Introduction to Mathematical Oncology

January 17, 2017

Greene	
Greene, .	۶.

Outline

1) What is Cancer?

- 2 Scientific Questions to Address
- 3 Mathematical Frameworks

4 Mathematical Modeling Applications

- Growth Law of Cancer
- Heterogeneous Populations and Competition
- Models of Chemotherapy and Radiotherapy
- Computing Optimal Treatment Protocols
- The Environment and its Effect on Tumor Development
- The Process of Carcinogenesis
- Dynamics of Drug Resistance
- Dynamics of Cancer and the Immune System

A collection of diseases in which abnormal cells grow uncontrollably by ignoring the normal rules of cell division

- Healthy life requires cooperation to survive
- Cancer is the collapse of this cooperation (selfish)
- A set of fundamental traits (more later):
 - Grow independently from signals (activation of oncogenes, e.g. ras)
 - Ignore stop signals (inactivation of tumor suppressor genes, e.g. p53)
 - Ignore cell death signals (apoptosis)
 - Become effectively immortal
 - Induce angiogenesis (blood supply formation)
 - Metastasize (invade other locations in the body)

Mutations and Clonal Expansion

Abnormal cells undergo **mutations** in their genetic material (DNA) to obtain traits

- Environmental tobacco, radiation, chemicals, diet
- Genetic factors i.e. inherited mutations, random events

Requires multiple mutations, and leads to clonal expansion



Figure 1.1 a: Clonal expansion. Cancer is a multi-gene, multi-step disease originating from single abnormal cell (clonal origin, changes in DNA sequences result in the cliporgersing diavity the mildity absent stage. Successive rounds of mutation & natural selection leads to a mass of abnormal cells called tumours. Some cells in the tumour undergo further rounds of mutations leading to the formation of malignant cells which cause metastasis. (Fig source Allona, <u>www.eds.cell</u>)



Figure 1:1b: Clonal expansion. Normal cells are subject to signals that regulate their proliferation and behaviour. All cancers dirung to most a controls of cell proliferations of no each cell there is a finite number of ways this disruption can occur. Cancer cells develop a degree of autonomy from external regulatory signals that are responsible for normal cellular homestasis. Multiper mutations lead to a tumour mass. Subsequent mutations lead to malignant tumour which break through the basal membrare and gread to distant locations. What would we like to know about disease (basic science, and clinically relevant)?

- How does cancer originate?
 - Time scales for multiple mutations
 - Oncogenes vs. tumor suppressor genes
 - Environment vs. genetic randomness
- How does a tumor grow?
- How does the immune system interact with cancer cells?
- What role does the physical environment play in growth and evolution?
- What is the "best" treatment strategy for chemotherapy or radiotherapy?
- When will a tumor metastasize?
- Why do treatments fail?
 - Interaction between sensitive (wild-type) and resistant clones

See how mathematics can help with some of these questions (and more!)

Mathematical Frameworks

Many different types of math in cancer modeling:

• Ordinary differential equations (ODEs)

$$\frac{dN}{dt} = rN\left(1 - \frac{N}{K}\right)$$

or

$$\frac{dP}{dt} = (\beta - \mu_P - r_0(N))P + r_i(N)Q$$
$$\frac{dQ}{dt} = r_0(N)P - (r_i(N) + \mu_q)Q$$

• Partial Differential Equations (PDEs)

$$\frac{d\sigma}{dt} = \frac{D}{r}\frac{\partial}{\partial r}\left(r^2\frac{\partial\sigma}{\partial r}\right) - A\sigma$$

Stochastic processes

$$\mathbf{A}_{(1)} \xrightarrow{\mathbf{u}_1} \mathbf{B}_{(r)} \xrightarrow{\mathbf{u}_2} \mathbf{C}_{(r_1)} \qquad P_{ij} = \begin{cases} \frac{u_1(N-1)+(1-u_2)rj}{N}, & j=i+1, \\ \frac{(1-u_1)(N-1)}{N}, & j=i-1, \\ \frac{u_2rj}{N}, & j=E, \\ 1-P_{i,i+1}-P_{i,i-1}-P_{i,j-1}, j=i, \\ 0 & \text{otherwise}, \\ \end{cases} \xrightarrow{\mathbb{R}} \mathcal{P} \otimes \mathbb{R} \xrightarrow{\mathcal{P}} \mathbb{Q} \otimes \mathbb{Q}$$

Greene, J.

Mathematical Frameworks continued

- Hybrid/multiscale models (and cellular automaton)
 - Combine multiple frameworks to capture different scales
 - Cells as discrete entities (stochastic evolution)
 - Nutrients diffusing and consumed (PDEs)
 - Chemical networks internal to cell, maybe governing cell-cycle (ODEs)



- Discuss some applications that use mathematics to address scientific questions
- Some will cover, some will not (dependent on time and student interest)

Growth Law of Cancer

Given a set of clinical measurements (volume, cell count, weight, PSA, etc), what type of growth law best describes it?

$$\begin{aligned} \frac{dN}{dt} &= \lambda N, \qquad \frac{dN}{dt} = rN\left(1 - \frac{N}{K}\right), \\ \frac{dN}{dt} &= re^{-aN}N, \qquad \frac{dN}{dt} = aN^{2/3} - bN. \end{aligned}$$

What does each type of model imply in regards to biological mechanisms?



Heterogeneous Populations and Competition

Cancer populations are heterogeneous in nature

• Genetic instability

Greene. J.

- Environmental factors (location w.r.t. stroma, bloody supply, etc.)
- Processes with intrinsic variation (e.g. cell-cycle)

Can clonal variants coexist, or does competition imply the selection of only one phenotype?

• Also understand competition between cancerous and healthy cells

$$\frac{dx_1}{dt} = \Phi(v)x_1$$

$$\frac{dx_2}{dt} = \Phi(v)x_2$$

$$\frac{dy}{dt} = (\alpha H(x_1, x_2, z) - \beta)y$$

$$\frac{dz}{dt} = \gamma y - \delta vz$$

$$v(t) = \frac{z}{x_1 + x_2}.$$
Must 495
January 17, 2017
10 / 20

Heterogeneity continued

Cancer cells also differ based on progression through cell-cycle

- Proliferative (P) vs. Quiescent (Q) vs. Qpoptotic (A)
- More specifically at location of cell-cycle



$$\frac{\partial}{\partial t}n(t,a) + \frac{\partial}{\partial a}n(t,a) = -d(a)n(t,a),$$
$$n(t,a=0) = \int_{a\geq 0} b(a)n(t,a) \, da,$$
$$n(t=0,a) = n_0(a).$$

Models of Chemotherapy and Radiotherapy

Chemotherapy - any chemical agent used to decrease a tumor load

- How do we implement effect of chemotherapy into models?
- Clearly should depend on class of drug AND mutations present in cell
- Drug alkylating agents vs. mitotic inhibitors
- Cell phase of cell-cycle, mutations
- log-kill $(S = e^{-kD})$ and Norton-Simon hypothesis (d(N) = kr(N))

Radiotherapy - treatment of cancer via X-rays

- Sufficient number of DNA "targets"" are "hit" by ionizing radiation \implies cell death (target-hit theory)
- Explain dose-response curves via underlying mechanisms (non-repair and repair)

$$S = 1 - (1 - e^{-D/D_0})^N$$

Computing Optimal Treatment Protocols

A central question: once we understand how chemotherapy effects our model, **what is the best protocol to apply?**

- Best is subjective:
 - Minimize tumor volume
 - Minimize resistant fraction of cells
 - Maximize survival time
- Subject to constraints (total amount of drug)

$$N' = \beta N(1 - N) - \gamma \frac{v(t)}{1 + v(t)} N$$

such that

$$\int_0^{t_f} v(t) \, dt \leq 1.$$

If we want to minimize $N(t_f)$, how should we apply v(t)?

Optimal Control continued

This is like a calculus problem, except instead of finding a number, we must find a **function** v(t).

Use calculus of variations (i.e. optimal control theory) and solve a minimization (or maximization) problem

$$H(N,\lambda,v) = \lambda^T f(N,v) - L(N,v)$$

Sufficient to minimize Hamiltonian H pointwise (often)



The Environment and its Effect on Tumor Development

Understand physical tumor formation and the role of the environment

- Structure of tumor (proliferating rim, quiescent layer, necrotic core)
- Tumor growth into surrounding tissue (nutrients and ECM)
- Angiogenesis and metastasis



Geometry is fundamentally important (location of cells, location of blood vessels, etc.)

The Environment and PDEs continued

Thus, variables should depend on position $x \in \mathbb{R}^{1,2,3}$ and time t

- Partial differential equations (PDEs)
- Often view in fluid mechanics framework (Euler equations, Navier-Stokes)
- Also combine systems (hybrid modeling)
 - Cells as discrete units
 - Nutrients, drugs as fluids

$$\begin{aligned} \frac{\partial n}{\partial t} + \nabla \cdot (n\mathbf{v}) &= (k_m(c) - k_d(c))n, \\ \frac{\partial m}{\partial t} + \nabla \cdot (m\mathbf{v}) &= k_d(c)n, \\ \frac{\partial \rho}{\partial t} + \nabla \cdot (\rho\mathbf{v}) &= 0, \\ \frac{\partial c}{\partial t} + \nabla \cdot (c\mathbf{v}) &= D\nabla^2 c - \gamma k_m(c)n. \end{aligned}$$

Cancer initiation and progression is fundamentally stochastic

- Accumulation of multiple genetic and epigenetic alterations in a single cell, which eventually allow cell to escape homeostatic mechanisms
- What is the probability of developing cancer as a function of age?
- Expected time to cancer initiation
- Mutational profile of a given tumor

Greene, J.

Thus, it is natural to model carcinogenesis via *random variables* (as opposed to deterministic quantities)



Dynamics of Drug Resistance

Although we can kill some (or even most) cancer cells, we cannot usually kill them all

• Cancer populations are extraordinarily heterogeneous



• Heterogeneity \iff heterogeneity

Greene, J.

Questions:

- Is resistance induced (by the drug) or random (and then selected)?
- Is there a qualitative and quantitative difference between induced vs. stochastic mutations (mathematically tractable, while experimentally subtle)?
- What is the optimal way to treat in presence of resistance?
 - Sensitive cells may be inhibiting growth of resistant cells
 - Not obvious what to do (kill too many sensitive, and resistant are free to grow)
- What is the probability resistance existed before treatment began?

$$P_R = 1 - \exp\left(-\frac{MuF}{1 - d/r}\right),$$
$$F = \int_0^1 \frac{1 - b/a}{1 - (b/a)y^{\alpha}} dy$$

Dynamics of Cancer and the Immune System

Why doesn't our immune system prevent (or eliminate) cancers?

- Can we induce an immune response?
- What is the best way to apply immunotherapies in conjunction with chemotherapies?
- Wave of the future (together with targeted therapies)
 - Guide development of combination therapies
 - Determine which patients will respond, and which won't
 - When are combinations needed (immunotherapy and chemotherapy)

$$\begin{split} &\frac{dT}{dt} = aT(1-bT) - cNT - DT - K_T(1-e^{-M})T \\ &\frac{dN}{dt} = eC - fN + g\frac{T^2}{h+T^3}N - pNT - K_N(1-e^{-M})t \\ &\frac{dN}{dt} = -mL + j\frac{D^2T^2}{h+D^2T^2}L - qLT + (r_1N + r_2C)T \\ &-uNL^2 - K_L(1-e^{-M})L + \frac{pLL}{g_I+I} + v_L(t) \\ &\frac{dC}{dt} = a - \beta C - K_C(1-e^{-M})C \\ &\frac{dM}{dt} = -\gamma M + v_M(t) \\ &\frac{dI}{dt} = -\mu_1 I + v_1(t) \\ &D = d\frac{(L/T)^3}{s + (L/T)^4} \end{split}$$