

# Mathematical modeling and analysis of tumor-immune interactions

Kevin Atsou <sup>1</sup>    Thierry Goudon <sup>1</sup>  
Biologists : Véronique Braud <sup>2</sup>    Fabienne Anjuere <sup>2</sup>

<sup>1</sup>Université Côte d'Azur, Inria (Team COFFEE), CNRS, LJAD

<sup>2</sup>CNRS, IPMC (Institut de Pharmacologie Moléculaire et Cellulaire)

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- 1 tumors and effector T-cells interactions : Biological context
- 2 Mathematical modeling : Earlier stages of tumor-growth
- 3 Numerical simulations

# Summary

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- 2 Mathematical modeling : Earlier stages of tumor-growth
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# The Immune system Attacks Tumor cells

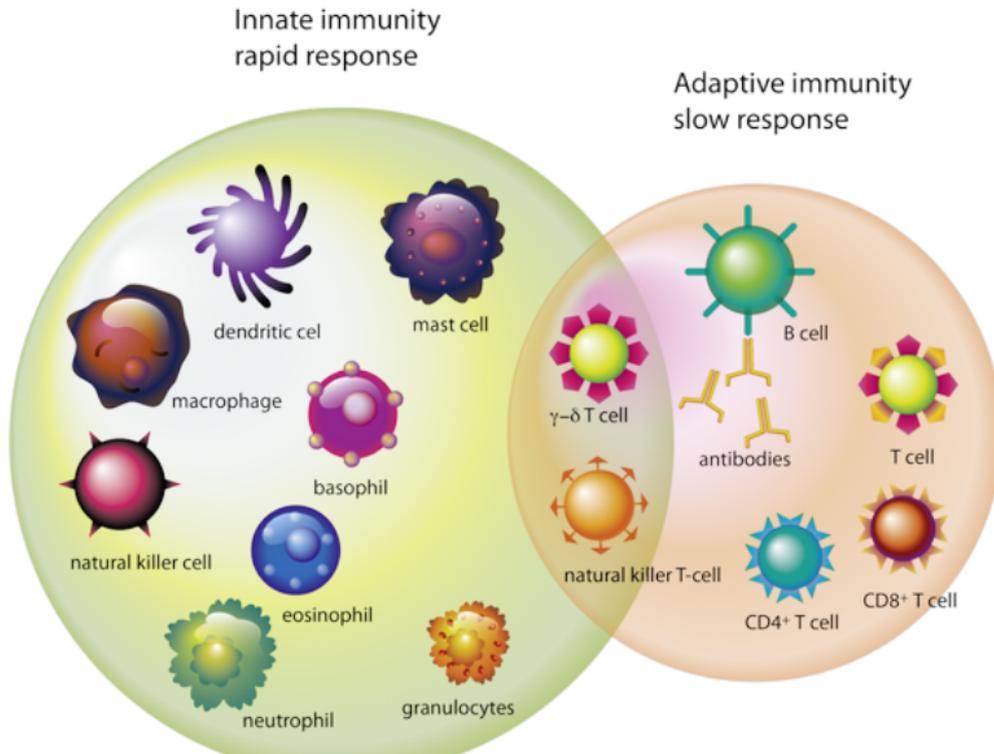
- **Immunotherapy** starts in 1891 with W. Coley (Coley's toxins);
- Immunotherapy restarted seriously With the appearance of AIDS ;
- **immunodeficient** patients often develop cancer
- **Immune cells** play an important role in the control or in the development of tumors

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• source : <https://gifer.com/en/7ftZ>

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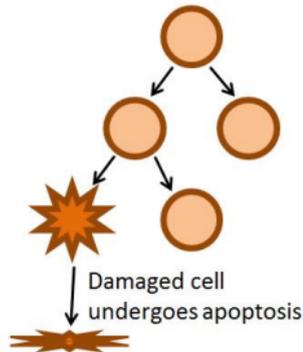
# Two types of Immune system



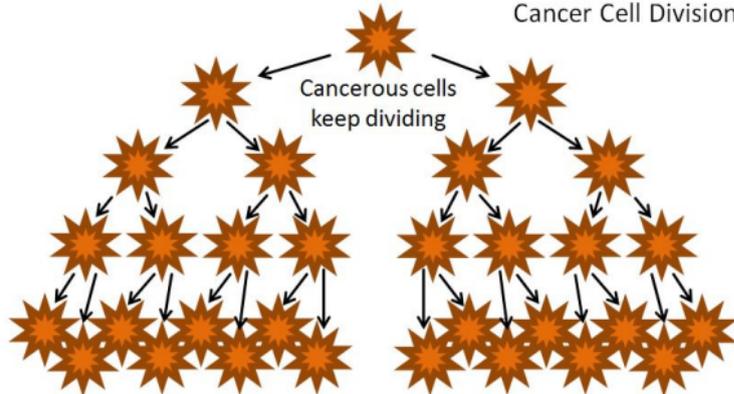
# Tumor cells cycle

- Genes involved in cell cycle control are subject to the **genetic alterations**  $\implies$  **cancer**  $\implies$  **Uncontrolled Cell division**

Normal Cell Division



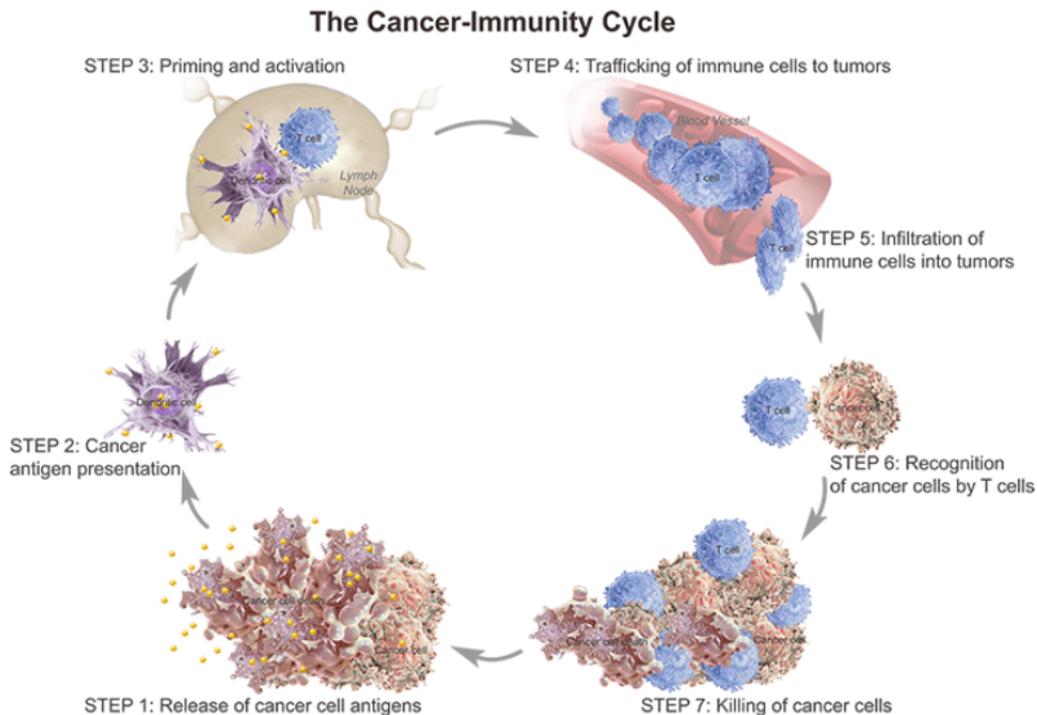
Cancer Cell Division



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3. source : <https://curiousciencewriters.org/>

# Tumor Immunity Cycle



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4. source : <http://biocc.hrbmu.edu.cn/TIP/>

## K.Atsou - Impediments of the tumor immunity cycle

### Tumor has many suppressive influences

The prevalence of cancer indicates that the immune system does not have a strong enough effect on Tumors.

- high levels of **suppressive cytokines** ;
- **T regulatory cells** (Tregs produce  $TGF-\beta$  and IL-10) ;
- high expression of **PD-L1** (Programmed death-ligand 1 ) by tumors ;
- Production of **VEGF** (Vascular Endothelial Growth Factor) for **angiogenesis** ;

# Summary

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- 2 **Mathematical modeling : Earlier stages of tumor-growth**
- 3 Numerical simulations

# A growth-fragmentation modeling approach : Tumor

- $(t, z) \mapsto n(t, z)$  (in  $cell_n \cdot \mu m^{-3}$ ) the size-structured tumor cells distribution
- $(t, x) \mapsto c(t, x)$  (in  $cell_c \cdot mm^{-3}$ ) the concentration of Effector T cells

# Tumor growth main processes

Tumor growth is splitted into two main processes :

- natural (microscopic) growth of the cells,
- cell division.

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• source : <https://gifer.com/en/9uMp>

# A growth-fragmentation modeling approach : Tumor

We model the tumor growth by the following equation :

$$\frac{\partial}{\partial t} n(t, z) = \underbrace{-\frac{\partial}{\partial z} (V(z)n(t, z))}_{\text{the cell's volume growth}} + \underbrace{Q(n(t, z))}_{\text{Cellular division operator}} .$$

We can complete the equation by the **initial data**,

$$n(t = 0, z) = n_0$$

and the **boundary condition**,

$$n(t, 0) = 0$$

# The cellular division

$$\frac{\partial}{\partial t} n(t, z) = \underbrace{-\frac{\partial}{\partial z} (V(z)n(t, z))}_{\text{the cell's volume growth}} + \underbrace{Q(n(t, z))}_{\text{Cellular division operator}} .$$

$Q(n(t, z))$  ?

- $a(z)$ , the **rate at which cells of size  $z$  process division.**
- $k(z'|z)$  **the distribution of products from a cells of size  $z$  dividing.**

# The cellular division

$k(z'|z)$  must be normalized so that mass is conserved :

$$\int_0^z z' k(z'|z) dz' = z. \quad (1)$$

Consequently,

$$Q(n) = \underbrace{-a(z)n(t, z)}_{\text{The loss of cells of size } z} + \underbrace{\int_z^\infty a(z') k(z|z') n(t, z') dz'}_{\text{gain of cells with size } z}. \quad (2)$$

## Binary and symmetric The cellular division

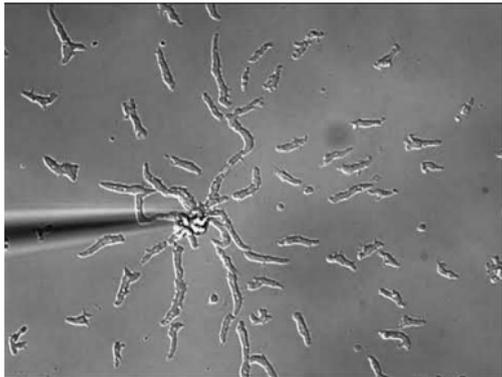
$$Q(n) = \underbrace{-a(z)n(t, z)}_{\text{The loss of cells of size } z} + \underbrace{\int_z^\infty a(z')k(z|z')n(t, z')dz'}_{\text{gain of cells with size } z}. \quad (3)$$

**We assume binary and symmetric division process.** Therefore cells of size  $2z$  give birth to cells of size  $z$ .  $k(z|2z) = 2\delta_{z'=2z}$

$$Q(n(t, z)) = -a(z)n(t, z) + 2a(2z)n(t, 2z)d(2z) \quad (4)$$

$$Q(n(t, z)) = \underbrace{4a(2z)n(t, 2z)}_{\text{gain of cells with size } z} - \underbrace{a(z)n(t, z)}_{\text{The loss of cells of size } z} \quad (5)$$

## Chemotaxis phenomenon



**T** cells displacement towards the tumor microenvironment is modelled by **chemotaxis**.

## K.A - The antigen-specific CD8+ effector T cells displacement

the activated T cells follows the gradient of the chemical signal

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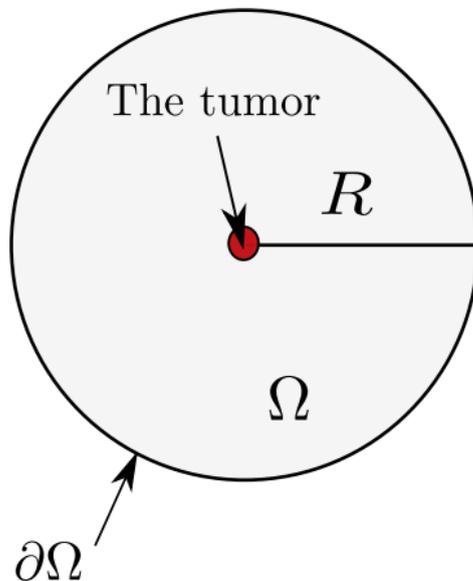
the activated T cells follows the gradient of the chemical signal

Let's denote by

$$\Omega = \{x \in \mathbb{R}^N, |x| \leq R\},$$

$c_\omega(t)$  the time dependent concentration of immune cells in a volume  $\omega \subset \Omega$  :

$$c_\omega(t) = \int_\omega c(t, x) dx \quad (6)$$



## The tumor mass : kth order moments

- Let's denote by  $\mu_0(t)$  (the zeroth order moment) **the total number of tumor cells** in the tumor ;

$$\mu_0(t) = \int_0^{\infty} n(t, z) dz \quad (7)$$

- Let's denote by  $\mu_1(t)$  (the first order moment) **the total volume of the tumor** ;

$$\mu_1(t) = \int_0^{\infty} zn(t, z) dz \quad (8)$$

# The Chemotaxis model

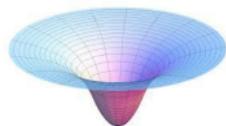
$$\begin{aligned}
 \frac{\partial c}{\partial t}(t, x) + \overbrace{\nabla \cdot (\chi_{\Phi}(\nabla_x \Phi) c)}^{\text{Convection term}} - \overbrace{D(\nabla_x c)}^{\text{Diffusion term}} = \overbrace{p_f g(\mu_1) S}^{\text{Conversion of Im. cells}} - \overbrace{\gamma c}^{\text{Natural Death}} \\
 c|_{\partial\Omega} = 0
 \end{aligned}
 \tag{9}$$

Typical conversion law of Naive T cells (*Saturated conversion*).

$$g(\mu_1) = \frac{\mu_1}{\delta + \mu_1}
 \tag{10}$$

# Tumor antigen-induced potential

The attractive potential  $\Phi$  satisfies a **Poisson equation** :



$$-\nabla \cdot (k(x)\nabla\Phi) = \langle\sigma\rangle \int_0^\infty zn(t,z)dz,$$

$$-\nabla \cdot (k(x)\nabla\Phi) = \mu_1(t)\langle\sigma\rangle,$$

With **Neumann homogeneous boundary condition**.

$$k(x)\nabla_x\Phi(x) \cdot n_{\partial\Omega} = 0$$

where  $\langle\sigma\rangle = \sigma - \frac{1}{|\Omega|} \int_\Omega \sigma dx$

# A chemotaxis model : Immune cells

## The resulting Chemotaxis model

$$\left\{ \begin{array}{l}
 \partial_t c(t, x) + \nabla \cdot (\chi(\phi)(\nabla_x \phi)c - (D(x)\nabla_x c)) = pg(\mu_1)S - \gamma c, \quad \forall t > 0, x \in \Omega \\
 -\nabla \cdot (k(x)\nabla \phi) = \int_0^\infty zn(t, z)\sigma(x, z)dz, \quad \forall t, z > 0, x \in \Omega \\
 c = 0, \quad \forall x \in \partial\Omega, \\
 c(t = 0, x) = c_0(x), \quad \forall x \in \Omega, \\
 k(x)\nabla \phi \cdot n_{\partial\Omega} = 0 \quad \forall x \in \partial\Omega.
 \end{array} \right.$$

## Tumor vs Immune cells : the interactions term

Let's denote by  $m(c, n)$  the death term describing the immune response. The tumor growth model becomes :

$$\frac{\partial}{\partial t} n(t, z) = -\frac{\partial}{\partial z} (Vn(t, z)) + Q(n(t, z)) - \underbrace{m(c, n)} .$$

The interaction Tumor-Im. Cells

## Tumor vs Immune cells : the interactions term

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$$m(c, n)(t) = \int_{\Omega} \delta(y) c(t, y) dy \times \frac{n(t, z)}{\alpha + n(t, z)} .$$

where  $\delta$  is an **interaction kernel**

## Asymptotic behavior : Tumor growth

We turn back to the tumor growth equation without immune response :

$$\begin{cases} \partial_t n(t, z) = -\partial_z(V(z)n(t, z)) - a(z)n(t, z) + 4a(z)n(t, 2z), & \forall (t, z) \in \mathbb{R}_+^* \times \mathbb{R}_+ \\ n(t, 0) = 0, & \forall t \geq 0, \\ n(t=0, z) = n_0(z) \in L^1(\mathbb{R}_+), & \forall z \geq 0, \end{cases}$$

The solution behaves like :  $(t, z) \mapsto \rho e^{\lambda t} N(z)$  with  $\rho = \int_{\mathbb{R}_+} n_0(z)$

## Asymptotic behavior : Tumor growth

The long time behavior of  $n$  is directly linked with the existence of a solution  $\lambda, N, \phi$ , in  $L^1$  sense, of the associated eigenproblem,

$$\left\{ \begin{array}{l} \partial_z(V(z)N(z)) + (\lambda + a(z))N(z) = 4a(z)N(2z), \quad z \geq 0 \\ N(0) = 0, \quad N(z) > 0 \text{ for } z > 0, \quad \int_0^\infty N(z) dz = 1 \\ V(z)\partial_z\psi(z) - (\lambda + a(z))\psi(z) = -2a(z)\psi\left(\frac{z}{2}\right), \quad x \geq 0 \\ \psi(z) > 0 \text{ for } x \geq 0, \quad \int_0^\infty N(z)\psi(z) dz = 1 \end{array} \right.$$

# Asymptotic behavior : Tumor growth

The solution of the dual equation allows to define a conservation law on  $n(t, z)$  :

## Mass Conservation

$$\int_0^\infty n(t, z) e^{-\lambda t} \psi(z) dz = \int_0^\infty n_0(z) \psi(z) dz$$

# Existence and uniqueness of the eigenelements

Theorem, B. Perthame et al., 2005

if **the division rate  $a$  is lower bounded and upper bounded**, then **there is a unique solution  $(\lambda, N, \psi)$**  to the eigenproblem with  $\psi, N$  in  $C^1(\mathbb{R}_+)$  and moreover, **all the moments of the asymptotic state  $N$  are bounded**.

The existence of the eigenelements is based on the **Krein-Rutman Theorem**

# Asymptotic behavior : Tumor growth

**Exponential decay to the leading eigenpair of the growth-fragmentation operator :**

Theorem, B. Perthame et al., 2004

If the **division rate** and the **growth rate of the tumor** are **constant**, there is a **unique solution** to **the first eigenvalue problem**. This solution belongs to the Schwartz space and furthermore, **all the solutions to the tumor growth equation converges exponentially to the first eigenvector** with a rate equal to the tumor division rate.

## Asymptotic behavior : Tumor growth

There is a semi-explicit formula for the leading eigenfunction  $N$  (F. Baccelli, 2002, B. Perthame et al., 2004 ) given by a series that **converges absolutely and uniformly**

Lemma, F.Baccelli, 2002

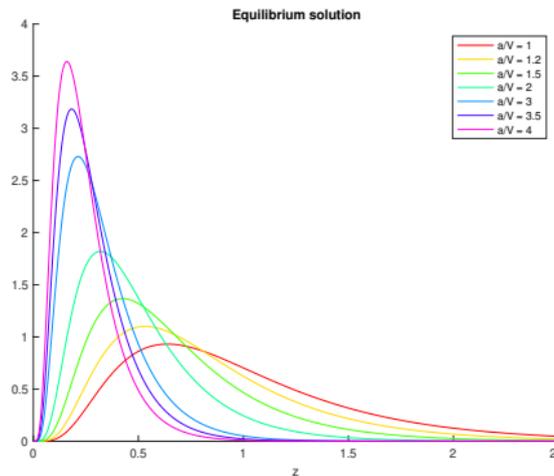
Let  $\alpha_0 = 1$ ,  $\alpha_n = \frac{2}{2^n - 1} \alpha_{n-1}$ , then the function

$$N(z) = \langle N \rangle \sum_{n=0}^{\infty} (-1)^n \alpha_n \exp \left( -2^{n+1} \frac{a}{V} z \right),$$

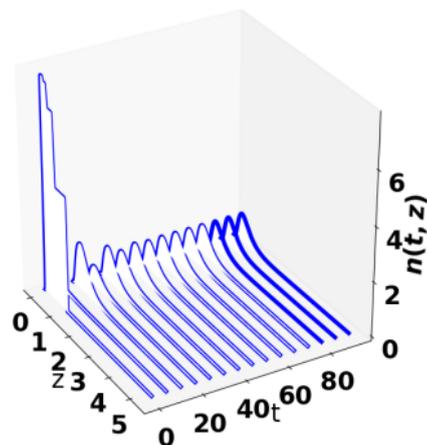
belongs to  $\mathcal{S}(\mathbb{R}^+)$  and is unique.

# Asymptotic behavior : Tumor growth

different profiles :



An example of time evolution :



# Asymptotic behavior : Tumor growth

Existence and uniqueness of eigenelements has been proved

## Residual distribution

- the system reaches a certain **equilibrium state**, with a non zero tumor cells distribution
- In which cases the immune system can control the tumor progression ?

## Asymptotic behavior : Immune cells taking control

An attempt which clarifies this issue is provided by the following statement for small division rates (non-aggressive tumors)

Proposition, Kevin ATSOU & Thierry Goudon

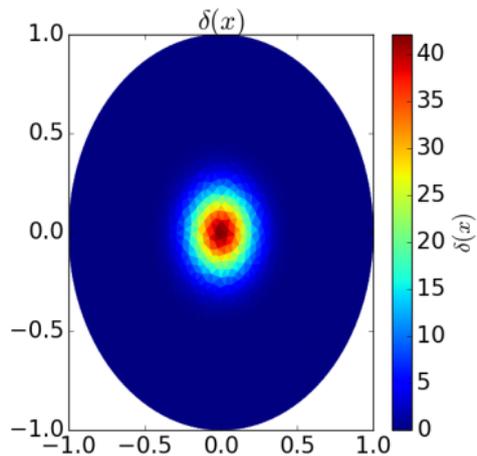
If the tumor is **non-aggressive**, there exist a **unique tumor mass** which depends on the tumor division rate at which the tumor is always **controlled by the immune cells**.

# Summary

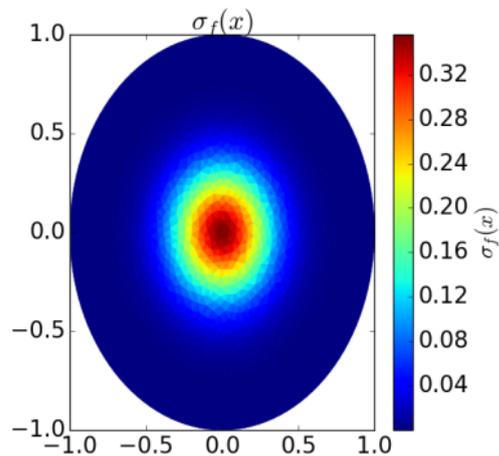
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# model parameters

- Variable parameters :



(a) The T cells Strength  $\delta$

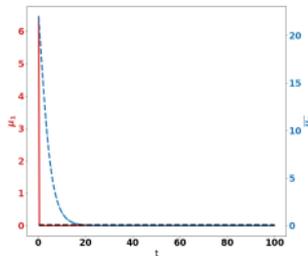


(b) The chemical signal  $\sigma_f$

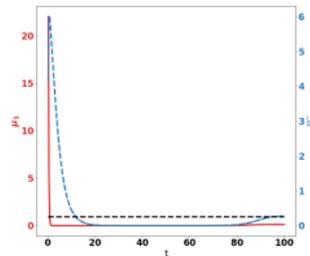
# Homogeneous distribution of Naive cells : $a = 1$

Tumor vs Homogeneous distribution of Immune cells :

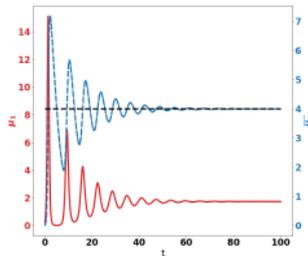
# Homogeneous distribution of Naive cells : $a$



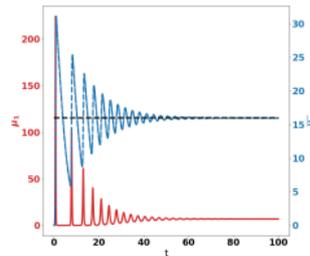
(c)  $a = 0.0625$



(d)  $a = 0.25$



(e)  $a = 4$

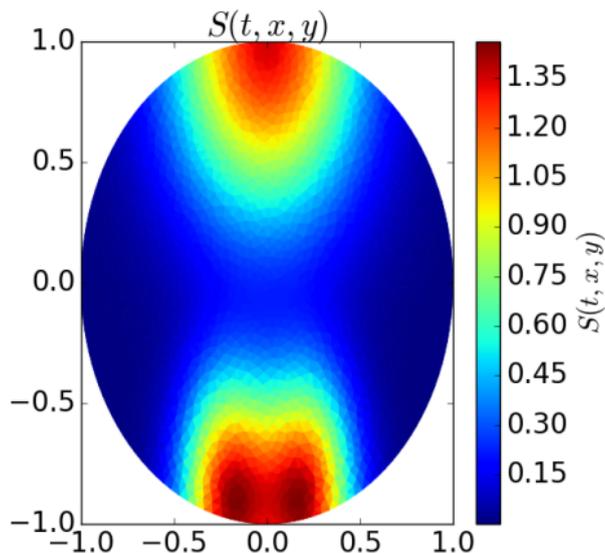


(f)  $a = 16$

FIGURE – behavior of the solutions. ( $\mu_1$  in red,  $\mu_c$  in blue,  $a$  in black)

# Heterogeneous distribution of Naive cells

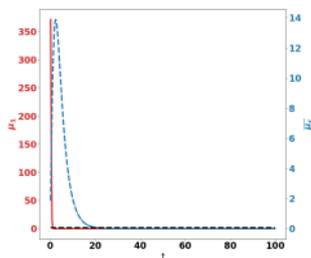
Tumor vs Heterogeneous distribution of Immune cells :



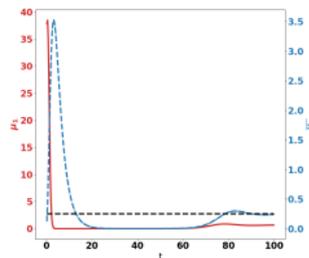
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Tumor vs Heterogeneous distribution of Immune cells :

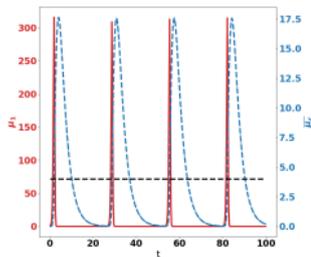
# Heterogeneous distribution of Naive cells



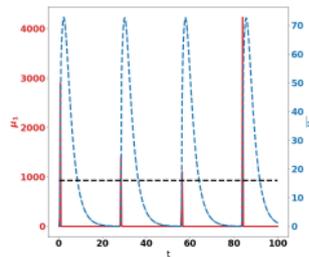
(a)  $a = 0.0625$



(b)  $a = 0.25$



(c)  $a = 4$



(d)  $a = 16$

**FIGURE** – Evolution of the tumor mass  $\mu_1$  (red curves, left axis), and of

Thank you for your attention...