

# State-Constrained Optimal Control Applied to Cell-Cycle-Specific Cancer Chemotherapy

Mazen Alamir and Sophie Chareyron

Laboratoire d'Automatique de Grenoble

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# Introduction

The optimal way to administer drugs in cancer treatment chemotherapy remains an open issue because of **a lack of unified models**. This is due to:

- the existence of many types of cancer
- the existence of many different families of drugs

**Our focus :** cell-cycle-specific chemotherapeutic drugs that act as killing agent

# Outlines

- Part I: Model and problematic presentation
- Part II: Preliminary results
- Part III: Main results
- Part IV: Numerical experiments
- Part V: Conclusion

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# Part I

## Model and problematic

# Cell-cycle dynamics

Dynamics of the cell-cycle of the bone-marrow cell transition between the proliferating phase and the rest phase:

$$\dot{P} = (\gamma - \delta - \alpha - f(t))P + \beta Q \quad (1)$$

$$\dot{Q} = \alpha P - (\lambda + \beta)Q \quad (2)$$

with

- $P$  number of cells in the proliferating phase
- $Q$  number of cells in the resting phase



K. R. Fister and J. C. Panetta *Optimal Control Applied to Cell-Cycle-Specific Cancer Chemotherapy*. *SIAM Journal on Applied Mathematics*, 60(3):1059-1072, 2000.

# Cell-cycle dynamics

$f$  : **control action** describing the effects of the chemotherapeutic treatment with  $f(t) \in [0, 1]$

- $f(t) = 0$  no drug injected at time  $t$
- $f(t) = 1$  maximal rate of drug is used

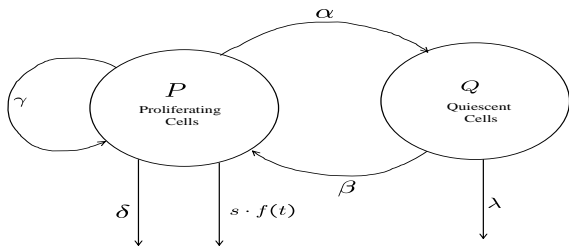


Figure: Schematic view of the cell cycle

**Design optimal control** to maximize the quantity of drug injected over the treatment period  $T$  while continuously respecting the following inequality constraint on the state trajectory

$$P(t) + Q(t) \geq \rho \quad \forall t \in [0, T]$$

Through the use of weighting parameters



K. R. Fister and J. C. Panetta *Optimal Control Applied to Cell-Cycle-Specific Cancer Chemotherapy*. *SIAM Journal on Applied Mathematics*, 60(3):1059-1072, 2000.



U. Ledzewicz and H. Schättler *Optimal Bang-Bang Controls for a Two-Compartment Model in Cancer Chemotherapy*, *Journal of Optimization Theory and Applications*, 114(3):609-637, 2002.

# Unconstrained optimal control problems

We propose to design **a family of unconstrained optimal control problems** that approximate to any desired precision the original constrained optimal control problem given by :

$$P_\rho(P_0, Q_0) : \min_{f(\cdot)} [J(f)] = \frac{1}{T} \int_0^T (1 - f(t))^2 dt$$

under  $f(t) \in [0, 1]$  and  $P(t) + Q(t) \geq \rho \quad \forall t \in [0, T]$  (3)

where  $(P(t), Q(t))$  solution of (1)-(2) under the control profile  $f(\cdot)$  starting from the initial condition  $(P_0, Q_0)$



## Part II

### Preliminary results

## Definition

Given some initial conditions  $(P_0, Q_0)$ , a control profile  $f(\cdot) \in [0, 1]^{[0, T]}$  leading to a state trajectory that meets the state constraint  $P(t) + Q(t) \geq \rho$  on  $[0, T]$  (for all  $t$ ) is said to be an **admissible profile for the optimal control problem**  $P_\rho(P_0, Q_0)$  defined by (3)

**But a solution of  $P_\rho(P_0, Q_0)$  may not exist for any pair  $(T, \rho)$**

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## Proposition

A **necessary and sufficient condition** for  $P_\rho(P_0, Q_0)$  to admit a non empty set of admissible profiles is that  $f \equiv 0$  is an admissible profile for the optimal control problem  $P_\rho(P_0, Q_0)$ , namely

$$\rho \leq \rho_{\min}(P_0, Q_0, T) := \min_{t \in [0, T]} \left[ C e^{A_0 t} \begin{pmatrix} P_0 \\ Q_0 \end{pmatrix} \right] \quad (4)$$

where  $C := \begin{pmatrix} 1 & 1 \end{pmatrix}$ .

Extended dynamical system  $\Sigma_{r_0}$  :

$$\Sigma_{r_0} \quad \dot{P} = (\gamma - \delta - \alpha - f(t))P + \beta Q \quad (5)$$

$$\dot{Q} = \alpha P - (\lambda + \beta)Q \quad (6)$$

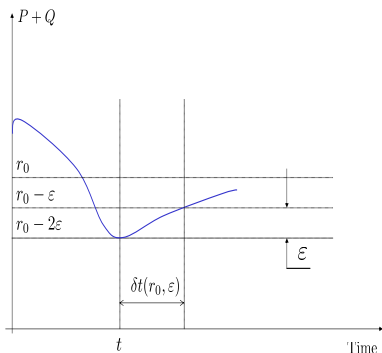
$$\dot{R} = \varphi\left(r_0 - (P + Q)\right) \quad ; \quad R(0) = 0 \quad (7)$$

where  $\varphi : \mathbb{R} \rightarrow \mathbb{R}_+$  is given by :

$$\varphi(r) = \max(0, r) \frac{a}{b+r} \quad ; \quad a > 0 \quad b > 0$$

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# Properties of the extended dynamical system



**Figure:** If at some instant  $t$ ,  $P(t) + Q(t) \leq r_0 - 2\varepsilon$ , then there is a computable duration  $\delta t(r_0, \varepsilon)$  during which  $P + Q$  is lower than  $r_0 - \varepsilon$ .

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## LEMMA

For any  $r_0 > 0$ , any  $\varepsilon \leq r_0/2$  and any control profile  $f(\cdot) \in [0, 1]^{[0, T]}$ , if at some time  $t$ , the system trajectory satisfies

$r_0 - (P(t) + Q(t)) = 2\varepsilon$ , then one has

$$r_0 - (P(\tau) + Q(\tau)) \geq \varepsilon \quad (8)$$

$$\forall \tau \in [t, t + \delta t(r_0, \varepsilon)] \quad (9)$$

where  $\delta t(r_0, \varepsilon) = \frac{1}{\gamma} \ln\left(\frac{r_0 - \varepsilon}{r_0 - 2\varepsilon}\right)$ .

# Indicator on the state constraint violation

## LEMMA

Let  $f \in [0, 1]^{[0, T]}$  be any control profile. Denote by  $P(\cdot)$ ,  $Q(\cdot)$  and  $R(\cdot)$  the corresponding state trajectories of the extended system (5)-(7). Let  $\mu$  be given by

$$\mu := \min_{t \in [0, T]} [P(t) + Q(t)] \quad (10)$$

then the following inequality holds

$$R(T) \geq G(r_0, \mu) := \frac{1}{\gamma} \ln\left(\frac{r_0 + \mu}{2\mu}\right) \varphi\left(\frac{r_0 - \mu}{2}\right) \quad (11)$$

Moreover,  $R(T) = 0$  if and only if  $\mu \geq r_0$ .

**Key property:**  $R(T)$  a relevant indicator on the state constraint violation over the treatment period  $[0, T]$

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## Part III

# Main results

# Computable approximate solution

**AIM:** find approximate solutions of the state-constrained optimization problem

$$P_\rho(P_0, Q_0) \quad : \quad \min_{f(\cdot)} [J(f)] = \frac{1}{T} \int_0^T (1 - f(t))^2 dt$$

under  $f(t) \in [0, 1]$  and  $P(t) + Q(t) \geq \rho \forall t \in [0, T]$  (12)

in which

$$\rho = \rho_{\min}(P_0, Q_0, T) - \eta_0; \quad \eta_0 > 0. \quad (13)$$

$\rho_{\min}(P_0, Q_0, T)$  is explicitly computable using (4) so that for any given  $\rho$ ,  $\eta_0$  is also computable

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# Family of state-unconstrained optimal control problem

For  $\eta \in (0, \eta_0)$ , the **family of state-unconstrained optimal control problem** is defined on the extended system  $\Sigma_{\rho+\eta}$  as follows:

$$P_{\rho}^{\eta}(P_0, Q_0) : \min_{f(\cdot) \in [0,1]^{[0,T]}} \frac{R(T)}{G(\rho + \eta, \rho)} + \frac{1}{T} \int_0^T (1 - f(t))^2 dt$$

with  $r_0 = \rho + \eta$  in (7) (14)

where  $R(T)$ : solution at  $t = T$  of the extended system (5)-(7) starting from the initial condition  $(P_0, Q_0, 0)$  at  $t = 0$  under the control profile  $f(\cdot)$  where  $G(\cdot, \cdot)$  is defined by (11)

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# Properties of the state-unconstrained optimal control problem

**For the optimal control problem  $P_\rho^\eta(P_0, Q_0)$ , we have:**

- existence of a solution of  $P_\rho^\eta(P_0, Q_0)$

## Proposition

*For all  $\eta \in (0, \eta_0]$ , the problem  $P_\rho^\eta(P_0, Q_0)$  admits a solution.*

- an optimal solution of  $P_\rho^\eta(P_0, Q_0)$  is an admissible profile for the original problem  $P_\rho(P_0, Q_0)$  for any value of  $\eta$ , thus  **$\eta$  is not a weighting coefficient**
- a lower and an upper bound on the exact solution of the constrained problem may be obtained **by solving only unconstrained problems**

# Properties of the state-unconstrained optimal control problem

## Proposition

- 1 For all  $\eta \in (0, \eta_0]$ , an optimal solution of the state-unconstrained problem  $P_\rho^\eta(P_0, Q_0)$  is an admissible profile for the original state-constrained problem  $P_\rho(P_0, Q_0)$ .
- 2 If  $\hat{J}_\rho$  [resp.  $\hat{f}_\rho^\eta$ ] denotes the minimal cost of the state-constrained problem  $P_\rho(P_0, Q_0)$  [resp. the state-unconstrained problem  $P_\rho^\eta(P_0, Q_0)$ ], then

$$\hat{J}_\rho \leq \frac{1}{T} \int_0^T \left(1 - \hat{f}_\rho^\eta(t)\right)^2 dt \leq \hat{J}_{\rho+\eta} \quad (15)$$

- 3 In particular, a lower and an upper bound on the exact solution of the constrained problem may be obtained by solving only unconstrained problems, namely

$$\frac{1}{T} \int_0^T \left(1 - \hat{f}_\rho^\eta(t)\right)^2 dt \leq \hat{J}_\rho \leq \frac{1}{T} \int_0^T \left(1 - \hat{f}_\rho^\eta(t)\right)^2 dt \quad (16)$$

## Part IV

# Numerical experiments

# Solutions of unconstrained optimal control problem

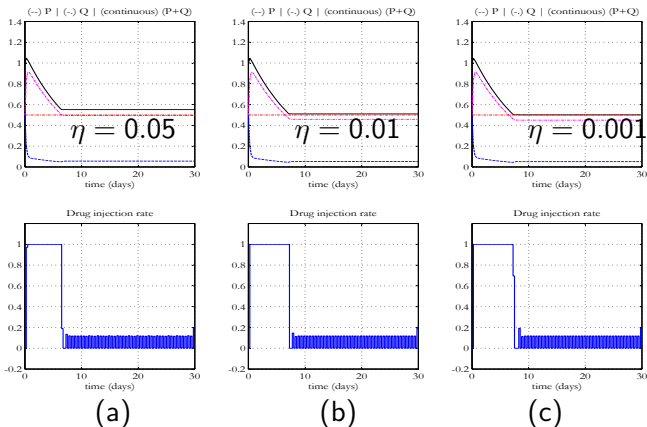


Figure: Solutions of the unconstrained optimal control problems  $P_\rho^\eta(0.5, 0.5)$  for different values of the parameter  $\eta$  and using the initialization  $f(\cdot) \equiv 0$

## Those simulations illustrate the following:

- the solutions of the unconstrained problems can approximate the state constraint problem to **any desired precision**
- the constraint is **tightly respected** for smaller values of  $\eta$  and that the quantity of the drug injected is greater when  $\eta$  is smaller
- constraint fulfillment is obtained for **any** positive value of  $\eta$
- the strategy of **intensive chemotherapy** seems to prevail



A. S. Matveev and A. V. Savkin *Application of optimal Control Theory to Analysis of Cancer Chemotherapy Regimens*. *Systems & Control Letters*, 46:311-321, 2002.

# Sensitivity to initial guess

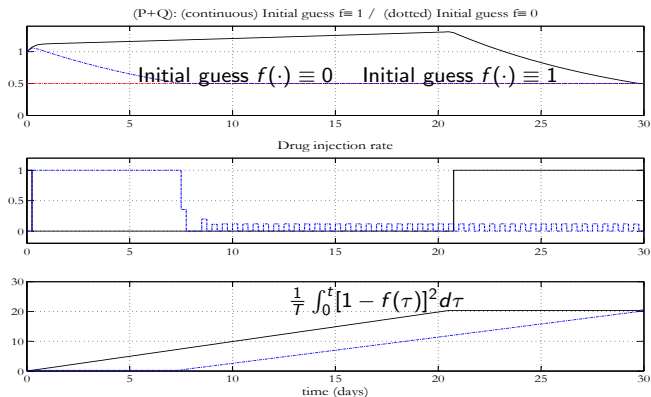


Figure: Comparison between two different solutions of the unconstrained problem  $P_{0.5}^{0.001}(0.6, 0.4)$  resulting from two different initial guess  $f \equiv 0$  (dotted line) and  $f \equiv 1$  (continuous line)

## Part V

# Conclusion



## By considering appropriate state-unconstrained optimal control problems

- the solution of the state-constrained problem occurring in cancer cell-cycle specific chemotherapy can be approximated **up to any desired precision**
- the resulting approximation error can be **explicitly computed** via the appropriate two-sided inequalities

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## Perspectives :

Realistic control schemes have to deal with **parametric uncertainties and/or measurement errors**

=> this work has to be seen as **an element in a more sophisticated control loop (predictive or adaptive)** in which the estimated values of the system's parameters have to be systematically updated based on appropriate assumptions