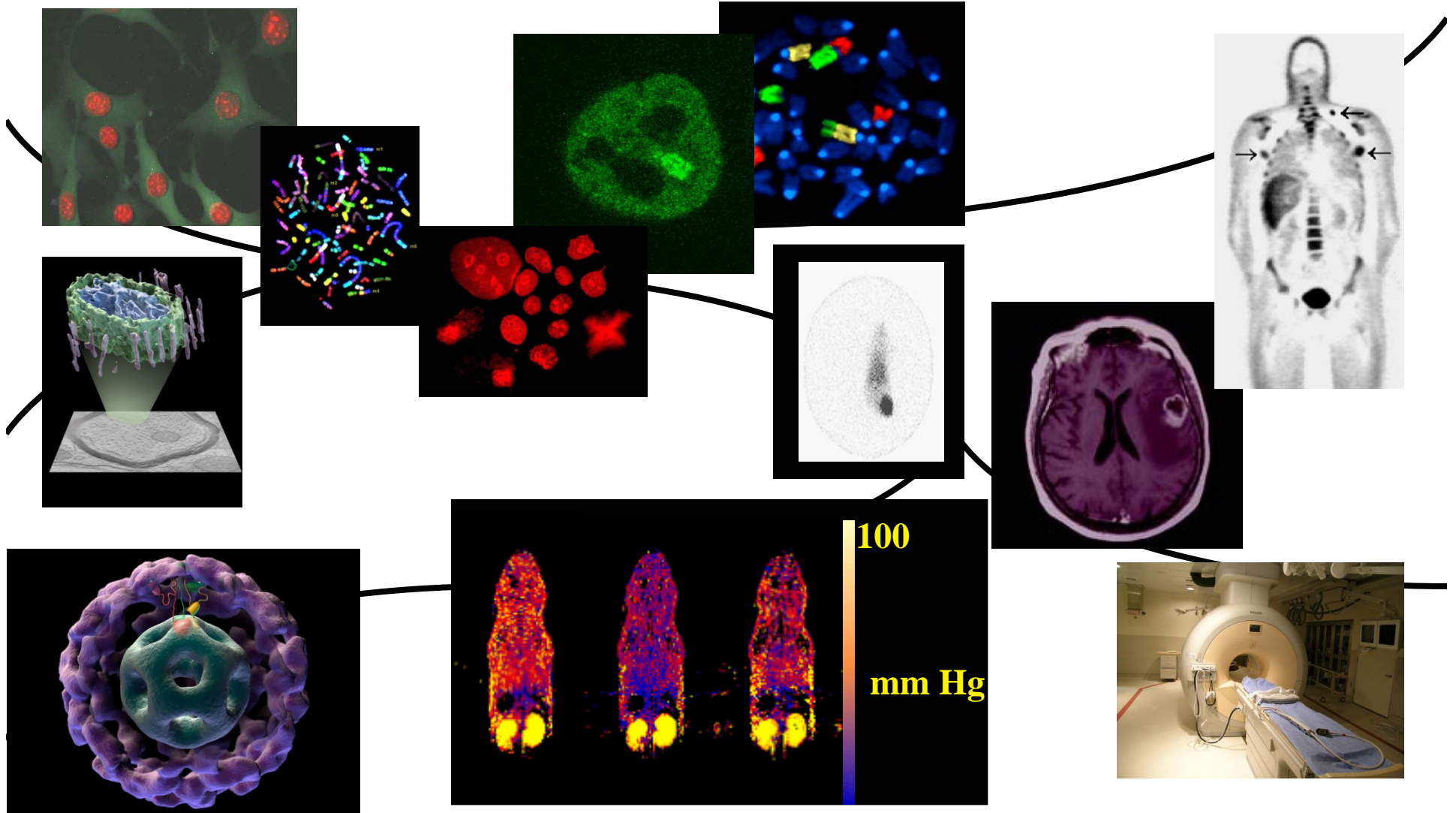


(pre-) malignant

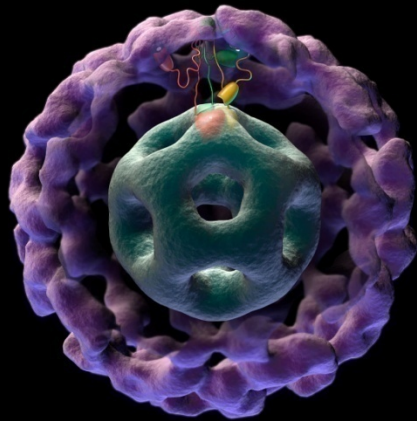


- Normal vs. malignant
- Metastatic cells
- Pre-malignant
- Metastatic potential
- Tumor type

From Molecules to Man

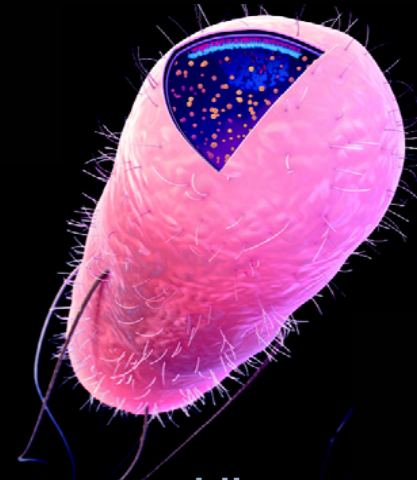


Bridging the imaging gap



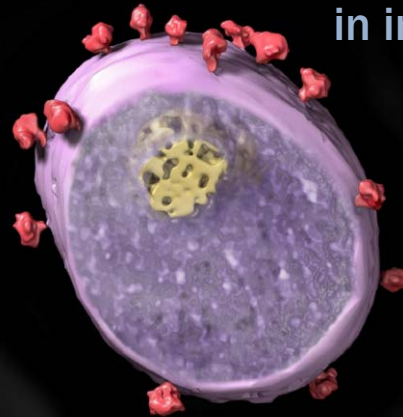
~50 nm

Dynamic multiprotein complexes



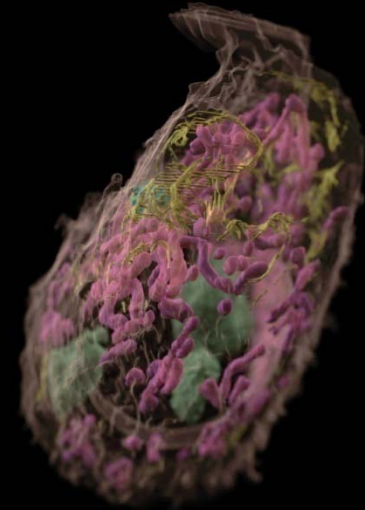
~2000 nm

Signaling assemblies in intact cells



~150 nm

HIV structure and cell entry mechanisms



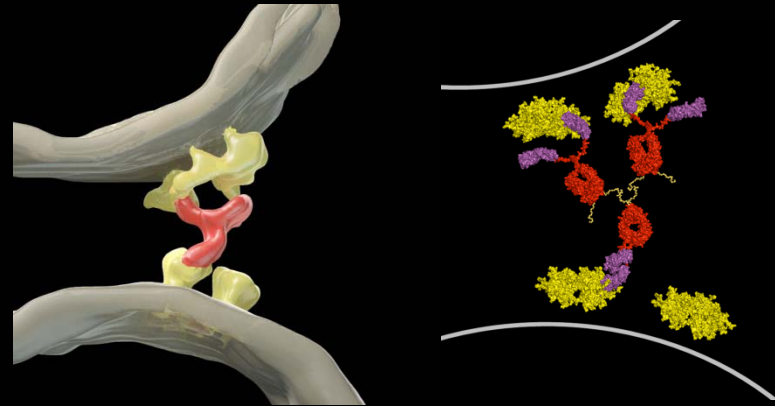
~30,000 nm

Subcellular architecture of melanoma cells



Sriram Subramaniam, Ph.D.

Molecular footprint of AIDS virus neutralization



Cryo-electron tomography of complex between viral gp120 and D1D2-IgP, a potent neutralizing antibody

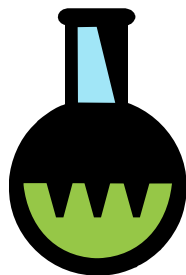
Conceptual approaches to complex problems of biology

Increasing the dimensions of biological science

- Protein-protein interactions
- Chemical gradients in vivo
- Energy and time as critical dimensions of interactions at the target
- Imaging the target reaction

Cancer Modeling Spectrum

All models are problematic but some are more useful than others.



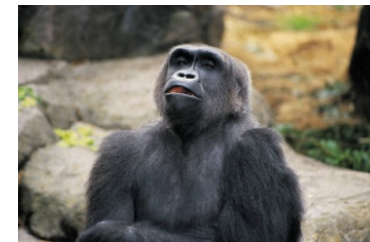
In vitro



Cell Lines

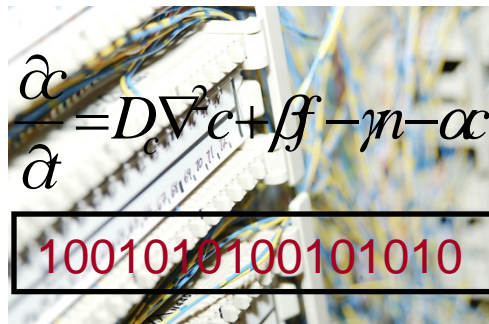


Mammalian



Primates

- Simple
- Adaptable
- Modular
- Personal
- Economical



Mathematical/Computational



Thoughts in Conclusion

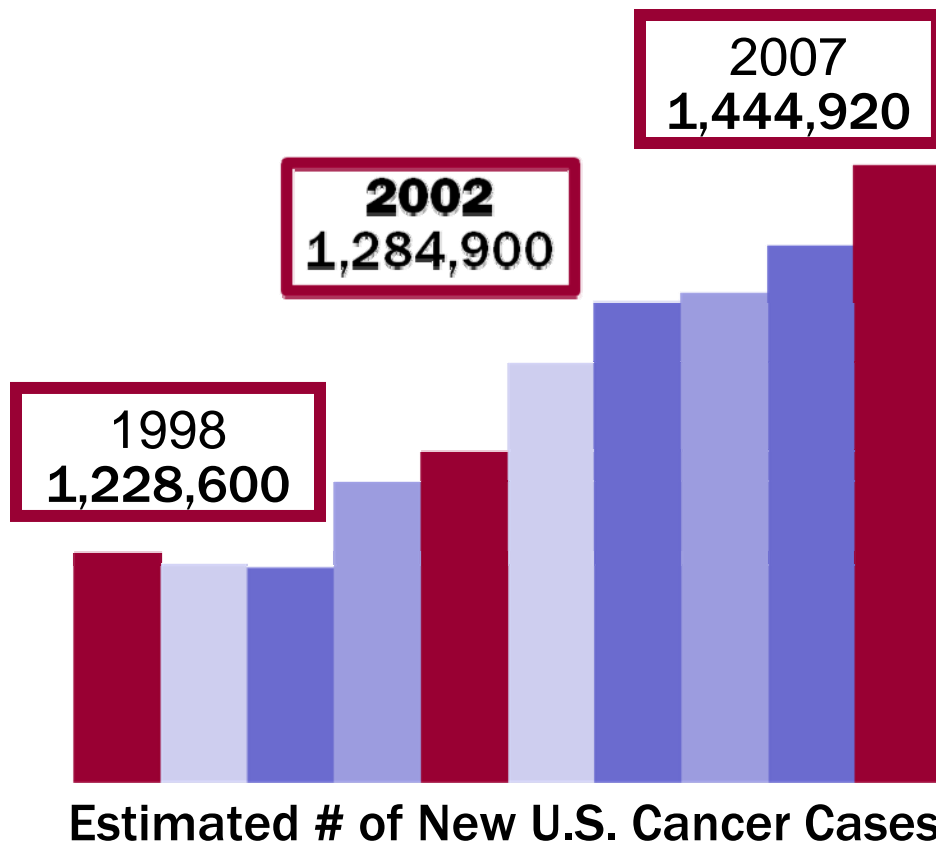
- **Never a more exciting time to work in science**
- **Rapidly advancing technology that drives complex research**
- **More and more a team approach**



www.cancer.gov

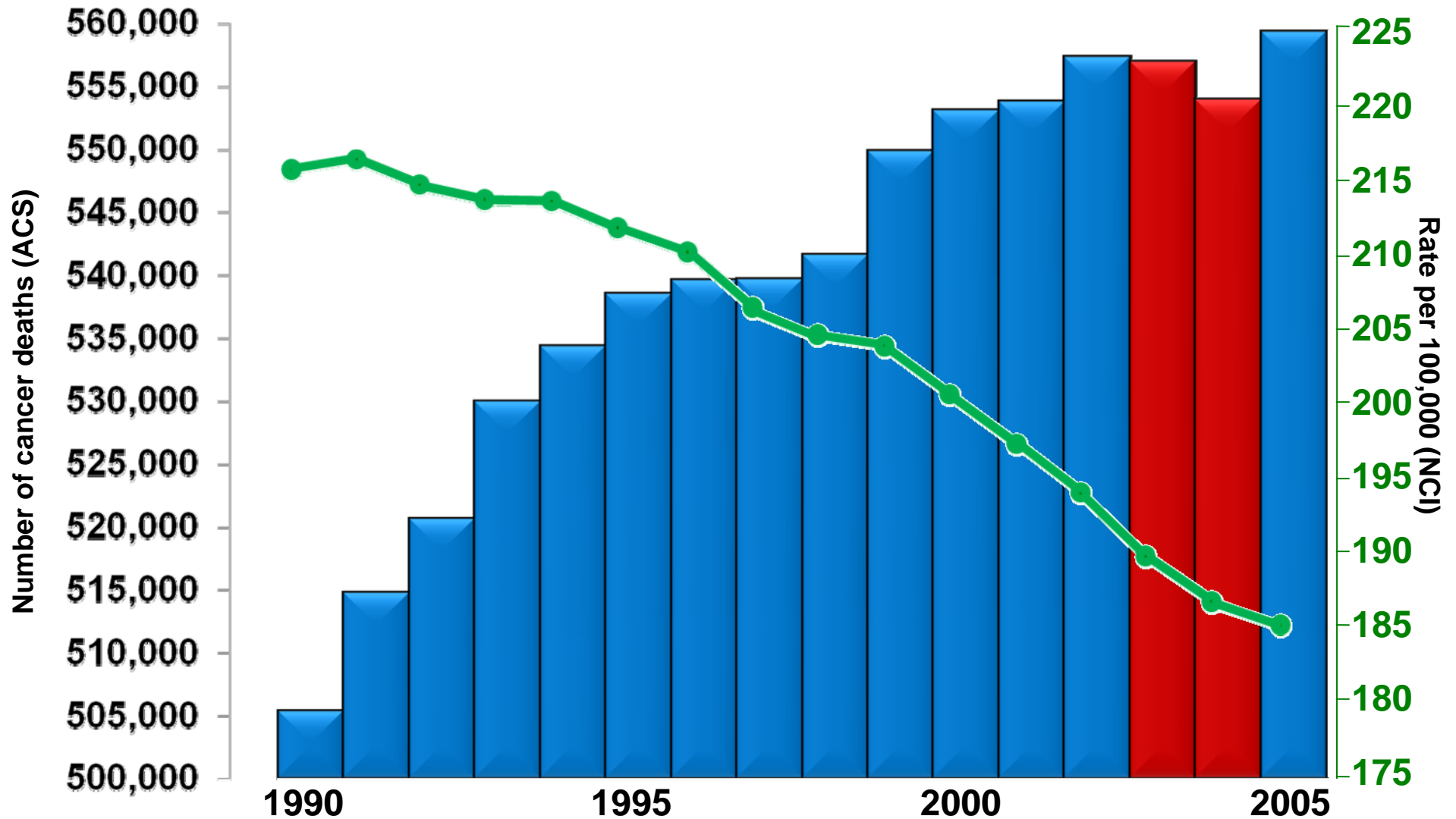
Why What We Are Doing is So Important

Human and Economic Burden of Cancer



- 1,444,920 Americans were diagnosed in 2007
- An estimated 559,650 died of cancer in 2007
- \$206.3 billion spent on healthcare costs for cancer in 2006
- 47 million Americans lack health insurance

Cancer Death Rates



A Physicist's Insight into Biology

“It is not that the subject
(biology) was simple enough
to be explained by
mathematics, but rather that
it was much too involved to
be fully accessible to
mathematics.”

$$\hat{H}|\psi\rangle = i\hbar\frac{d}{dt}|\psi\rangle$$



Erwin Schrödinger

1944

“Biology is more like history than it is like physics. You have to know the past to understand the present; there is no predictive theory of biology just as there is no predictive theory of history. The reason is the same; both subjects are too complicated for us.”

Carl Sagan