

# A stochastic model of metastatic proliferation

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# The deterministic results

K.Iwata, K.Kawasaki and N.Shigesada [IKS] model for the growth and spread of secondary tumors (metastases):

- the primary tumor, of size  $x_p$  (the cell number, being very large, is considered as a continuous variable), grows with a Gompertzian deterministic law  $\dot{x}_p = g(x_p)$  and  $g(x) := ax \log(N/x)$  here  $a > 0$ ,  $0 \ll N$  (i.e.  $N$  is “macroscopic”, e.g.  $10^{11}$ ).
- meanwhile, it produces one-cell metastases with a size-dependent rate  $\beta(x_p) = mx_p^\alpha$ , with parameters  $m > 0$  and  $\alpha \in (0, 1]$ :  $\alpha$  is related to the connection between tumor and blood circulation (angiogenesis), vascularization is on the surface  $\Rightarrow \alpha \simeq 2/3$ .
- the new metastases grow with the same law and produce other metastases with the same mechanism.



The evolution (von Forster) equation for the size distribution, in the continuum approximation:

$$\frac{\partial \rho(x, t)}{\partial t} + \frac{\partial (g(x) \rho(x, t))}{\partial x} = 0, \quad x > 1 \quad (1.1)$$

$$g(1) \rho(1, t) = \int_1^\infty \beta(x) \rho(x, t) dx + \beta(x_p(t)); \quad \rho(x, 0) = 0 \quad (1.2)$$

Remark:  $x_p(t) = N^{(1-\exp(-at))}$  solves the Gompertz equation with  $x_p(0) = 1$ .

Accurate analysis in more recent papers by two french groups [DGL, BBHV].

The main point is the evaluation of the malthusian rate associated to the asymptotic exponential growth of the metastatic population.





# The stochastic model

First step: modelling growth of a single tumour.

- It is a BD (birth-death) process (equivalently, a random walk on the space of sizes  $\mathbb{Z}_+ = \{0, 1, 2, \dots\}$ ), with suitable rates.
- In the initial segment  $[1, N]$ , its size tends to increase ( $B > D$ ), while beyond  $N$ , the size tends to decrease ( $D > B$ ).
- 0 is a.s. absorbing.

Remark: the stochastic model of a single tumor has a behavior not so different from the deterministic one; if the tumor reaches a macroscopic size the mean time to extinction becomes extremely long. What is actually observed in the long run, if no extinction occurred, is the quasi-stationary state, i.e. the distribution of sizes, conditioned to not being absorbed, see the recent review on quasi-stationary distributions [vDP]



Proposed rates:

- birth rate,  $\lambda_n = an \log(N + 1)$ ,  $n \in \mathbb{Z}_+$
- death rate  $\mu_n = an \log(n + 1)$ ,  $n \in \mathbb{Z}_+$ .

Remark:  $N$  is a large bounding size, the drift  $\lambda_n - \mu_n$  is positive from 1 to  $N - 1$ , zero on  $N$  and negative after  $N$ , such that asymptotically reproduces the deterministic, Gompertzian law.



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For general positive rates of a birth-death process, we consider the potential coefficients  $\pi_n$ ,  $n \in \mathbb{N}$ , i.e. on the set of transient states  $\mathbb{N} = \{1, 2, \dots\}$

$$\pi_1 = 1, \pi_n = \frac{\lambda_1 \lambda_2 \dots \lambda_{n-1}}{\mu_2 \mu_3 \dots \mu_n}, \quad n = 2, 3, \dots \quad (2.1)$$

and suppose that the following conditions are fulfilled:

$$\sum_{n=1}^{\infty} \left( \frac{1}{\lambda_n \pi_n} \right) = \infty \text{ i.e. non-explosion}$$

and

$$\sum_{n=1}^{\infty} \left( \frac{1}{\lambda_n \pi_n} \sum_{i=1}^n \pi_i \right) = \infty$$

These conditions (fulfilled in our case) imply that absorption in 0 is certain. The transition probabilities

$$P_{i,j}(t) = \mathbb{P}\{X(t) = j | X(0) = i\}$$

are the unique solution of the Kolmogorov Backward and Forward Equation (KBE, KFE) systems.



These equations can be compactly formulated by introducing the generator  $Q$  of the process:  $Q$  is a tridiagonal matrix such that

$$Q_{i,i} = -(\lambda_i + \mu_i), \quad i = 0, 1, \dots, \quad Q_{i,i-1} = \mu_i, \quad i = 1, 2, \dots; \\ Q_{i,i+1} = \lambda_i, \quad i = 0, 1, \dots; \quad Q_{i,k} = 0 \text{ otherwise}$$

Let  $P(t)$  denote the transition probability matrix, we shortly write the two Kolmogorov equations:

$$\dot{P} = QP, \quad (\text{KBE})$$

$$\dot{P} = PQ, \quad (\text{KFE})$$

Remark: as absorption in 0 is certain, the stationary distribution of the process is just  $\delta_0$ .



We moreover have that a positive decay parameter  $\Lambda_+$ ,  $\Lambda_+ := \lim_{t \rightarrow \infty} 1/t |\log P_{ij}(t)|$ ,  $i, j \in \mathbb{N}$  exists; this comes from the asymptotic behavior of the  $\pi_j$  for large  $j$ , which allows the following result for the quantity  $R_n := (\sum_{j=1}^n \frac{1}{\mu_j \pi_j})(\sum_{j=n}^{\infty} \pi_j)$  see [SZP]:

$$\sup_{n \geq 1} R_n := R < \infty \quad (2.2)$$

This implies that the decay parameter is positive, as  $(4R)^{-1} \leq \Lambda_+ \leq R^{-1}$ .







# The reduced equations

Model proliferation i.e. the appearance of secondary tumours at small, unit size: it is represented by a given rate of creation of “particles” in the site 1, through a linear functional of the system configuration.

Remark: at the initial stage of growth (size = 1!) extinction happens with “reasonable” mean times; this is not present in the deterministic model, hence different results are expected.

The proliferation process is represented in the configuration space  $\mathbb{S} = \{\eta : (\eta_1, \dots, \eta_n, \dots), \eta_n = \# \text{metastases of size } n, n = 1, 2, \dots\}$ , in the following way: the population on the site 1 (i.e. the number of tumours of size one) increases with a rate  $C(\eta)$  (the colonization rate) which depends linearly on the current configuration  $C(\eta) = \sum \beta_n \eta_n$ .



We just analyze the evolution of the expected values.

Let the expected occupation numbers be  $\rho_k(t) = \langle \eta_k(t) \rangle$ ,  $k = 1, 2, \dots$ . The initial measure is concentrated on the configuration with just one particle in the site 1, i.e. just one (ancestor) cell.

and let  $\underline{e}_1$  denote the unit vector on the site 1.



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By linearity we get the following system for the vector  $\underline{\rho}(t)$ , with  $C(\underline{\rho}) = \langle C(\eta) \rangle$ :

$$\begin{aligned}\dot{\underline{\rho}} &= \underline{\rho}Q + C(\underline{\rho})\underline{e}_1 \\ \underline{\rho}^0 &= \underline{e}_1\end{aligned}\tag{3.1}$$

Let  $c(t) \equiv C(\underline{\rho}) = \sum \beta_n \rho_n(t)$ , and write the associated componentwise integral equation

$$\rho_k(t) = P_{1,k}(t) + \int_0^t c(s)P_{1,k}(t-s)ds\tag{3.2}$$

Multiplying by  $\beta_k$  and sum over  $k$ , a Volterra integral equation for the colonization rate  $c(t)$  is got:

$$c(t) = \gamma(t) + \int_0^t c(s)\gamma(t-s)ds\tag{3.3}$$

where  $\gamma(t) \equiv \sum_k \beta_k P_{1,k}(t) = \mathbb{E}(\beta_{X(t)})$ .



Key estimates:

- for the transition probabilities

$$P_{1,k} \leq M_k \exp(-\Lambda_+ t) \quad (3.4)$$

- similar for  $\gamma(t)$ ,

The following integral plays a key role:

$\int_0^\infty \gamma(t) dt = \hat{\gamma}(0)$ , where  $\hat{f}$  denotes Laplace transform of  $f$ .

If  $\hat{\gamma}(0) < 1$ ,  $c(t)$  decays exponentially, while exponential growth comes out if  $\hat{\gamma}(0) > 1$ ;  $c(t)$  goes to a constant if  $\hat{\gamma}(0) = 1$ . Some properties of the solution  $c(t)$  propagate to the distribution function  $\rho(t)$ , as the inhomogenous term  $P_{1,k}(t)$  is decaying to zero and the integral term is a convolution between  $\gamma$  and  $P_{1,k}(t)$ .



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Final observations on possible actions on the parameters in order to get the desired, decaying behavior.

Formula giving the relevant integral  $\int_0^\infty \gamma(t)dt$  in terms of the rates:

$$\int_0^\infty \gamma(t)dt = \mathbb{E}\left(\int_0^\tau \beta_{X(t)}\right) = \frac{1}{\mu_1} \sum_{k=1}^\infty \beta_k H(k) \quad (3.5)$$

where  $H(1) = 1$ ,  $H(k) = \frac{\lambda_1 \dots \lambda_{k-1}}{\mu_2 \dots \mu_k}$ , see [SW].

This formula allows to evaluate the role of different parameters in order to decrease the value of the integral below 1, in particular by suitable decreasing the birth rates in a initial segment.



Forthcoming work (with K.Ravishankar) will focus on the full many-particle system, where the absorption competes with proliferation; it may be interesting to investigate the condition for the existence of a stationary distribution on the transient set. A resemblance with the Fleming-Viot system may be noted, see [FM].



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