Population and evolutionary dynamics of tumour growth

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Background

Motivation & model formulation

Well-stirred systems: From Invasion to Latency

Systems with spatial inhomogeneities: The role of cell motility

Conclusions & summary

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• Cancer is a disease of clonal evolution within the body¹

¹Nowell Nature (1976)

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- Cancer is a disease of clonal evolution within the body¹
- Although this idea of cancer as an evolutionary problem is not new, it has received less attention than it perhaps deserves

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- The succession of somatic mutations to which cancer cells are subjected leads to clonal expansion and heterogeneity

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- The succession of somatic mutations to which cancer cells are subjected leads to clonal expansion and heterogeneity
- Heterogeneity is a key aspect since it almost directly leads to drug resistance

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Effects on targeted cance therapy³

Evolutionary dynamics of cancer poses a barrier to targeted therapy efficiency

Evolutionary dynamics of carcinogenesis and why targeted therapy does not work

Robert J. Gillies, Daniel Verduzco and Robert A. Gatenby

Abstract | All malignant cancers, whether inherited or sporadic, are fundamentally governed by Darwinian dynamics. The process of carcinogenesis requires genetic instability and highly selective local microenvironments, the combination of which promotes somatic evolution. These microenvironmental forces, specifically hypoxia, acidosis and reactive oxygen species, are not only highly selective, but are also able to induce genetic instability. As a result, malignant cancers are dynamically evolving clades of cells living in distinct microhabitats that almost certainly ensure the emergence of therapy-resistant populations. Cytotoxic cancer therapies also impose intense evolutionary selection pressures on the surviving cells and thus increase the evolutionary rate. Importantly, the principles of Darwinian dynamics also embody fundamental principles that can illuminate strategies for the successful management of cancer.

³Gillies, Verduzco & Gatenby. Nature Rev. Cancer (2012)

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Cancer as an evolutionary ecology problem

Competition between normal and cancer cells



• Tumour dormancy in cancer refers to an extended period of growth restriction of undetected metastases

⁴Willis et al. Cancer Res. **70**, 4310-4317 (2010)

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 - Solitary cells which persist in a quiescent state for months or even years post-resection
 - Non-vascularised, non-angiogenic micro-metastases restricted to a size of 1 to 2 mm in diameter
 - Vascularised metastases that are held at an equilibrium size by the immune system

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Two strategies to use up available resources



Two cellular populations

• Switch-like cells.

Birth pro

$$\begin{array}{l} \mbox{bability} \left\{ \begin{array}{l} \sim 1 \mbox{ when } c > c_2 \\ \sim 1/2 \mbox{ when } c \simeq c_0 \\ \sim 0 \mbox{ when } c < c_1 \end{array} \right. \mbox{ Death probability} \left\{ \begin{array}{l} \sim 0 \mbox{ when } c > c_2 \\ \sim 1/2 \mbox{ when } c \simeq c_0 \\ \sim 1 \mbox{ when } c < c_1 \end{array} \right. \end{array}$$

• Bistable cells.

- Two phenotypes: Proliferating and Quiescent
- Proliferating cells duplicate or change phenotype to quiescent
- Quiescent cells die or change phenotype to proliferating
- The dynamics of the population is then controlled by the rates at which cells switch phenotype:



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Example: Model for the cell-cycle

Simple model for the G_1/S transition

• $x_1 = [Cdh1]$, $x_7 = [CycB/Cdk1]$, m = cell mass

$$\frac{dx_1}{dt} = \frac{(k_{31} + k_{32}A)(1 - x_1)}{J_3 + 1 - x_1} - \frac{k_4 m x_7 x_1}{J_4 + x_1},$$
$$\frac{dx_7}{dt} = k_1 - (k_{21} + k_{22} x_1) x_7$$







Population dynamics

The system we are considering is the competition between an incumbent population of switch-like cells and an invasor made out of bistable cells. These two populations compete by a shared resource, eg oxygen.

Oxygen concentration

$$\frac{\partial c}{\partial \tau} = \nabla^2 c - \kappa c \sum_{k=1}^{N_T} \chi^{(k)}, \quad \frac{\partial c}{\partial x} \Big|_{x=0} = h_0, \quad \frac{\partial c}{\partial x} \Big|_{x=L} = -h_L \tag{1}$$

with $h_0 - h_L = Lh^{-1}\bar{\kappa}\Omega c_0$. The parameter Ω corresponds to the total cell population that a uniform concentration of oxygen $c(x, t) = c_0$ is capable of sustaining. $\chi^{(k)}$ is the population vector corresponding to the *k*th cellular type $k = 1, \ldots, N_T$. $N_T = 3$: Switch-like cells, proliferating bistable cells, and quiescent bistable cells.

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Population dynamics (cont.)

Master Equation for the cellular populations

 $\frac{\partial P(\chi^{(1)},\ldots,\chi^{(N_T)},t)}{\partial t} =$ $\sum_{x_{i}^{(k)}}^{N_{T}} \sum_{x_{i}^{(k)}}^{N} \left(\mathcal{E}_{\chi_{i}^{(k)}}^{-} - 1 \right) T_{\chi_{i}^{(k)} + 1|\chi_{i}^{(k)}} P + \left(\mathcal{E}_{\chi_{i}^{(k)}}^{+} - 1 \right) T_{\chi_{i}^{(k)} - 1|\chi_{i}^{(k)}} P$ $\sum_{i=1}^{N_T}\sum_{j=1}^{N}\sum_{i=1}^{N}\sum_{j=1}^{N}\left[\left(\mathcal{E}^{-}_{\chi^{(k)}_i}\mathcal{E}^{+}_{\chi^{(j)}_i}-1\right)T_{\chi^{(k)}_i+1\chi^{(j)}_i-1|\chi^{(k)}_i\chi^{(j)}_i}P+\right.$ $\left(\mathcal{E}^{+}_{\chi^{(k)}}\mathcal{E}^{-}_{\chi^{(j)}_{i}}-1
ight)T^{(k)}_{\chi^{(k)}_{i}-1\chi^{(j)}_{i}+1|\chi^{(k)}_{i}\chi^{(j)}_{i}}P
ight|+$ $\sum_{i=1}^{N_T} \sum_{j=1}^{N} \sum_{k=1}^{N} \sum_{j=1}^{N} \left[\left(\mathcal{E}_{\chi_i^{(k)}}^- \mathcal{E}_{\chi_j^{(k)}}^+ - 1 \right) T_{\chi_i^{(k)} + 1\chi_j^{(k)} - 1 |\chi_i^{(k)} \chi_j^{(k)}} P + \right.$ $\left(\mathcal{E}_{\chi_{i}^{(k)}}^{+}\mathcal{E}_{\chi_{i}^{(k)}}^{-}-1\right)T_{\chi_{i}^{(k)}-1\chi_{i}^{(k)}+1|\chi_{i}^{(k)}\chi_{i}^{(k)}}P\right]$

where $\chi^{(k)}$ is the population vector corresponding to the *k*th cellular type $k = 1, ..., N_T$ and $\mathcal{E}_{\chi^{(k)}_i}^{\pm} f(\chi^{(k)}_i) = f(\chi^{(k)}_i \pm 1).$

Population dynamics (cont.)

Dimensionless transition rates

Transition rate	Event
$\overline{T_{\chi_i^{(1)}+1 \chi_i^{(1)}}} = \epsilon \frac{e^{\alpha_0(c-1)}}{e^{\alpha_0(c-1)}+e^{-\alpha_0(c-1)}} \chi_i^{(1)}$	Switch-like cell division
$T_{\chi_{i}^{(1)}-1 \chi_{i}^{(1)}} = \epsilon \frac{e^{-\alpha_{0}(c-1)}}{e^{\alpha_{0}(c-1)}+e^{-\alpha_{0}(c-1)}} \chi_{i}^{(1)}$	Switch-like cell death
${\cal T}_{\chi^{(2)}_i+1 \chi^{(2)}_i}=\epsilon\chi^{(2)}_i$	Bistable proliferating cell division
$T_{\chi_{i}^{(3)}+1\chi_{i}^{(2)}-1 \chi_{i}^{(3)}\chi_{i}^{(2)}} = \epsilon w_{0}e^{-H_{PA}(c)}\chi_{i}^{(2)}$	Proliferating-to-quiescent switch
$T_{\chi_{i}^{(2)}+1\chi_{i}^{(3)}-1 \chi_{i}^{(2)}\chi_{i}^{(3)}} = \epsilon w_{0} e^{-H_{AP}(c)} \chi_{i}^{(3)}$	Quiescent-to-proliferating switch
$T_{\chi_{i}^{(3)}-1 \chi_{i}^{(3)}}=\epsilon\chi_{i}^{(3)}$	Bistable quiescent cell death
${\cal T}_{\chi_i^{(k)}-1\chi_j^{(k)}+1 \chi_i^{(k)}\chi_j^{(k)}}=\epsilon u_k\chi_i^{(k)}$	Cell migration $\forall k$ and $\forall j \in $

Table : Dimensionless transition rates for our stochastic model of competition between switch-like and bistable populations.

Aims

- Examine the effect of intracellular noise on the long-term behaviour of the competition between sitch-like and bistable populations
- Explore intracellular noise as a mechanism for latency (i.e. long-term survival of a non-invading mutant)
- Analyse the role of cell motility
- Study the therapeutic implications of our model

Well-stirred systems

Inhomogeneous systems

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Low-noise behaviour: Invasion in well-stirred systems

Hybrid Gillespie simulation results



Figure : Simulation results for the fixation probability of the mutant (bistable) cell population as a function of the barrier-to-noise ratio α . The initial mutant-to-incumbent ratio y = 1/500. $w_0 = 10^2$.

Supression of intracellular noise reduces population noise

Comparison between Gillespie simulations and the mean-field approximation

 $\Omega = 100$

 $\Omega = 1000$



e^a

Linear analysis predicts abrupt transition between invasive a non-invasive regimes as a function of intracellular noise

Linear analysis

The ability of invasion by a small mutant population is provided by linearising of the dynamics of the mutant around the mutant-free steady state. In our case, this correponds to the steady state of Eqs. (3) and (4) with $n_p = n_q = 0$, i.e. c = 1 and $n_{\rm s} = \Omega$:



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High-noise behaviour: Emergence of latency

Hybrid Gillespie simulation results



Figure : Simulation results for the fixation probability of the mutant (bistable) cell population as a function of the barrier-to-noise ratio α . The initial mutant-to-incumbent ratio y = 1/500. $w_0 = 10^2$.

The mutant population exhibits vanishing growth rate under elevated intracellular noise

Neutral invasion dynamics



- We observe that $c(t) = c_0$
- \bullet Constant oxygen concentration implies that Switch-like cells+Proliferating cells+Quiescent cells $\simeq \Omega$
- This allows us to reduce the system and study only the dynamics of the mutant population

Embedded branching process

In order to formulate an analytical theory, we consider the corresponding embedded branching process equivalent to our multi-type birth-death process.

Embedded branching process

- The process of embedding consists of a coarse-graining of time variable carryied out in the following way.
- 2 After birth, each individual (of type *j*) lives for a length of time which is exponentially distributed with characteristic time $\tau_c^{(j)} = \left(\epsilon(1+w_0e^{-\alpha/2})\right)^{-1}$.

At the end of their life-span, each individual produces offspring according to the corresponding generating functions of the per-cell offspring probabilities.

$$G_{P}(x,y) = \frac{w_{0}e^{-\alpha/2}}{1+w_{0}e^{-\alpha/2}}y + \frac{1}{1+w_{0}e^{-\alpha/2}}x^{2}$$
(7)
$$G_{Q}(x,y) = \frac{1}{1+w_{0}e^{-\alpha/2}} + \frac{w_{0}e^{-\alpha/2}}{1+w_{0}e^{-\alpha/2}}x$$
(8)

For full specification of the age-dependent, embedded branching process, we need to give the age distribution which in this case is f(τ) = τ_c⁻¹e^{-τ/τ_c}.

The embedded branching process is critical for high-noise intensity

Let M be the matrix whose entries, $m_{i,j}$, correspond to the average offspring of type j produced by each individual of type j, simply by computing the derivatives of the generating functions: $m_{PP} = \partial_x G_P|_{x=y=1}$, $m_{PQ} = \partial_y G_P|_{x=y=1}$, $m_{QP} = \partial_x G_Q|_{x=y=1}$, and $m_{QQ} = \partial_y G_Q|_{x=y=1}$

Dominant eigenvalue of the offspring matrix



Survival probability for critical branching processes $(\lambda_1 = 1)$

Probability of eventual extincition (P_E)

- If $\lambda_1 > 1 \Rightarrow P_E < 1$
- $If \lambda_1 \leq 1 \Rightarrow P_E = 1$

Asymptotic behaviour of the survival probability $(P_S(t))$ for critical branching processes

- If $\lambda_1=1$ then $P_S(t)\sim \sigma^{-1}t^{-1}$ as $t
 ightarrow\infty$
- $P_S(t) \sim t^{-1}$ implies that there is no characteristic survival time and that both shortand (aribitrarily) long-lived states are possible

The structure of the mutant population supresses population noise (σ)

σ is supressed as ω increases

By supressing population noise, $P_S(t)$, which is inversely proportional to σ , increases and, thus, so does the probability of latency, i.e. a state where growth is supressed by that can last for aribitraly long time



Analytical and numerical results show that:

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- The structure of the bistable population acts as a buffer for population noise, thus contributing to long-lived, growth supressed states to be more likely (latency)

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Low-intracellular noise: Mean-field inhomogeneous model

Mean-field inhomogeneous model

$$\frac{\partial c}{\partial \tau} = \frac{\partial^2 c}{\partial x^2} - \kappa (n_s + n_p + n_q)c, \quad \frac{\partial c}{\partial x}\Big|_{x=0} = h_0, \quad \frac{\partial c}{\partial x}\Big|_{x=L} = -h_L \tag{9}$$

$$\frac{\partial n_s}{\partial \tau} = \epsilon \nu_1 \frac{\partial^2 n_s}{\partial x^2} + \epsilon \tanh\left(\alpha_0(c-1)\right) n_s \tag{10}$$

$$\frac{\partial n_{p}}{\partial \tau} = \epsilon \nu_{2} \frac{\partial^{2} n_{p}}{\partial x^{2}} + \epsilon n_{p} - \epsilon w_{0} P_{PA}(c) n_{p} + \epsilon w_{0} P_{AP}(c) n_{q}$$
(11)

$$\frac{\partial n_q}{\partial \tau} = \epsilon \nu_3 \frac{\partial^2 n_q}{\partial x^2} - \epsilon n_q + \epsilon w_0 P_{PA}(c) n_p - \epsilon w_0 P_{AP}(c) n_q$$
(12)

with $h_0 - h_L = Lh^{-1}\bar{\kappa}\Omega c_0$ no-flux boundary conditions for the equations for the cellular populations

Initial coniditions

$$c(x, t = 0) = 1$$
, $n_p(x, t = 0) = \delta(x - L/2)$, $n_q(x, t = 0) = \delta(x - L/2)$, and $n_s(x, t = 0) = \Omega/L$ if $x \neq L/2$ and $n_s(x = L/2, t = 0) = \Omega/L - 2$

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Bistable mutant invades switch-like incumbent by migrating towards better-perfused regions

Numerical results



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Increased mutant motility increases its aggresiveness

Time to incumbent extinction as a function of mutant motility

 $\bar{\nu}_2 = \bar{\nu}_3 = 10^{-1}$

 $\bar{\nu}_2 = \bar{\nu}_3 = 10^{-2}$





Non-motile mutants are non-invasive but generate long-term coexistence

Non-motile mutants: Stasis followed by transient bursts of mutant



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Mere upregulation of motility is not enough for invasion

Competition between non-motile (incumbent) and motile (mutant) switch-like populations



Summary of results for inhomogeneous systems

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- Spatial inhomogeneities have a non-trivial effect on the competition between our two cell types
- Increased cell motility leads to more aggresive mutants
- Ong-term coexistence between mutant and is possible provided mutant cells are not motile. This is in contrast to the well-stirred case, where no coexistence is possible for low-noise mutant cells

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- Formulation, analytical and numerical results for a model of competition between two populations characterised by two different strategies (response curves) to turn nutrients into offspring
- Intracellular noise shown to be a major factor controlling the outcome of the competition
- Increased cell motility leads to more aggresive mutants although, on its own, it is insufficient to give the advantage to the mutant (previous transformation is required)
- Coexistence between the switch-like and low-noise, non-motile bistable cells is possible in spatially homogeneous systems. It consists of long periods of stasis followed by sudden, short-lived bursts of the mutant population
- High level of intracellular noise leads to critical growth dynamics: growth-supressed states but with aribitrarily long life span

Future work

Explore the therapeutic implications of our model

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