



Optimal schedules for therapies in metastatic cancers

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ECMTB, Krakow, June 28, 2011

Clinical problematics

Clinical problematics

- In the case of breast cancer, 10% of cancers in stage T1N0M0 give rise to metastasis.
Elias (2006), Spielmann & al (2006)
- Micrometastases (size $< 10^8$ cells \simeq 100 mg) invisible with imaging techniques. How to administer adjuvant therapy (after surgery) without seeing anything?
- What is the best **scheduling** for : chemotherapy (CT), anti-angiogenics (AA), CT + AA?

The tool : a "simple" modelling (with **few parameters**)

- **tumoral growth model (ODE)**
- **renewal model for the metastases (PDE)**

Outline

- 1 A model for metastatic evolution
- 2 An optimal control problem for the metastases
 - Formulation of an optimal control problem
 - Theoretical study
- 3 Numerical study in a simplified case
 - First examples
 - Optima comparison in a two-dimensional case

ODE model of tumoral growth under angiogenic control

Hahnfeldt et al., Cancer Research 1999

Gompertzian growth x = Size of the tumor

$$\frac{dx}{dt} = ax \ln\left(\frac{\theta}{x}\right)$$

Consider θ as a **variable** : the **vascular capacity**

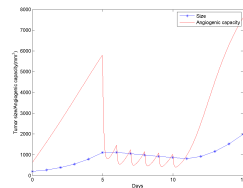
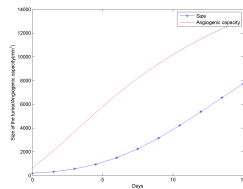
$$\frac{d\theta}{dt} = \underbrace{cx}_{\text{Stimulation by the tumor}} - \underbrace{d\theta x^{\frac{2}{3}}}_{\text{Inhibition}}$$

Hahnfeldt model and CT/AA combination

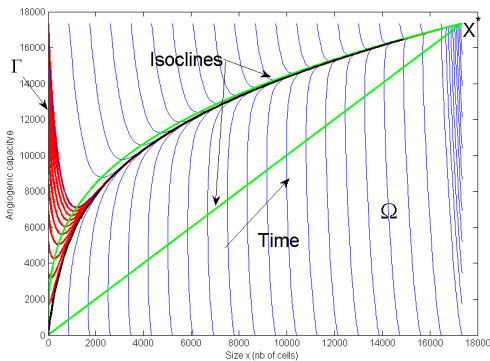
$$\frac{dx}{dt} = ax \ln\left(\frac{\theta}{x}\right) - f u^1(t)(x - x_{min})^+$$

$$\frac{d\theta}{dt} = cx - d\theta x^{\frac{2}{3}} - e u^2(t)(\theta - \theta_{min})^+$$

- Log-kill term of the chemotherapy
- AA drugs impact on the tumoral vasculature

Studied in **d'Onofrio, Gandolfi, 2004**

Transport equation for the metastases population



$$\Omega = \left] 1, \left(\frac{c}{d} \right)^{\frac{3}{2}} \right[$$

$$G(X) = G(x, \theta) = \begin{pmatrix} ax \ln \left(\frac{\theta}{x} \right) \\ cx - d\theta x^{\frac{3}{2}} \end{pmatrix}$$

$$u(t) = (u^1(t), u^2(t))$$

$$\bar{G}(t, X; u) = G(X) - B(X)u(t)$$

$$\frac{dX}{dt} = G(X)$$

All the tumors follow the ODE model.

Population of the metastases structured in size x and vascular capacity θ
density $\rho(t, x, \theta) \in L^1(\Omega)$.

Balance law :

$$\partial_t \rho + \operatorname{div}(\rho \bar{G}) = 0$$

Boundary condition. Birth of new metastases

We assume

- That the metastases are born with size 1
- **Independance** between the vascular capacity of the neo-metastasis and the mother-tumour which emitted it

Birth rate of new metastases of parameter σ per meta of size x and angiogenic capacity θ per unit of time :

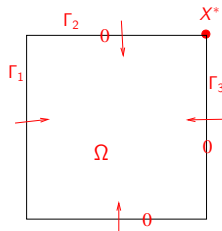
$$B(\sigma, x, \theta) = N(\sigma)\beta(x, \theta), \quad \sigma \in \partial\Omega$$

We choose :

$$N(\sigma) = \frac{1}{2\Delta\sigma} \mathbf{1}_{\sigma \in [\sigma_0 - \Delta\sigma, \sigma_0 + \Delta\sigma]}$$

$$\beta(x, \theta) = mx^\alpha$$

Iwata et al., J. Theor. Biol., 2000



Two sources of new metastases :

- Primitive tumor $X_p(t)$ with $\frac{dX_p}{dt} = \overline{G}(X_p) : N(\sigma)\beta(X_p(t))$
- Metastases themselves : $N(\sigma) \int_{\Omega} \beta(x, \theta) \rho(t, x, \theta) dx d\theta$

Metastatic model

$$(E) \begin{cases} \partial_t \rho(u) + \operatorname{div}(\overline{G}(u)\rho(u)) = 0 &]0, \infty[\times \Omega \\ -\overline{G}(t, \sigma; u) \cdot \nu(\sigma)\rho(t, \sigma; u) = N(\sigma) \left\{ \int_{\Omega} \beta(X)\rho(t, X; u) dX + \beta(X_p(t; u)) \right\} & \partial\Omega \\ \rho(0) = \rho^0 & \Omega \end{cases}$$

$$\overline{G}(t, X; u) = G(X) - B(X)u(t), \quad u(t) = (u^1(t), u^2(t))$$

Number of metastases : $\int_{\Omega} \rho(t, X) dX$. **Metastatic mass :** $\int_{\Omega} x\rho(t, X) dX$

- Linear transport equation in dimension 2 with nonlocal boundary condition.
- Theoretical and numerical analysis has been performed B., J. Evol. Equ., 2011
B., M2AN, 2011
Barbolosi, Benabdallah, Hubert, Verga, Math. Biosc., 2008
- Original idea for a structured population equation for the metastases Iwata et al., J. Theor. Biol.

- ① A model for metastatic evolution
- ② An optimal control problem for the metastases
 - Formulation of an optimal control problem
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Formulation of an optimal control problem

On the primary tumor growth : $\dot{X}_p(t; u) = G(X_p(t; u)) - B(X_p(t; u))u(t)$

- Already studied

A. d'Onofrio, U. Ledzewicz, H. Maurer and H. Schättler, *Math. Biosc.*, 2009

A. Ergun, K. Camphausen, L. M. Wein, *Bull. Math. Biol.*, 2003 (radiotherapy)

- Two possible criteria to be minimized for the primary tumor size

$$J_T(u) = x_p(T; u) \quad \text{and} \quad J_m(u) = \min_{t \in [0, T]} x_p(t; u)$$

- Toxicity constraints

$$\mathcal{U}_{ad} = \left\{ u \in (L^\infty(0, T))^2; \begin{pmatrix} 0 \\ 0 \end{pmatrix} \leq u(t) \leq \begin{pmatrix} v_{max} \\ u_{max} \end{pmatrix} \forall t \text{ and } \int_0^T u(t) dt \leq \begin{pmatrix} C_{max} \\ A_{max} \end{pmatrix} \right\}$$

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On the metastases

$$J(u) = \underbrace{\int_{\Omega} \rho(T, X; u) dX}_{\text{Total number of metastases}} \quad \text{and} \quad J_M(u) = \underbrace{\int_{\Omega} x \rho(T, X; u) dX}_{\text{Metastatic mass}}$$

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Is there a difference in the optimal minimizer u^* for J_T , J_m , J , J_M ?

Existence of an optimal solution

Theorem

Under some regularity assumptions **there exists** $(u^*, u_M^*) \in \mathcal{U}_{ad}$ such that

$$J(u^*) \leq J(u), \quad \forall u \in \mathcal{U}_{ad}, \quad J_M(u_M^*) \leq J_M(u), \quad \forall u \in \mathcal{U}_{ad}$$

The proof is based on the following proposition

Proposition

Under some regularity assumptions if $\rho(u)$ is the solution of (E), then $\rho \in W^{1,\infty}(Q)$ and there exists a continuous function C which can be explicited in terms of $\|\beta\|_{W^{1,\infty}(\Omega)}$, $\|N\|_{W^{1,\infty}(\partial\Omega)}$, $\|G\|_{L^\infty(\Omega)}$ and $\|B\|_{L^\infty(\Omega)}$ such that, for all $u \in \mathcal{U}_{ad}$

$$\|\rho(u)\|_{W^{1,\infty}(Q)} \leq C(\|u\|_{L^\infty(Q)})$$

Optimality system for J

In the case of $J(u) = \int_{\Omega} \rho(T, X; u) dX$ and without the source term in the boundary condition

Proposition

Let u^* be a solution of the optimal control problem. We have the following **optimality system** :

$$\left\{ \begin{array}{l} \partial_t \rho^* + \operatorname{div}(\rho^* \overline{G}(u^*)) = 0 \\ -G \cdot \nu(t, \sigma; u^*) \rho^*(t, \sigma; u^*) = N(\sigma) \left\{ \int_{\Omega} \beta(X) \rho^*(t, X; u^*) dX + \beta(X_p(t; u^*)) \right\} \\ \rho^*(0, X; u^*) = \rho^0 \end{array} \right.$$

$$\left\{ \begin{array}{l} -\partial_t \rho^*(t, X; u^*) - \overline{G}(X; u^*) \nabla \rho^*(t, X; u^*) - \beta(X) \int_{\partial\Omega} N(\sigma) \rho^*(t, \sigma) d\sigma = 0 \\ \rho^*(T) = -1. \end{array} \right.$$

$$\int_0^T \int_{\Omega} \rho^* \operatorname{div}(\rho^* B(X) \cdot (v - u^*)) dX dt \leq 0, \quad \forall v \in \mathcal{U}_{ad}.$$

- ① A model for metastatic evolution

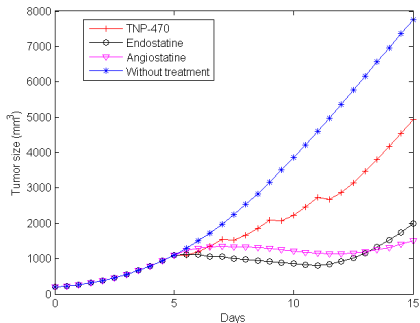
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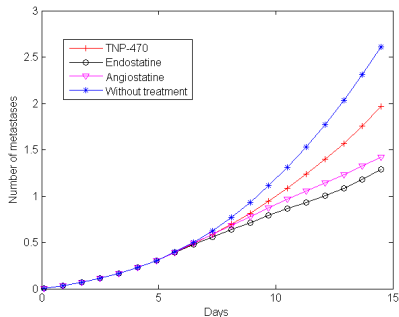
Anti-angiogenic therapy

Testing the drugs from **Hahnfeldt et al., Cancer Res. 99** (mice data) :

- Endostatine 20 mg/kg/day
- Angiostatine 20 mg/kg/day
- TNP-470 30 mg/kg/q.o.d



Primary tumor growth

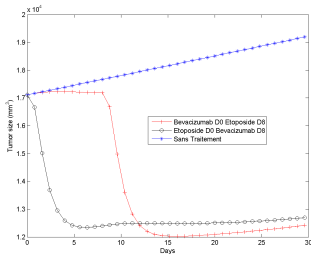


Metastatic evolution

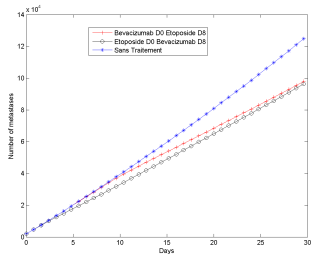
Combination of cytotoxic and anti-angiogenic therapy

Human parameters. Etoposide/Bevacizumab combination. Order of administration?

- Bevacizumab D0 Etoposide D8 VS Etoposide D0 Bevacizumab D8



Tumoral growth



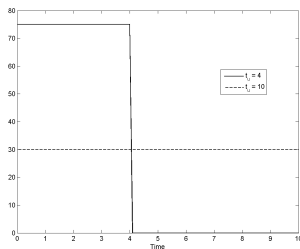
Total number of metastases

The best protocol is **not the same** for the primary tumor and for the number of metastases.

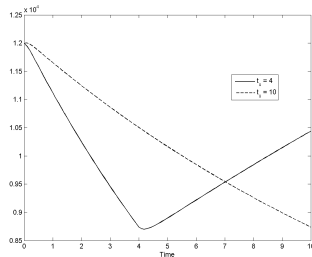
Optima comparison in a two-dimensional case

Administer total given amounts of agents (C_{max} , A_{max}) from time 0 to times (t_v, t_u) at constant rates $V = \frac{C_{max}}{t_v}$ and $U = \frac{A_{max}}{t_u}$.

U. Ledzewicz et al., *Math. Medic. and Biol.*, 2010



Examples of administration of the AA drug

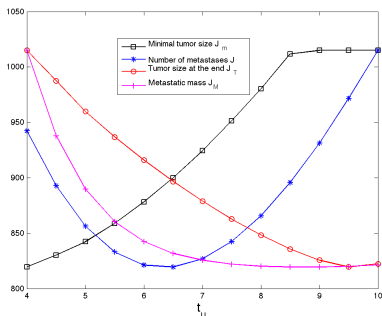
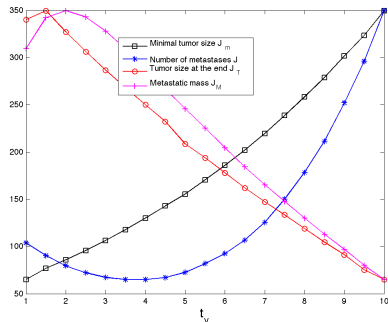


Primary tumor evolution

$$U_{ad} = \{u^1(t) = \frac{C_{max}}{t_v} \mathbf{1}_{[0, t_v]}(t), u^2(t) = \frac{A_{max}}{t_u} \mathbf{1}_{[0, t_u]}(t), \left(\frac{C_{max}}{v_{max}}, \frac{A_{max}}{u_{max}} \right) \leq (t_v, t_u)\}.$$

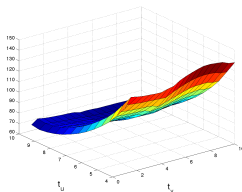
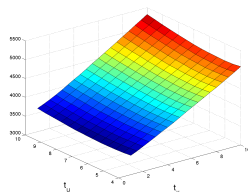
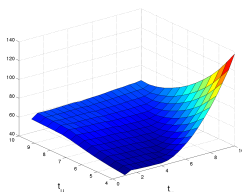
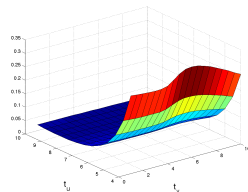
$$U_{ad} \simeq \left[\frac{C_{max}}{v_{max}}, T \right] \times \left[\frac{A_{max}}{u_{max}}, T \right] = [1, 10] \times [4, 10]$$

Monotherapy cases

AA alone. $C_{max} = 0$ CT alone. $A_{max} = 0$.

- Criteria J_T , J_m and J give **different optimal values**.
- Metastatic mass J_M gives the same optimal value as J_T .
- Difference between CT and AA : shape of J_M .

CT-AA combination

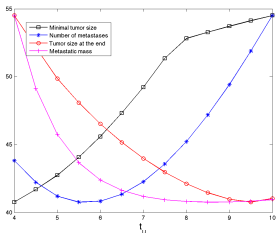
Tumor size J_T Number of metastases J Minimal tumor size J_m Metastatic mass J_M

Criterion	J_T	J_m	J	J_M
(t_v^*, t_u^*)	(9.5, 9.5)	(1, 4)	(1, 5.5)	(10, 9)

- We can regroup J and J_m under the **strong dose/short time** strategy and J_T and J_M under the **low dose/large time** one.

Influence of the presence of a drug on the behavior of the other

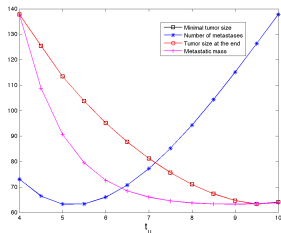
$$t_V = 1$$



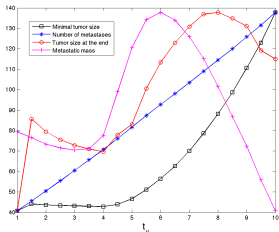
Effect of CT on AA

J_T , J and J_M are almost identical but J_m has the reverse behavior

$$t_V = 10$$



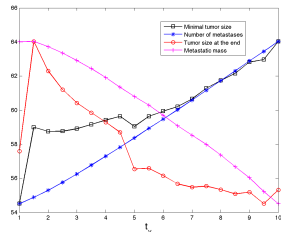
$$t_U = 4$$



Effect of AA on CT

Drastical changes for J_T and J_M . J is stable

$$t_U = 10$$



Conclusion

- Simple model for metastatic evolution
- Difference of the optimal solution between the primary tumor and the metastases.
- Necessity to define precisely the objective(s) to be minimized.

How to **cleverly combine** tumoral and metastatic reduction?

Use the **metastatic mass** $J_M = \int_{\Omega} x\rho(t, x, \theta) dx d\theta$?

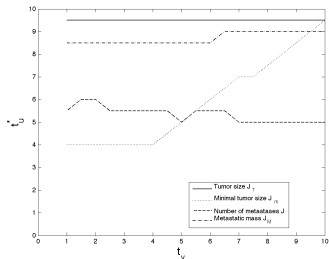
Linear combination between a tumoral criterion and the number of metastases?

Perspectives

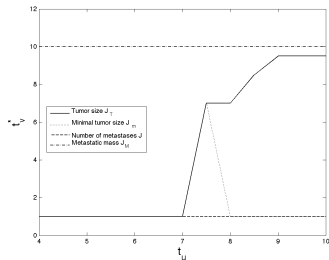
- Numerical method for the infinite-dimensional optimal control problem on the metastases (PDE)
- Further theoretical study of the theoretical optimality system

CT-AA combination

CT-AA combination

Graphs of $t_v \mapsto \arg\min_{t_u} J_X(t_u, t_v)$

- For J_m : **synchronization effect**
- J , J_M and J_T are stable

Graphs of $t_u \mapsto \arg\min_{t_v} J_X(t_u, t_v)$

- For J_T : change in the optimal value t_v^* .
- J , J_M and J_m are stable