

# Some insights for the impact of intratumoral heterogeneity on tumor progression

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# The Go or Grow(GoG) System

Due to the migration/proliferation dichotomy, we can distinguish the total population density of cancerous cells  $\rho$  in two groups,  $\rho^p$  the proliferating cell density and with  $\rho^m$  the density of moving cells. Then we end up with the GoG model, **Böttger et al. 2015**,

$$\frac{\partial \rho^m}{\partial t} = \Delta \rho^m + r_{p \rightarrow m}(\rho) \rho^m - r_{m \rightarrow p}(\rho) \rho^m - r_d \rho^m, \quad (1)$$

$$\frac{\partial \rho^p}{\partial t} = r_b \rho^p (1 - \rho^p) + r_{m \rightarrow p}(\rho) \rho^p - r_{p \rightarrow m}(\rho) \rho^p - r_d \rho^p, \quad (2)$$

where  $r_b$  is the birth rate,  $r_d$  is the death rate,  $r_{p \rightarrow m}(\rho)$  is the switch rate from proliferating type to motile type and  $r_{m \rightarrow p}(\rho)$  is the switch rate from motile type to proliferating type of the cancerous cells.

# The Go or Grow(GoG) Single Equation

To obtain a unique equation for  $\rho = \rho^m + \rho^p$ , we refer to detailed balance condition

$$\rho^p r_{p \rightarrow m}(\rho) = \rho^m r_{m \rightarrow p}(\rho).$$

Through this condition we can rewrite the preceding system as a single PDE

$$\frac{\partial \rho}{\partial t} = \Delta(r_{p \rightarrow m}(\rho)\rho) + r_b r_{m \rightarrow p}(\rho)\rho(1 - \rho) - r_d \rho. \quad (3)$$

A possible choice of the switching rates is given by, **Böttger et al. 2015**,

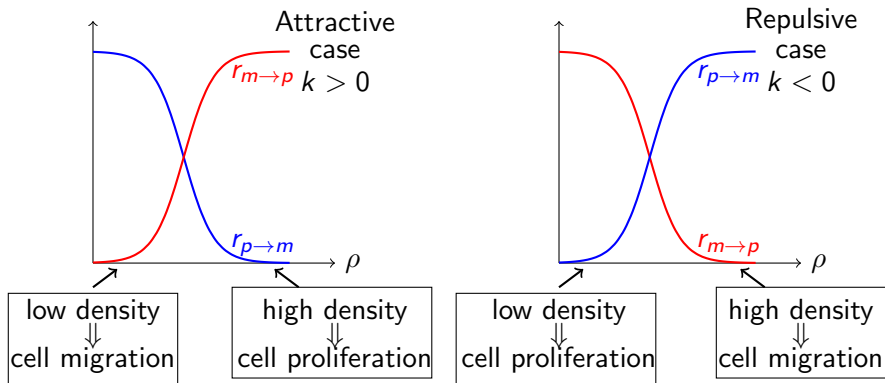
$$r_{m \rightarrow p}(\rho) = \frac{1}{2} (1 + \tanh(k(\rho - \theta))) := r_s(\rho, k), \quad (4)$$

$$r_{p \rightarrow m}(\rho) = 1 - r_s(\rho, k), \quad (5)$$

where the switching parameter  $k$  determines how each tumor cell interprets its microenvironment. Then (3) takes the form

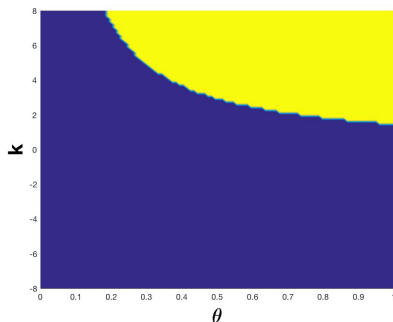
$$\frac{\partial \rho}{\partial t} = \Delta((1 - r_s(\rho, k))\rho) + r_b r_s(\rho, k)\rho(1 - \rho) - r_d \rho. \quad (6)$$

# Role of Switching parameter $k$ .



**Figure 1:** Sketch of cellular mechanism for different kind of phenotypic plasticity. In the left aggregative configuration is represented, while in the right figure, the repulsive configuration is represented, **Böttger et al. 2015**

# Allee Effect (Deterministic Case– Constant Heterogeneity)

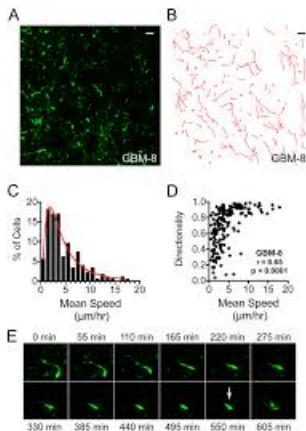


**Figure 2:** Allee effect in the deterministic system (7)-(8). The yellow area represents the area where 0 is stable for the above system, whereas the blue one depicts the area where 0 becomes unstable, **Böttger et al. 2015**

$$\frac{d\rho}{dt} = r_b r_s(\rho, k)\rho(1 - \rho) - r_d \rho := F(\rho), t \in (0, T], \quad (7)$$

$$\rho(0) = \rho_0. \quad (8)$$

# Data driven motivation for introduction of heterogeneity



**Figure 3:** High temporal resolution path tracking highlights intratumoral heterogeneity of speed and directionality of tumor cell migration in focal micro-regions, **J. J. Parker et. al 2018**.

# Intrinsic Heterogeneity as Noise

- Main Assumption: The phenotypic switch is in-homogeneously regulated through the glioma cell populations.

Then the desirable heterogeneous regulation is realised by  $k$  following a probability distribution, i.e. we heuristically take

$$k \mapsto k_0 + \text{Noise} := \xi_t.$$

Thus (7)-(8) takes the form

$$\begin{aligned} \frac{d\rho}{dt} &= F(\rho, \xi_t), t \in (0, T], \\ \rho(0) &= \rho_0 \end{aligned}$$

which is not so well behaved since  $F$  is a nonlinear function.



## Approximation of White Noise

We consider the following approximation of the White Noise  $\xi_t$

$$\xi_t^N = \sum_{i=1}^N \alpha_i^N \mathbb{1}_{A_i^N}(t) \quad \text{where} \quad \alpha_i^N = \frac{B_{\frac{i-1}{N}T + \Delta_N} - B_{\frac{i-1}{N}T}}{\Delta_N} \sim \mathcal{N}(0, N),$$

(9)

where  $B_j$  are Brownian motions and

$$A_i^N = \left[ \frac{(i-1)}{N}T, \frac{i}{N}T \right], \quad i = 1, \dots, N.$$

Therefore we end up with the following well posed system

$$\frac{d\rho^N}{dt} = F(\rho^N, \xi_t^N(\omega)), \quad t \in (0, T], \quad \omega \in \Omega = \text{probability space} \quad (10)$$

$$\rho^N(0) = \rho_0. \quad (11)$$

# Convergence as $N \rightarrow \infty$ (White Noise)

## Theorem

Let  $\xi_t^N$  be the process described by (9). Then the solution trajectories  $\rho_t^N$  of

$$\begin{cases} \frac{\partial \rho^N}{\partial t} = r_b r_s(\rho^N, \xi_t^N) \rho^N (1 - \rho^N) - r_d \rho^N, \\ 0 \leq \rho(0) = \rho_0 \leq 1, \end{cases}$$

converge uniformly in time and almost surely to the solution  $\rho$  of the following ODE problem

$$\begin{cases} \frac{d\rho}{dt} = \frac{1}{2} r_b \rho (1 - \rho) - r_d \rho \\ 0 \leq \rho^N(0) = \rho_0 \leq 1. \end{cases} \quad (12)$$

Namely,

$$\lim_{N \rightarrow \infty} \sup_{t \in [0, T]} |\rho_t^N - \rho_t| = 0 \quad a.s. .$$

# Elimination of Allee Effect (White Noise)

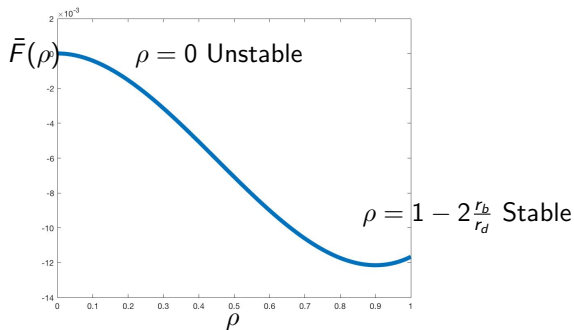


Figure 4: Plot of the potential of equation (12), namely  $\bar{F}(\rho) := \frac{1}{2}r_b\rho(1 - \rho) - r_d\rho$  with  $r_b = 0.1$  and  $r_d = 0.02$ .

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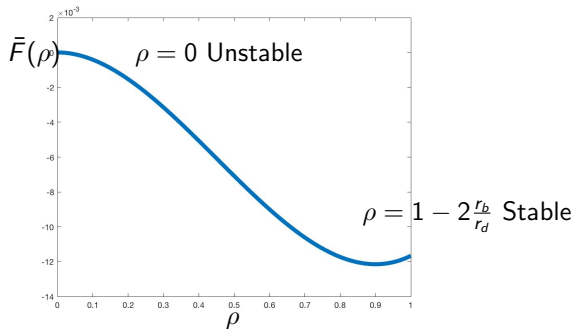


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- The strength of White Noise is so big that annihilates any other influence and force the system to return to a monostable configuration!

## $\xi_t = \text{Gaussian Noise}$

By clinical data provided in **J. J. Parker et. al 2018** we get that the mean speed of cells follows a lognormal distribution  $\implies k \sim \mathcal{N}(\mu_1, D_1)$ . Therefore we consider

$$\xi_t^D = \xi_t \sim \mathcal{N}(\mu, D), D = \text{noise strength.}$$

### Approximation of Gaussian Noise

We consider the following approximation of the Gaussian Noise  $\xi_t$

$$\xi_t^{D,N} = \sum_{i=1}^N \alpha_i^N \mathbb{1}_{A_i^N}(t) \quad \text{where} \quad \alpha_i^N \sim \mathcal{N}(\mu, D), \quad (13)$$

and thus we end up with the model

$$\frac{\partial \rho^{D,N}}{\partial t} = r_b r_s (\rho^{D,N}, \xi_t^{D,N}) \rho^{D,N} (1 - \rho^{D,N}) - r_d \rho^{D,N} := \bar{F}(\rho), \quad (14)$$

$$0 \leq \rho^{D,N}(0) = \rho_0 \leq 1. \quad (15)$$

# Convergence as $N \rightarrow \infty$ (Gaussian Noise)

## Theorem

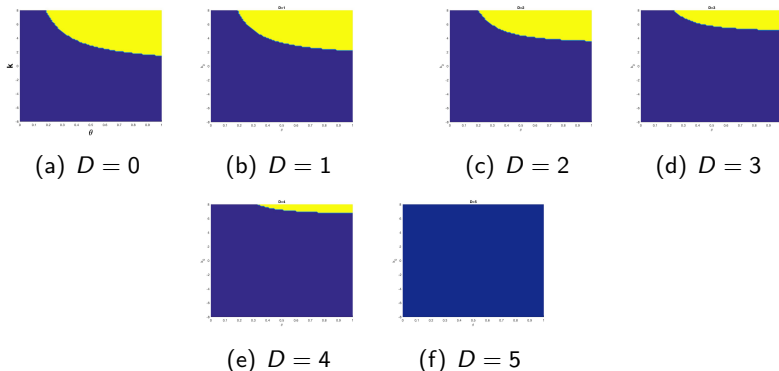
Assume that  $\xi_t^{D,N}$  is given by (13). Then each solution trajectory of (14)-(15) converges uniformly in time and almost surely to the unique solution of

$$\begin{cases} \frac{d\rho^D}{dt} = \mathbb{E} [f^D(\rho^D)] r_b \rho^D (1 - \rho^D) - r_d \rho^D \\ 0 \leq \rho^D(0) = \rho_0 \leq 1, \end{cases} \quad (16)$$

where

$$f^D(x)[y] = \begin{cases} 0, & y \leq 0, \\ \text{sign}(x - \theta) \frac{2e^{-\left(\left(\frac{\tanh^{-1}(2y-1)}{(x-\theta)} - k_0\right) \frac{1}{D\sqrt{2}}\right)^2}}{D\sqrt{2\pi}(x-\theta)(1-(2y-1)^2)}, & y \in (0, 1), \\ 0, & y \geq 1. \end{cases}$$

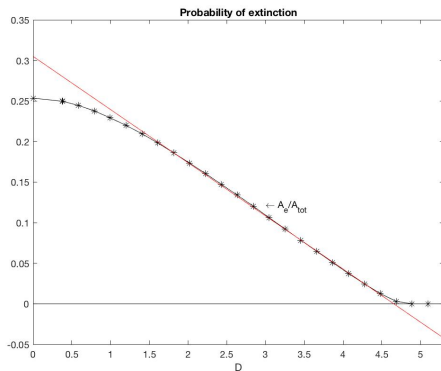
# Elimination of Allee Effect (Gaussian Noise)



**Figure 5:** Plot of the indicator function  $S_{t_0}(\theta, k_0, D)$ . In yellow area there holds  $S_{t_0}(\theta, k_0, D) = 1$ , whilst  $S_{t_0}(\theta, k_0, D) = 0$  in blue area.

$$S_{t_0}(\theta, k_0, D) = \begin{cases} 1 & \text{If } 0 \text{ is stable for the system (16)} \\ 0 & \text{If } 0 \text{ is unstable for the system (16)} \end{cases}$$

# Probability of Tumor's Extinction (Gaussian Noise)



**Figure 6:** Probability of extinction:  $\frac{A_e}{A_{tot}}$  as a function of variance  $D$ . Notably this probability decreases almost linearly as  $D$  increases.



# Probability of Tumor's Extinction (Gaussian Noise)

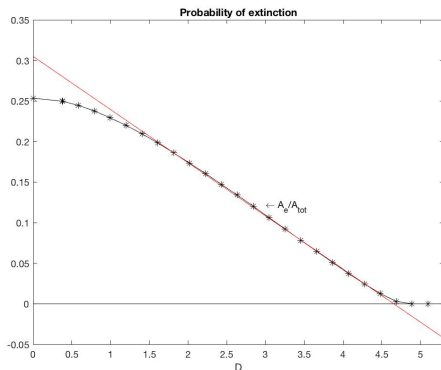


Figure 6: Probability of extinction:  $\frac{A_e}{A_{tot}}$  as a function of variance  $D$ . Notably this probability decreases almost linearly as  $D$  increases.

- There is a critical level of heterogeneity (or noise magnitude e.g.  $D = 5$ ), above which the tumor survives.

# Spatio-temporal GoG model (White Noise)

## Theorem

For every  $T > 0$  the unique solution  $\rho^N$  of problem

$$\begin{aligned}\frac{\partial \rho^N}{\partial t} &= \Delta((1 - r_s(\rho^N, \xi_t^N))\rho^N) + F(\rho^N, \xi_t^N), \quad (x, t) \in D_T := D \times (0, T), \\ \rho^N(x, t) &= 0, \quad (x, t) \in \Gamma_T := \partial D \times (0, T), \\ 0 \leq \rho^N(x, 0) &= \rho_0(x) \leq 1, \quad x \in D,\end{aligned}$$

converges as  $N \rightarrow \infty$  into  $\mathcal{H}_T = C(0, T; L^2(D))$  to the unique solution of the average problem

$$\begin{aligned}\frac{\partial \rho}{\partial t} &= \frac{1}{2} \Delta \rho + \bar{F}(\rho), \quad (x, t) \in D_T, \\ \rho(x, t) &= 0, \quad (x, t) \in \Gamma_T, \\ 0 \leq \rho(x, 0) &= \rho_0(x) \leq 1, \quad x \in D.\end{aligned}$$

# Summary

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- Linear stability analysis for spatial-temporal GoG models again reveals the existence of a critical threshold of heterogeneity above which the glioblastoma tumor survives.
- Therefore the provided stochastic models predict that: the higher the heterogeneity of the microenvironment of the cancerous cells is then the higher the probability of survival of the tumor becomes.

# Key References

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Thank you for your attention!