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

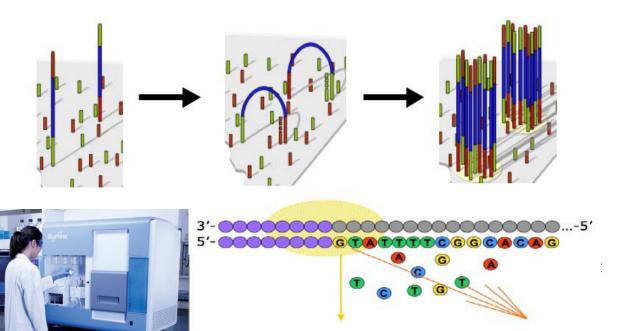
February 27, 2008

U.S. DEPARTMENT OF HEALTH AND HUMAN SERVICES

National Institutes of Health

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

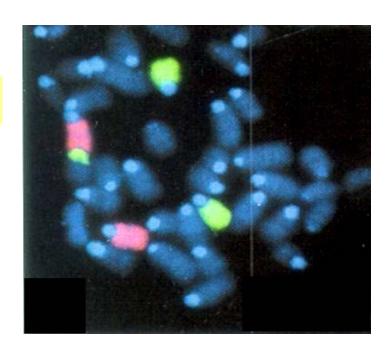




Cancer is a Disease of the Genome

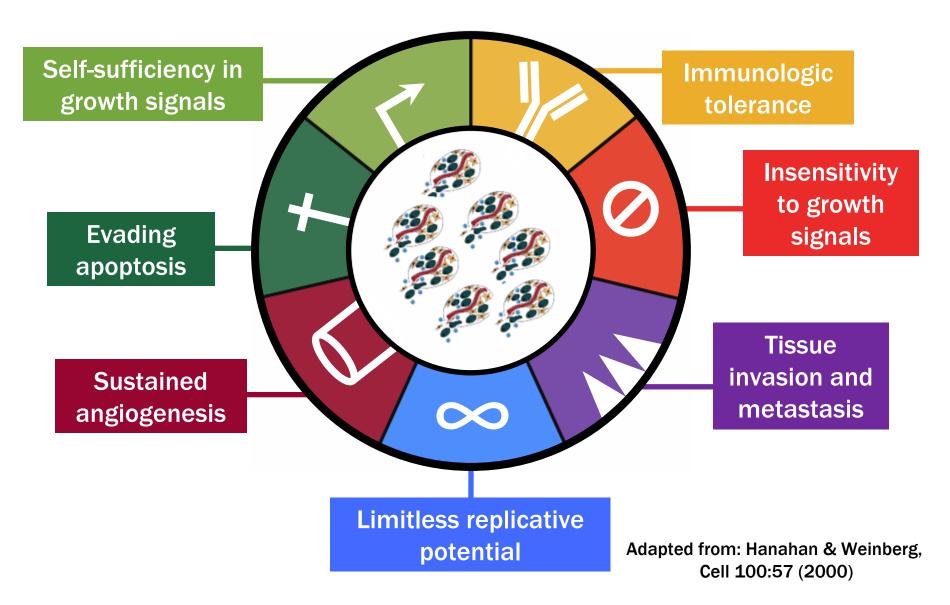
It arises from changes within the DNA of our cells during their lifespan

- Deletions
- Amplifications
- Mutations
- Translocations
- Epigenetic changes



A Complex Foe

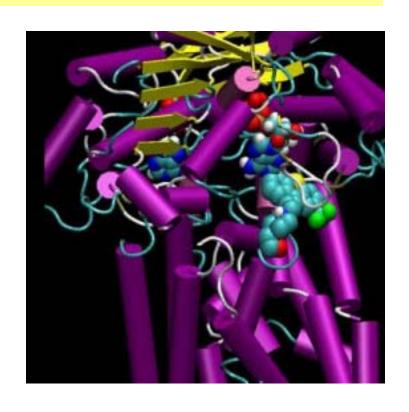
The essential aberrations of cancer



The Complexity of the "Interactome"

Molecular interaction networks

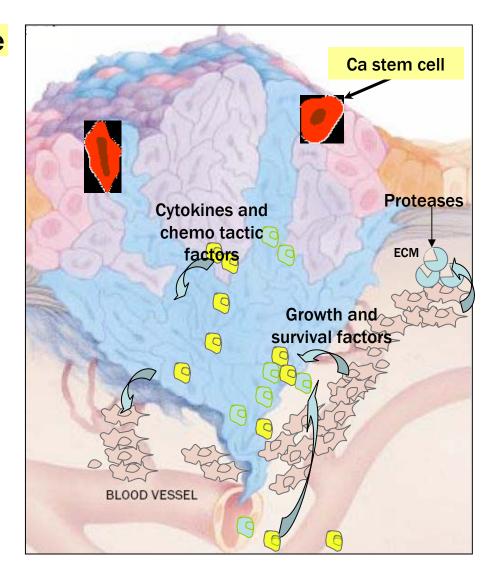
- Protein-protein interactions
- Protein-DNA interactions
- MicroRNA-mRNA interactions



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

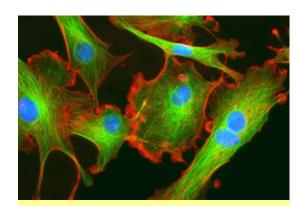
Average Consistent Fold

Name P-Value Change

| 0.0072 | -1.88 |
|--------|--|
| 0.0015 | -1.81 |
| 0.0015 | -1.80 |
| 0.0072 | -1.80 |
| 0.0072 | -1.57 |
| 0.0072 | -1.54 |
| 0.0266 | -1.50 |
| 0.0266 | 1.53 |
| 0.0266 | 1.57 |
| 0.0002 | 1.61 |
| 0.0072 | 3.66 |
| | 0.0015 0.0015 0.0072 0.0072 0.0266 0.0266 0.0266 0.0002 |

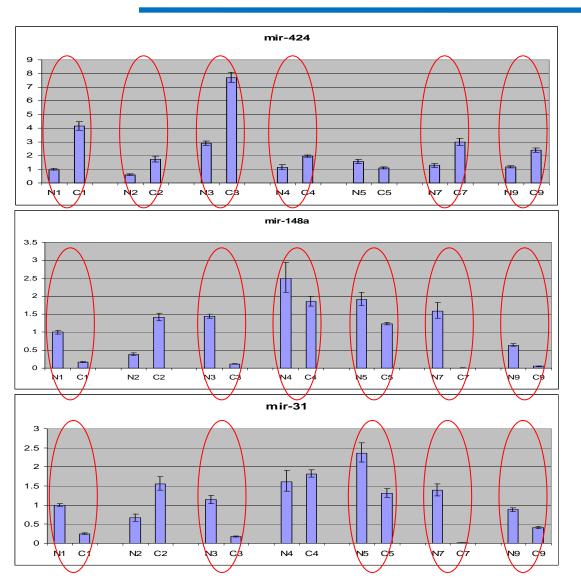
Higher in cancer fibroblasts

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

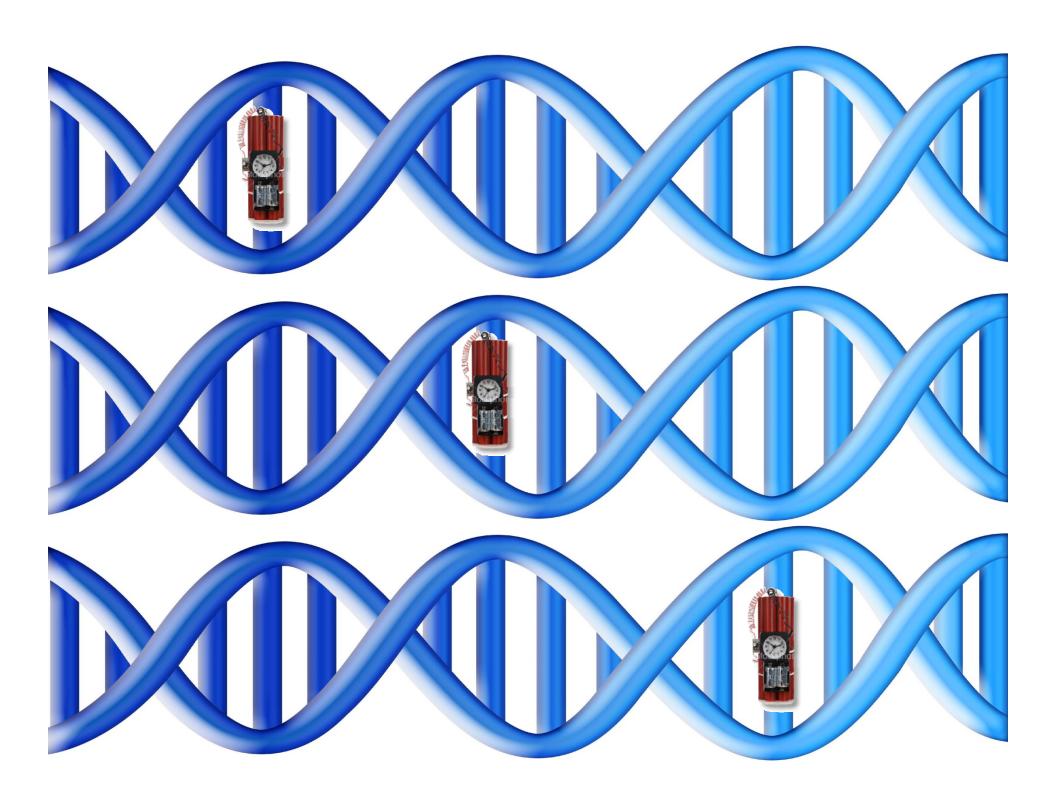
Down regulated in cancer fibroblasts

Down regulated in cancer fibroblasts

gggcataaaggtgatgtcttatcacagatca acattetetataceeteaagteeaagtatataa tctctgagagtctgtaggagtcttagtcacga tactactacgctccactgcagtgatctgctta tcagaaaaacagtgctgcattaattgaattga gaatgtaagagtaaattaacaggagctgcca caggtatttggagtttattatgtgctaaatactg attgacatgaacacctgtcaaacctgcaactc aaggatccaagactatgaagggagaagaga cgtcatcacagaagccttcttggtcccattgg agttttcatagccaagcaaatttccaccattg accatttaatgctgatagtatcctgtgagctag agatctgcagcgtttgtccacacggcagggc tgctggaatctgtgtgtgtgtgtgtgtgtgtgt ctcacatccctccactctaaccttgggtag attgaaaaaatgttagaggcggttcaggcag ccccagtgacccctgatagaattattgcctcc gactctcagaggaggatggagtgtagctgc cagtcactttttaaatcagaagctcgcagga attgttgtgagccctctgctaccacgtggaga ttctcacaaaaccccttggggaagtgaaaca ggaaatgaactccttgagaaagctgacattc gccaatatcaagacactagtggaaaggaac ccaccactccagacaccactcaccgctctag ctactggctagggtactgtaaaacacatggg aaaaccagtgaggctgaaagaacggctgcc tttgcaactctgccggaaacacttctgctcag aacctctgtctgttggggtgggcataaaggtg gggtttcctgctctccactacattctctataccc caaacacaattctatgcccaccattcataccc gatggcacatggcagagcccaagccctctc

THE INTERNATIONAL WEEKLY JOURNAL OF SCIENCE 27 October 2005 www.nature.com/nature ГНЕ НАРМАР PROJECT Chapter and verse on human genetic variation

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Identifying Genetic Markers for Prostate & Breast Cancer

Initial Study

1,150 cases/**1,150** controls

540,000 Tag SNPs

Follow-up Study #1

4,600 cases/ 4,600 controls

>28,000 SNPs

Follow-up Study #2

5,500 cases/ 5,500 controls

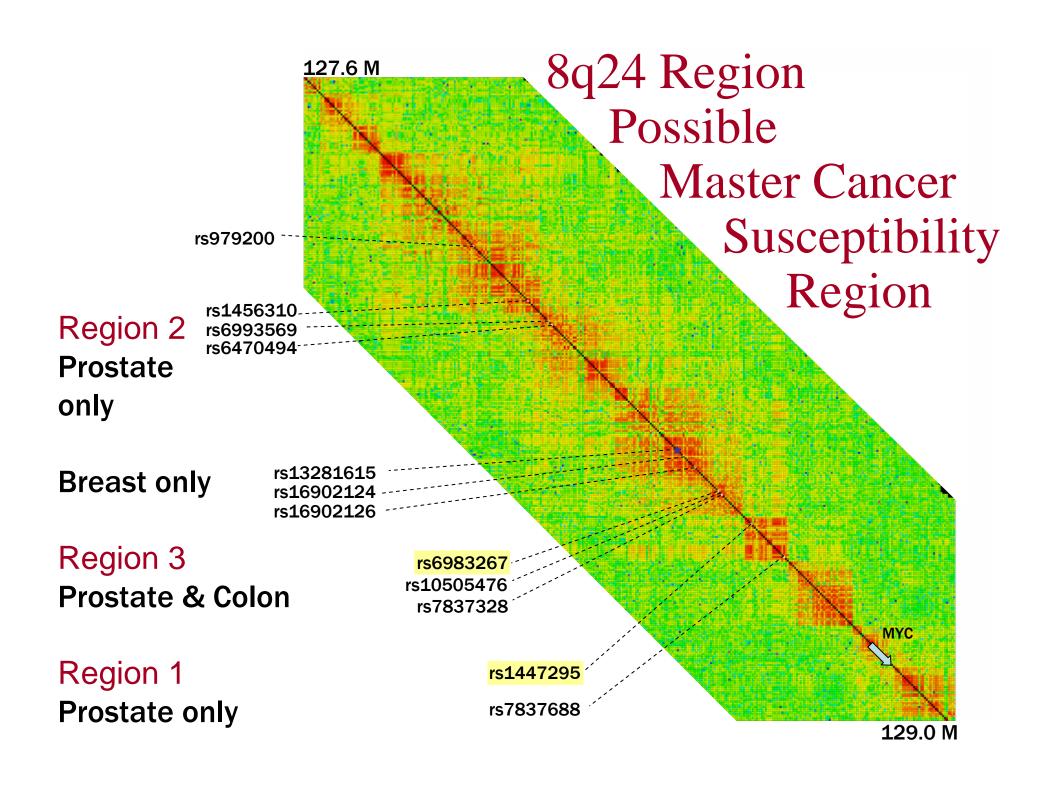
at least7,600SNPs



Stephen Chanock, M.D. Cancer Genome Anatomy Project

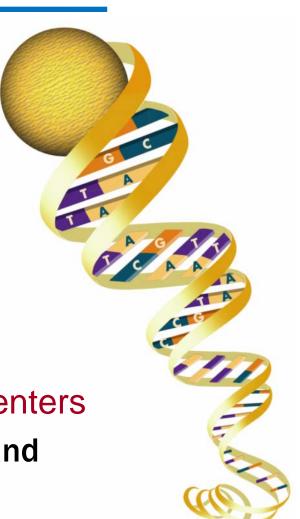
Genotype, Haplotype, Sequence

Determine Causal Variant(s)



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

Therapy Prevention

Alterations in cellular function

Cell biology and protein carbohydrate chemistry

Normal vs. abnormal phenotype

Behavioral science

Regulation of gene expression

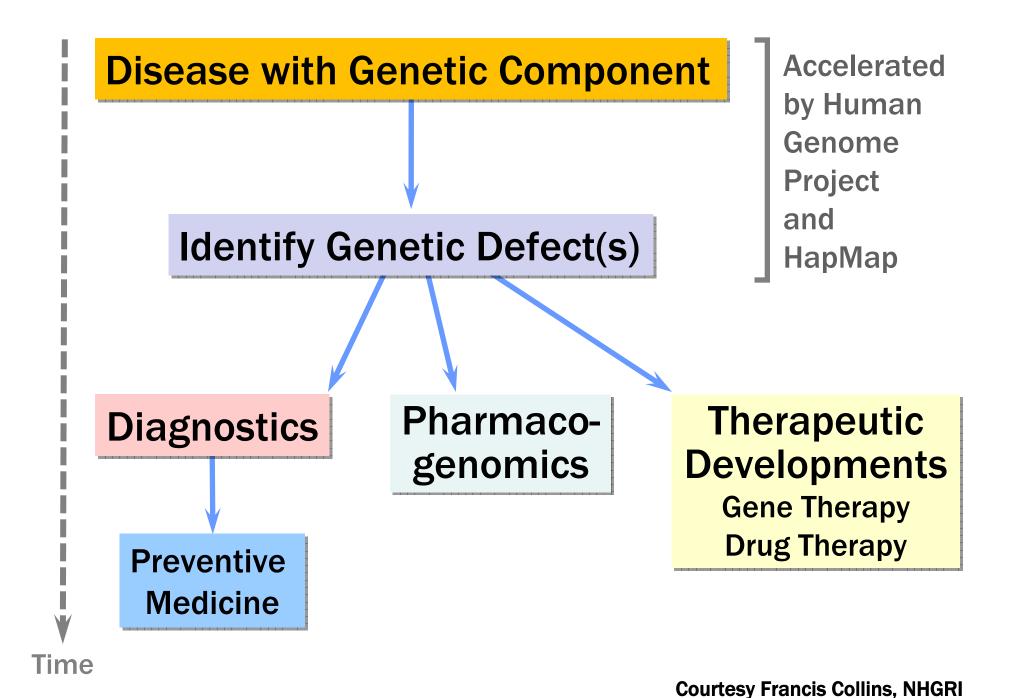
Genes and environment

Populations w/disease (environmental risk exposure) Whole genome sequencing

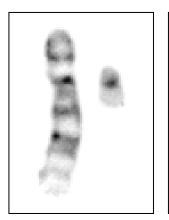
Somatic mutations

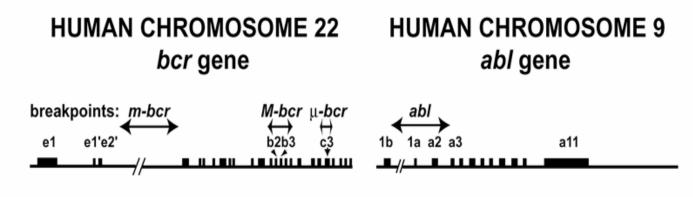
Large populations Longitudinal history Genome- wide association studies

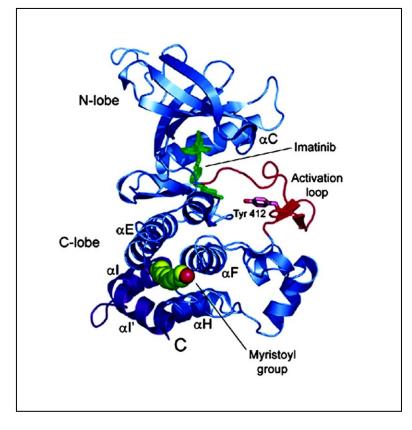
Germline markers of risk

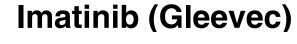


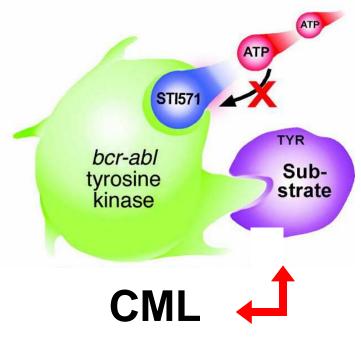
CML t(9;22)



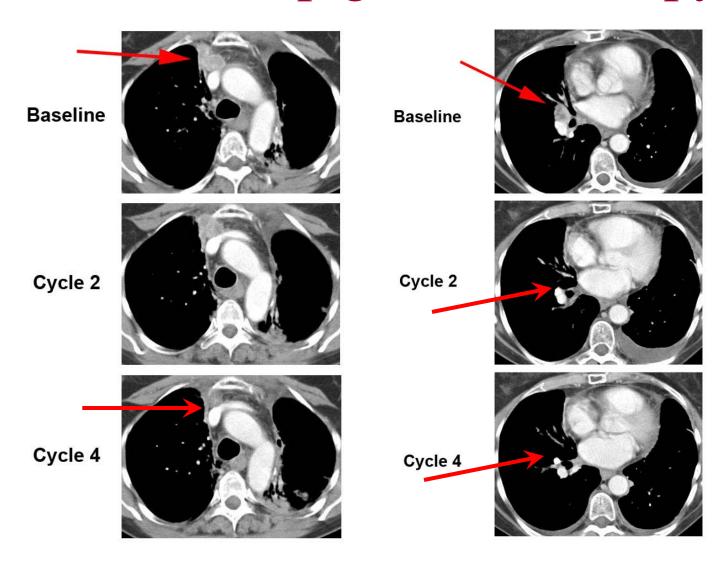




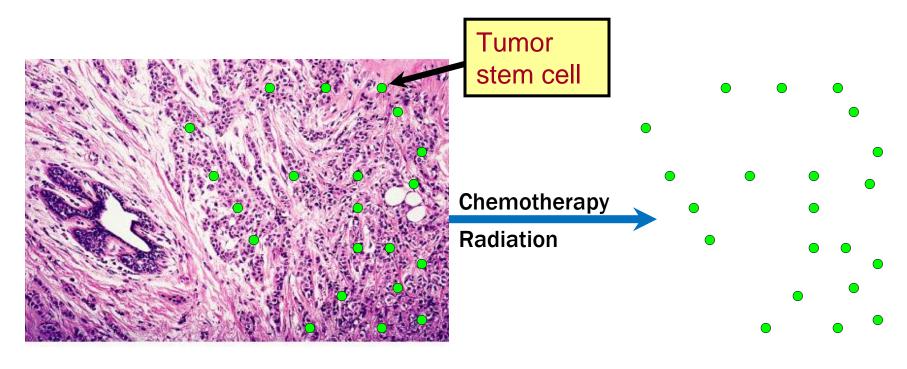




Response in Patient with Metastatic NSCLC to Epigenetic Therapy



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 |

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

Source: Dontu, Genes Dev 17:1253 (2003)

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

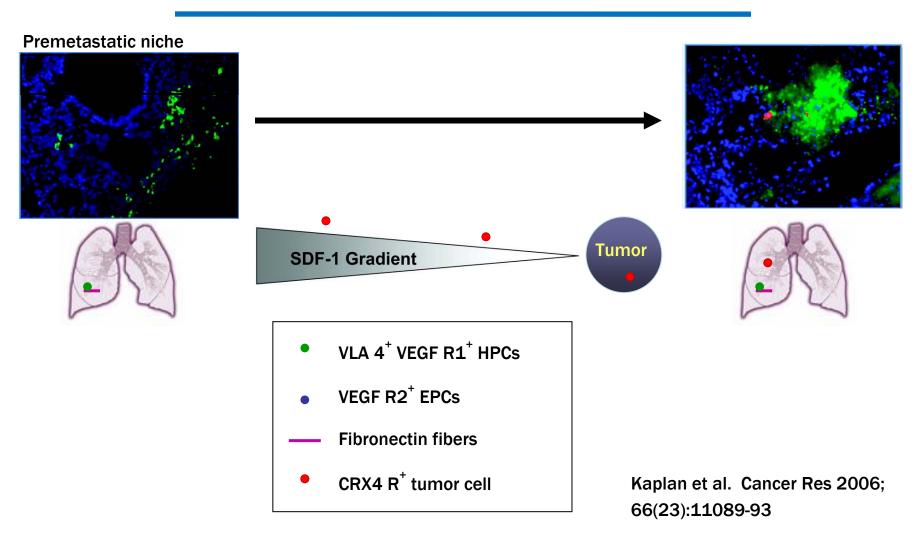
- What role do tissue stem cells play in the process of metastasis?
- What is the role of cancer stem cells/tumor initiator cells in the development of a metastatic lesion?

Chemokines & Role in the Premetastatic Niche

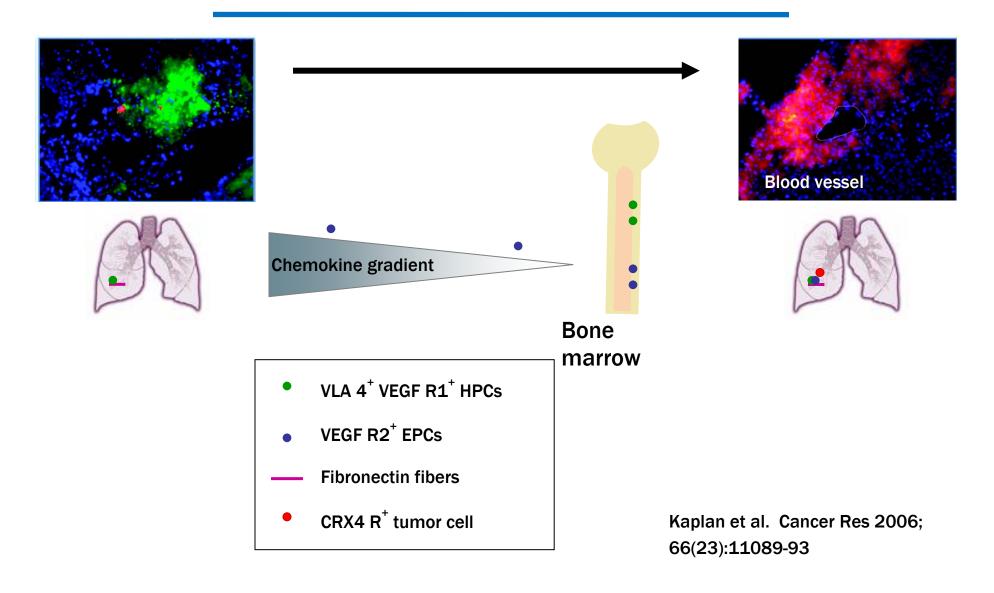
1. Fibronectin recruitment Target Organ Fibronectin expression **Tumor** VLA 4⁺ VFGF R1⁺ HPCs VEGF & other chemokines **Baseline** Day 3 **Day 14 VEGF R2**⁺ **EPCs Fibronectin fibers** 2. Premetastatic niche Chemokine gradient VLA 4 cells preferentially bind to fibronectin Kaplan et al. Nature 2005;438(8): 820 -7 **Bone marrow**

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

Possible Mechanism for Recruitment of Tumor Stem Cells to Premetastatic Niche



Vascular Organization of Metastasis



Neutrophils in a Chemoattractant (fMLP) Gradient



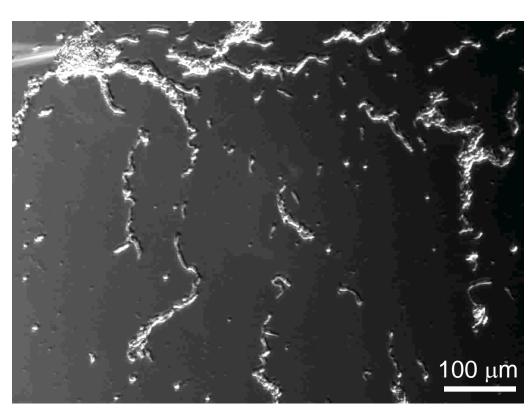


Carole A. Parent, Ph.D.

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

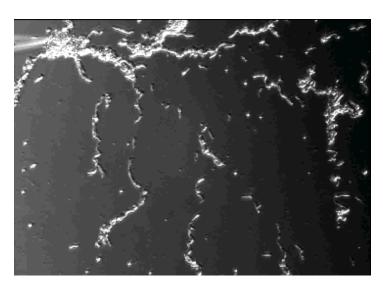


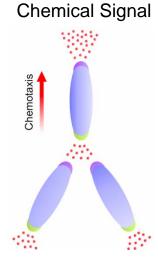
Dictyostelium cells migrating towards a signal-releasing pipette (upper left) over the course of 2 hrs

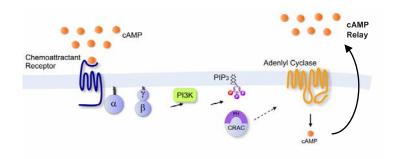
Group Cell Migration

- 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?

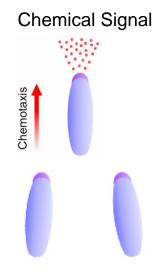


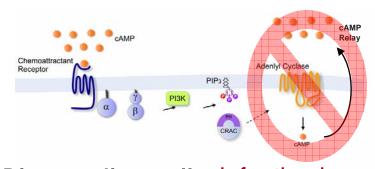




Dictyostelium cells migrating towards a signal-releasing pipette form groups by relaying chemical signals

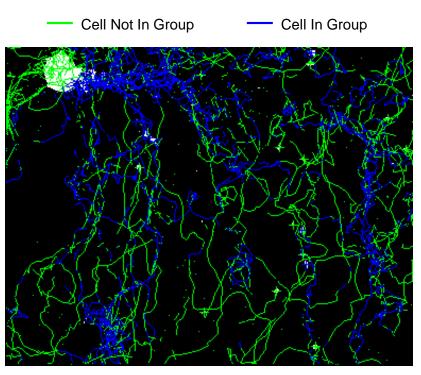




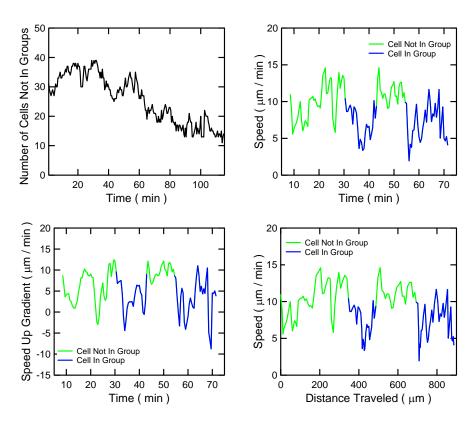


Dictyostelium cells defective in signal relay do not form groups and instead migrate independently

Computational Tools for Statistical Analysis of Complex Biological Behaviors



Automated software tracks the migration of cells from the previous movie

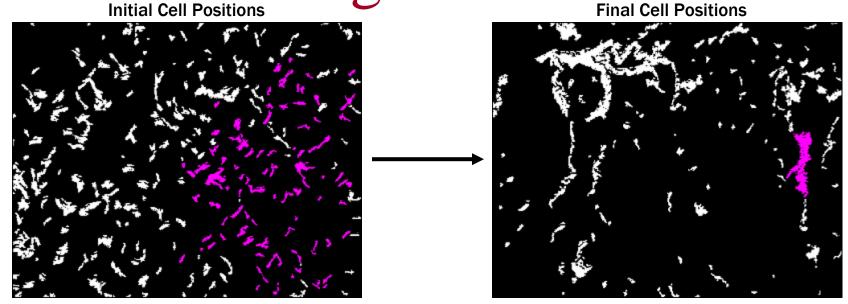


Various statistics for a representative cell from the movie on the left

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

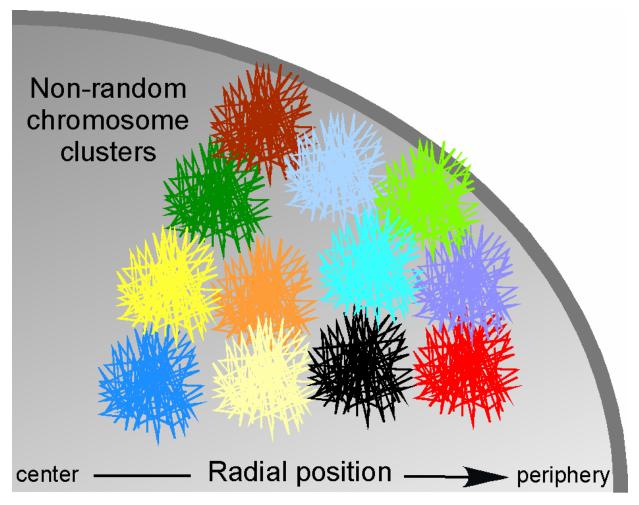
Organization of the Genome in 3D Space



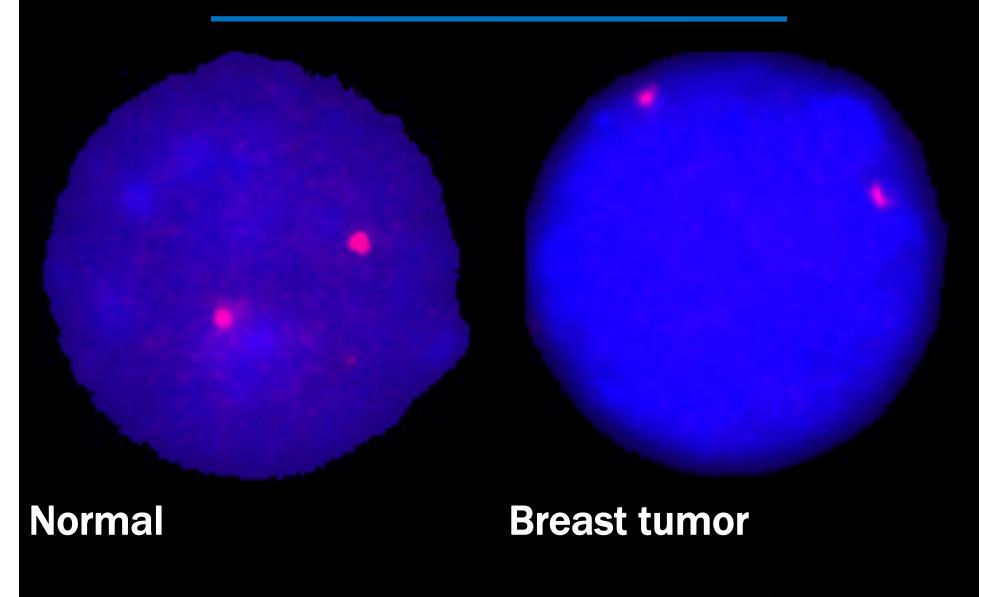
Tom Misteli, Ph.D., NCI



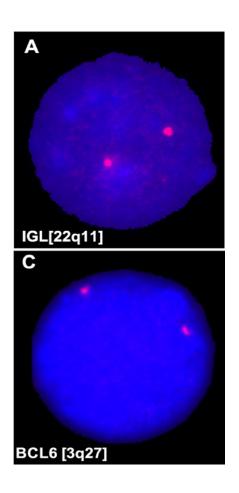
Karen Meaburn, Ph.D., NCI



Diagnostic applications



Non-Random Genome Organization



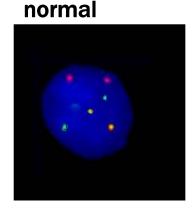
Non-random chromosome position Non-random gene position

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

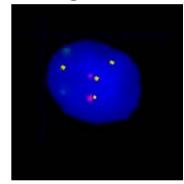
Spatial positioning as a potential diagnostic tool

Interphase Cytogenetics

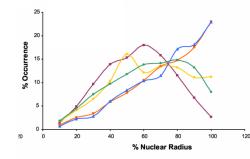




(pre-) malignant

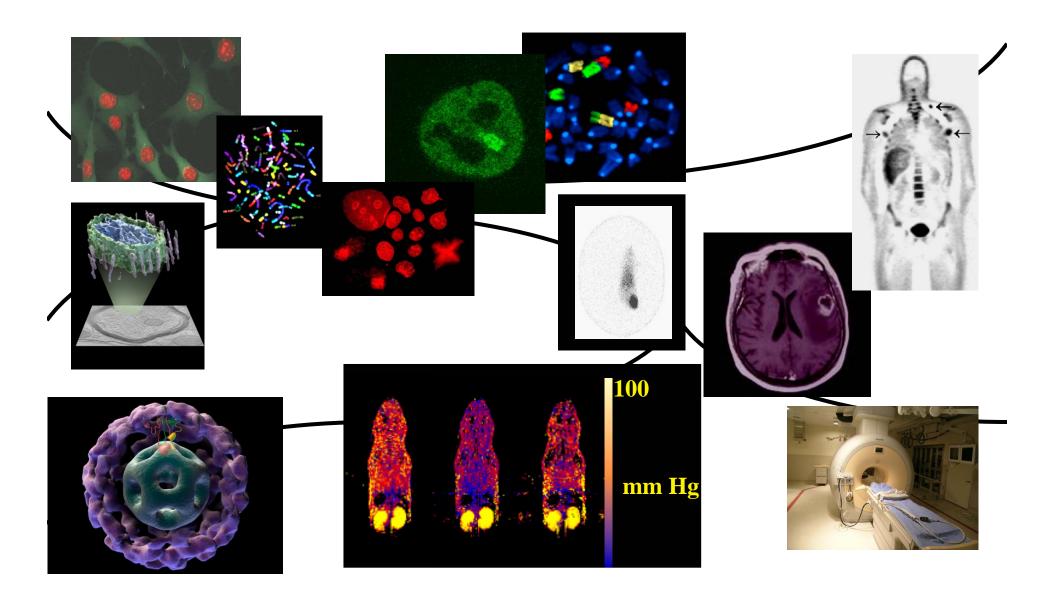


25 20 15 0 0 20 40 60 80 100 1

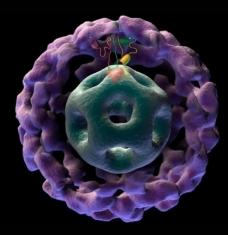


- Normal vs. malignant
- Metastatic cells
- Pre-malignant
- Metastatic potential
- Tumor type

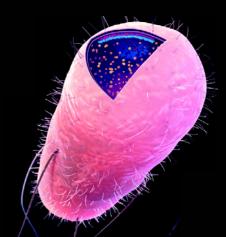
From Molecules to Man



Bridging the imaging gap



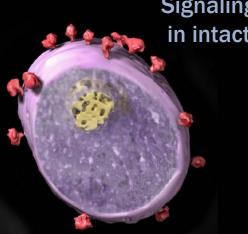
~50 nm



~2000 nm

Dynamic multiprotein complexes

~150 nm



Signaling assemblies in intact cells

~30,000 nm



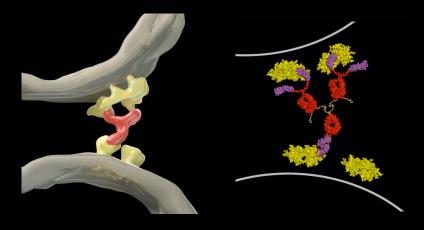
HIV structure and cell entry mechanisms

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-lgP, 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.







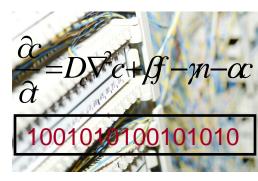


Mammalian



Primates

- Simple
- Adaptable
- Modular
- Personal
- Economical



Mathematical/Computational



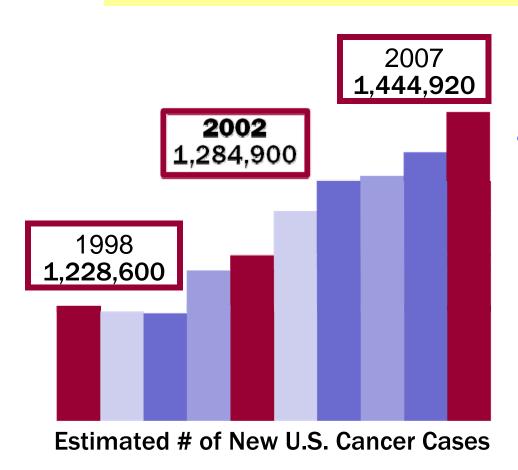
Thoughts in Conclusion

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



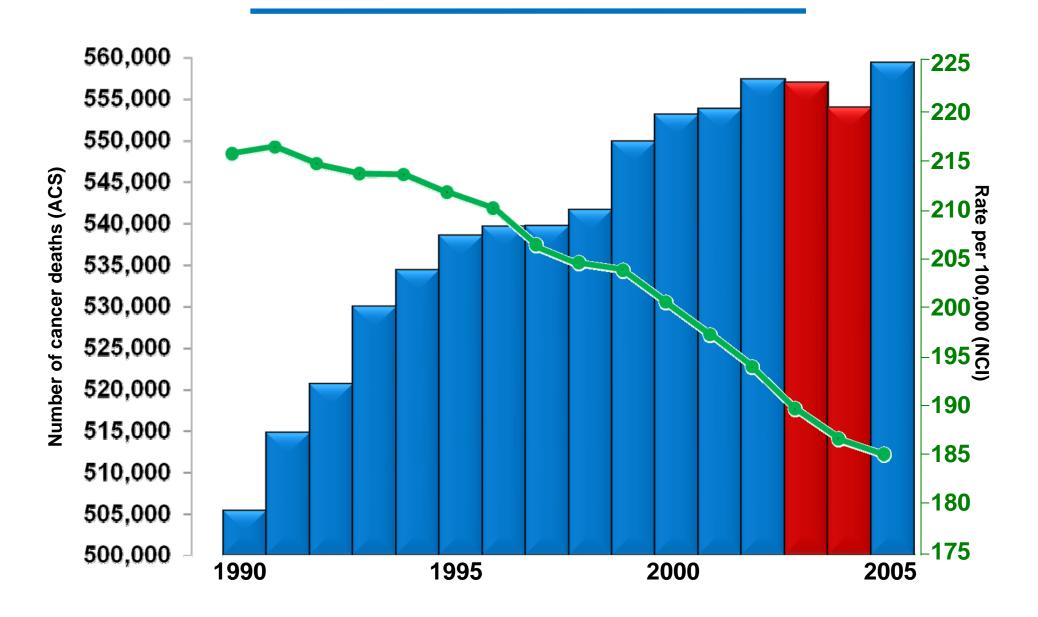
Why What We Are Doing is So Important

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

"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