



# Geometric and numerical methods in optimal control for the time minimal saturation in Magnetic Resonance

Jérémie Rouot, Bernard Bonnard, Olivier Cots, Thibaut Verron

## ► To cite this version:

Jérémie Rouot, Bernard Bonnard, Olivier Cots, Thibaut Verron. Geometric and numerical methods in optimal control for the time minimal saturation in Magnetic Resonance. Dynamics, Control and Geometry In honor of Bronislaw Jakubczyk's 70th birthday, Sep 2018, Varsovie, Poland. hal-02483317

HAL Id: hal-02483317

<https://inria.hal.science/hal-02483317>

Submitted on 18 Feb 2020

**HAL** is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

# **Geometric and numerical methods in optimal control for the time minimal saturation in Magnetic Resonance Imaging**

*DYNAMICS, CONTROL, and GEOMETRY*

In honor of Bronisław Jakubczyk's 70th birthday

12.09.2018 - 15.09.2018 | Banach Center, Warsaw

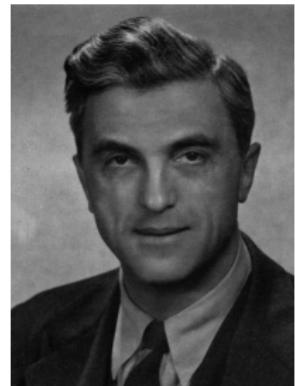
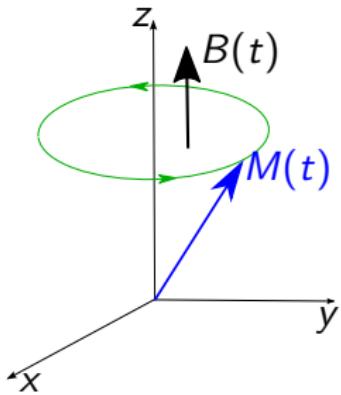
J. Rouot\*, B. Bonnard, O. Cots, T. Verron

\*EPF.Troyes, France, [jeremy.rouot@epf.fr](mailto:jeremy.rouot@epf.fr)

# Magnetization vector

- **Bloch equation:**  $M$ : magnetization vector of the spin-1/2 particle in a magnetic field  $B(t)$ .

$$\dot{M}(t) = -\kappa M(t) \times B(t)$$



F. Bloch Nobel Prize (1952)

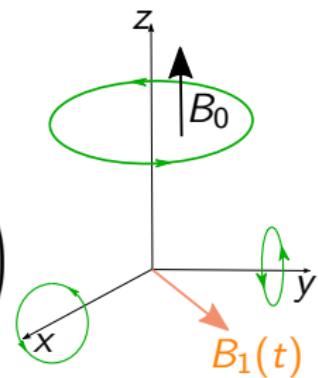
# Experimental model in Nuclear Magnetic Resonance

- **Two magnetic fields** : controlled field  $B_1(t)$  and a strong static field  $B_0$

# Experimental model in Nuclear Magnetic Resonance

- Two magnetic fields : controlled field  $B_1(t)$  and a strong static field  $B_0$

$$\begin{pmatrix} \dot{M_x} \\ \dot{M_y} \\ \dot{M_z} \end{pmatrix} = \begin{pmatrix} -\Gamma M_x \\ -\Gamma M_y \\ -\gamma(M_0 - M_z) \end{pmatrix} + \begin{pmatrix} 0 & -\omega_0 & \omega_y \\ \omega_0 & 0 & -\omega_x \\ -\omega_y & \omega_x & 0 \end{pmatrix} \begin{pmatrix} M_x \\ M_y \\ M_z \end{pmatrix}$$

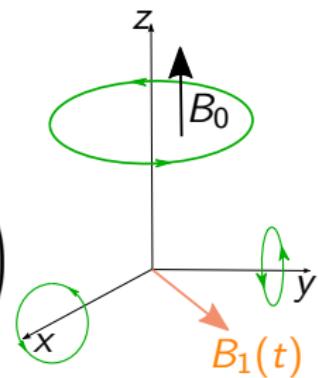


- $\Gamma, \gamma$  are parameters related to the observed species
- $\omega_0$  is fixed and associated to  $B_0$
- $\omega_x, \omega_y$  are related to the controlled magnetic field  $B_1(t)$

# Experimental model in Nuclear Magnetic Resonance

- Two magnetic fields : controlled field  $B_1(t)$  and a strong static field  $B_0$

$$\begin{pmatrix} \dot{M}_x \\ \dot{M}_y \\ \dot{M}_z \end{pmatrix} = \begin{pmatrix} -\Gamma M_x \\ -\Gamma M_y \\ -\gamma(M_0 - M_z) \end{pmatrix} + \begin{pmatrix} 0 & -\omega_0 & \omega_y \\ \omega_0 & 0 & -\omega_x \\ -\omega_y & \omega_x & 0 \end{pmatrix} \begin{pmatrix} M_x \\ M_y \\ M_z \end{pmatrix}$$



- $\Gamma, \gamma$  are parameters related to the observed species
  - $\omega_0$  is fixed and associated to  $B_0$
  - $\omega_x, \omega_y$  are related to the controlled magnetic field  $B_1(t)$
- $M(t) \in S(O, |M(0)|)$ ,  
 $B_1 \equiv 0 \Rightarrow$  relaxation to the stable equilibrium  $M = (0, 0, |M(0)|)$ .

- Normalized Bloch equation in the *rotating frame* ( $\omega_0, (Oz)$ )

$$\dot{x}(t) = -\Gamma x(t) + \textcolor{orange}{u_y(t)} z(t),$$

$$\dot{y}(t) = -\Gamma y(t) - \textcolor{orange}{u_x(t)} z(t),$$

$$\dot{z}(t) = \gamma(1 - z(t)) - \textcolor{orange}{u_y(t)} x(t) + \textcolor{orange}{u_x(t)} y(t).$$

- $q = (x, y, z) = M/M(0)$  is the normalized magnetization vector,
- $(\textcolor{orange}{u_x}, \textcolor{orange}{u_y})$  is the control.

- Normalized Bloch equation in the *rotating frame* ( $\omega_0, (Oz)$ )

$$\dot{x}(t) = -\Gamma x(t) + \textcolor{orange}{u_y(t)} z(t),$$

$$\dot{y}(t) = -\Gamma y(t) - \textcolor{orange}{u_x(t)} z(t),$$

$$\dot{z}(t) = \gamma(1 - z(t)) - \textcolor{orange}{u_y(t)} x(t) + \textcolor{orange}{u_x(t)} y(t).$$

- $q = (x, y, z) = M/M(0)$  is the normalized magnetization vector,
  - $(\textcolor{orange}{u_x}, \textcolor{orange}{u_y})$  is the control.
- Symmetry of revolution around  $(Oz)$ , we set:  $u_y = 0$  and we obtain the **planar control system**

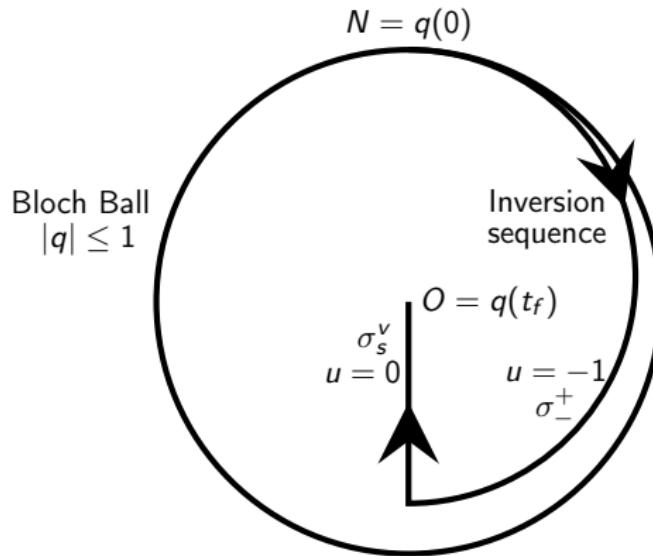
$$\dot{y}(t) = -\Gamma y(t) - \textcolor{orange}{u(t)} z(t),$$

$$\dot{z}(t) = \gamma(1 - z(t)) + \textcolor{orange}{u(t)} y(t),$$

and  $\textcolor{orange}{u} = u_x$  is the control satisfying  $|\textcolor{orange}{u}| \leq 1$ .

## Saturation of a single spin in minimum time

- **Aim.** Steer the North pole  $N = (0, 1)$  of the Bloch ball  $\{|q| \leq 1\}$  to the center  $O$  in minimum time.



The inversion sequence  $\sigma_-^N \sigma_s^v$  is not optimal in many physical cases

- **Pontryagin Maximum Principle.**

- Pseudo-Hamiltonian:  $H(q, p, u) = p \cdot (F(q) + u G(q)) = H_F + u H_G$
- $u(\cdot)$  optimal  $\Rightarrow \exists p(\cdot) \in \mathbb{R}^2 \setminus \{0\}$ :

$$\dot{q} = \frac{\partial H}{\partial p}, \quad \dot{p} = -\frac{\partial H}{\partial q}$$

$$H(q(t), p(t), u(t)) = \max_{|v| \leq 1} H(q(t), p(t), v) = cst \geq 0$$

- Regular and bang-bang controls:  
 $u(t) = \text{sign}(H_G(q(t), p(t))), H_G(q(t), p(t)) \neq 0$
- Singular trajectories are contained in  $\{q, \det(G, [F, G])(q) = 0\}$ :

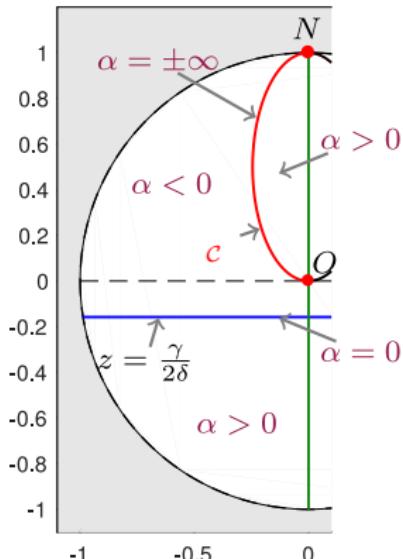
$z = \gamma / (2 \delta) = z_s(\gamma, \Gamma), \delta = \gamma - \Gamma \quad \text{and} \quad y = 0.$

## Computations:

$$D'(q) + u D(q) = 0$$

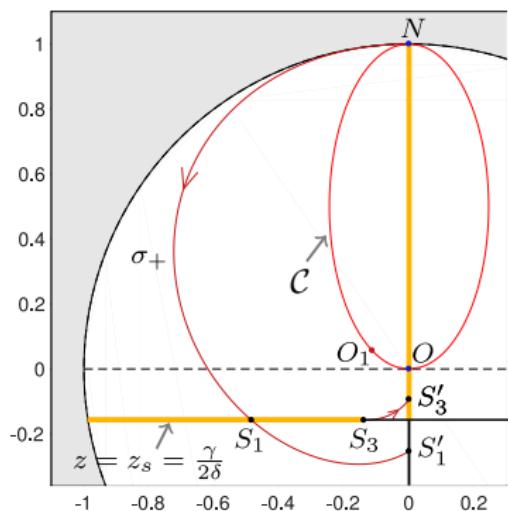
with  $D = \det(G, [G, [F, G]])$  and  $D' = \det(G, [F, [F, G]])$ .  
 We obtain:

- $u_s = \gamma(2\Gamma - \gamma)/(2\delta y)$  on the horizontal singular line.
- $u_s = 0$  on the vertical singular line



- Symmetry:  $u \leftarrow -u$  corresponds to  $y \leftarrow -y$
- Collinearity set:  
 $\mathcal{C} = \{q \mid \det(F, G)(q) = 0\}$
- Switching function:  
 $\Phi(t) = p(t) \cdot G(q(t))$  and outside  
 the set  $\mathcal{C}$ ,  
 $\text{sign}(\dot{\Phi}(t)) = \text{sign}(\alpha(q))$ ,  $\alpha(q) \neq 0$   
 where  $\alpha(q(t)) = \frac{\det(G, [F, G])(q)}{\det(G, F)(q)}$ .

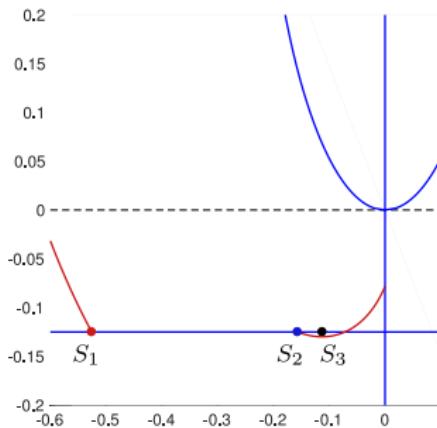
# Definition of the points $S_1, S_3$



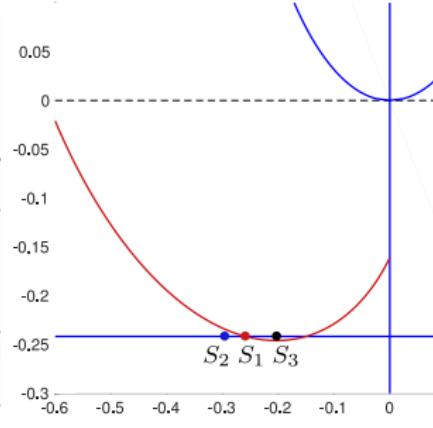
The singular trajectory  $q(\cdot)$  is called

- **Hyperbolic** if  $p(t) \cdot [G, [F, G]](q(t)) = \frac{\partial}{\partial u} \frac{d^2}{dt^2} \frac{\partial H}{\partial u}(q(t), p(t)) > 0$ .
- **Elliptic** if  $p(t) \cdot [G, [F, G]](q(t)) = \frac{\partial}{\partial u} \frac{d^2}{dt^2} \frac{\partial H}{\partial u}(q(t), p(t)) < 0$ .

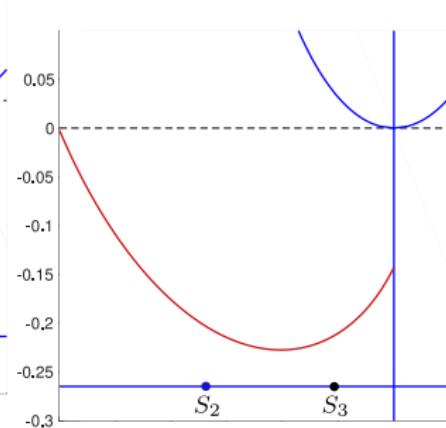
Optimal synthesis depends on the ratio  $\frac{\gamma}{\Gamma}$ .



**Case 1:**  $S_1$  exists and  
 $S_2 \in S_1 S_3$

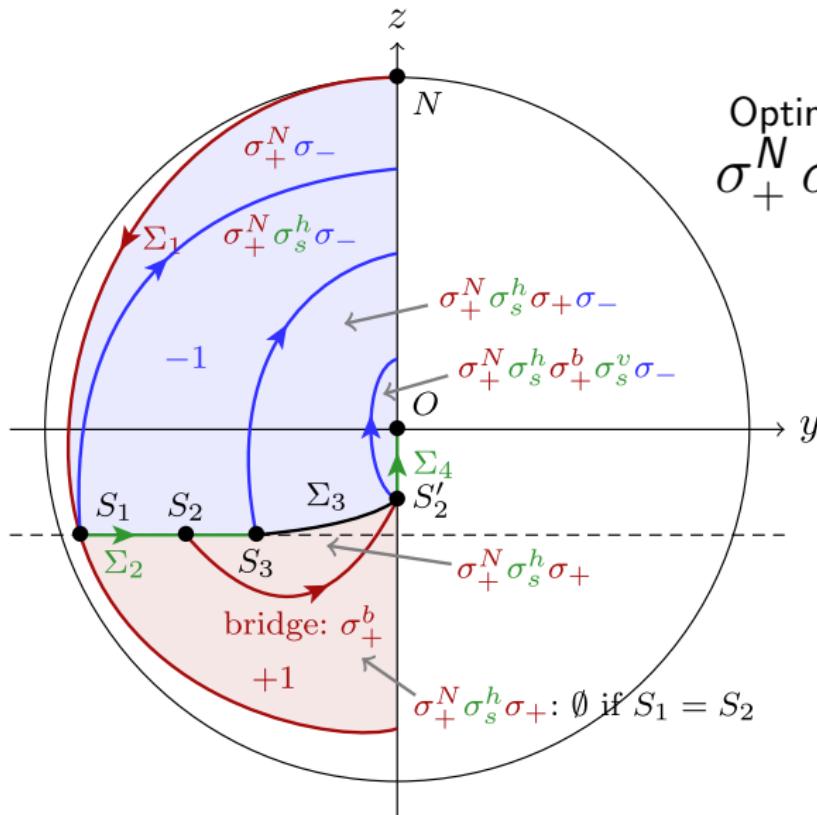


**Case 2:**  $S_1$  exists and  
 $S_2 \notin S_1 S_3$



**Case 3:**  $S_1$  doesn't exist

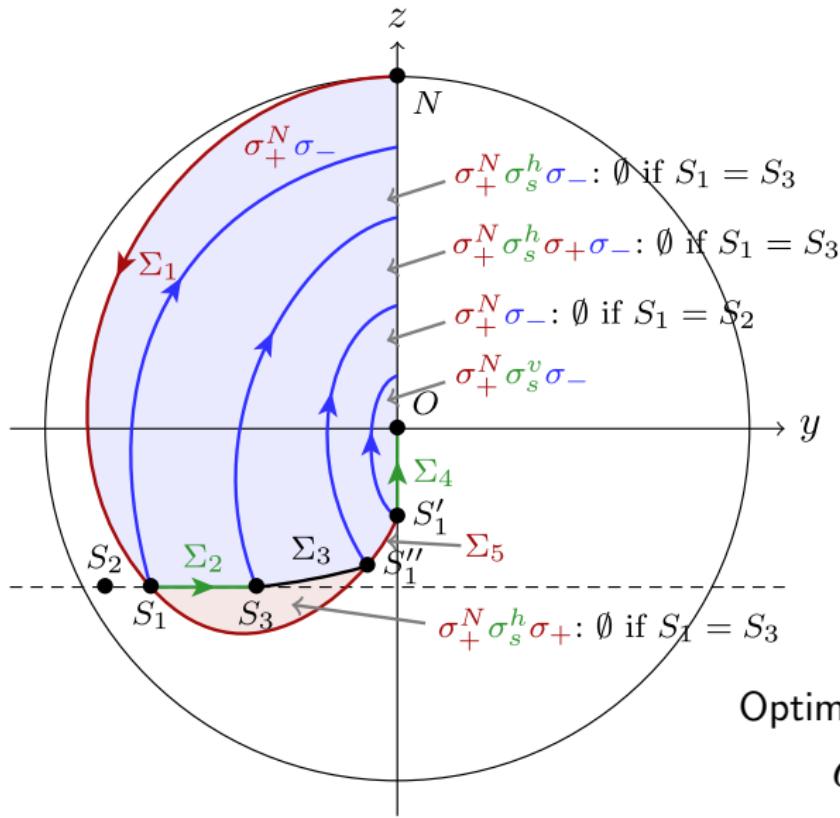
## Case 1: $S_1$ exists and $S_2 \in S_1 S_3$



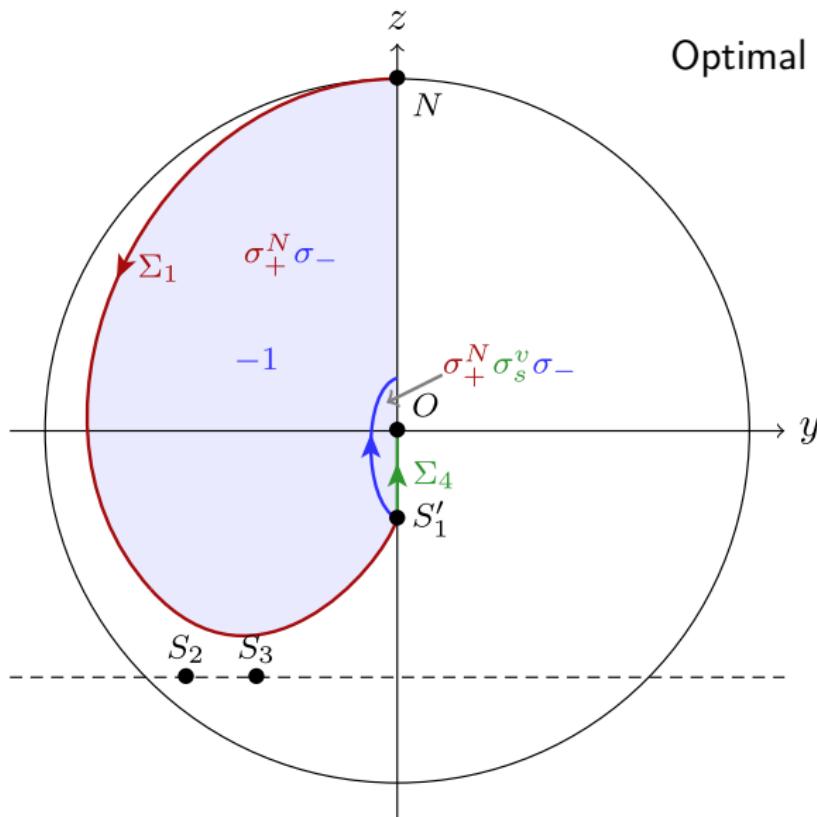
Optimal trajectory from  $N$  to  $O$ :

$$\sigma_+^N \sigma_s^h \sigma_+^b \sigma_s^v \sigma_-$$

## Case 2: $S_1$ exists and $S_2 \notin S_1 S_3$



## Case 3: $S_1$ doesn't exists



Optimal trajectory from  $N$  to  $O$ :

$$\sigma_+^N \sigma_s^v$$

## Theorem

*The time optimal trajectory for the saturation problem of 1-spin is of the form:*

$$\sigma_+^N \quad \underbrace{\sigma_s^h \quad \sigma_+^b}_{\text{empty if } S_2 \leq S_1} \quad \sigma_s^v$$

# Numerical validations using Moments/LMI techniques

*Aim:* Provide lower bounds on the global optimal time.

- Numerical times obtain with the HamPath software to validate :

Case	$\Gamma$	$\gamma$	$t_f$
$C_1$	$9.855 \times 10^{-2}$	$3.65 \times 10^{-3}$	42.685
$C_2$	$2.464 \times 10^{-2}$	$3.65 \times 10^{-3}$	110.44
$C_3$	$1.642 \times 10^{-2}$	$2.464 \times 10^{-3}$	164.46
$C_4$	$9.855 \times 10^{-2}$	$9.855 \times 10^{-2}$	8.7445

# Context

$$t_f = \inf_{\underline{u}(\cdot)} T$$

$$\dot{x}(t) = f(x(t), u(t)),$$

$$x(t) \in X, \quad u(t) \in U, \quad x(0) \in X_0, \quad x(T) \in X_T$$

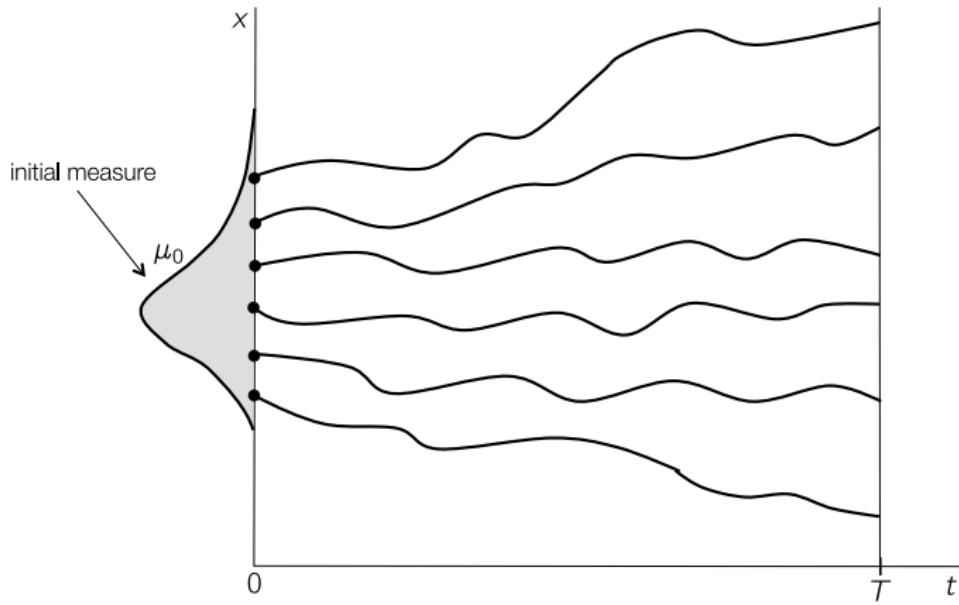
$X, U, X_0, X_T$  are subsets of  $\mathbb{R}^n$  which can be written as

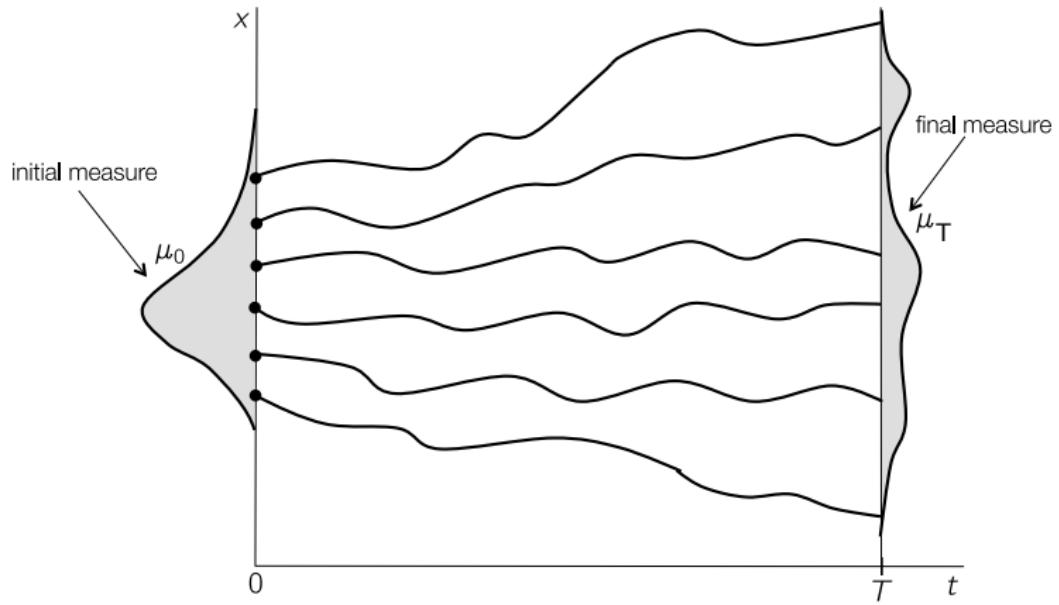
$$X = \{(t, x) : p_k(t, x) \geq 0, k = 0, \dots, n_x\}, \quad U = \{u : q_k(u) \geq 0, k = 0, \dots, n_u\}$$

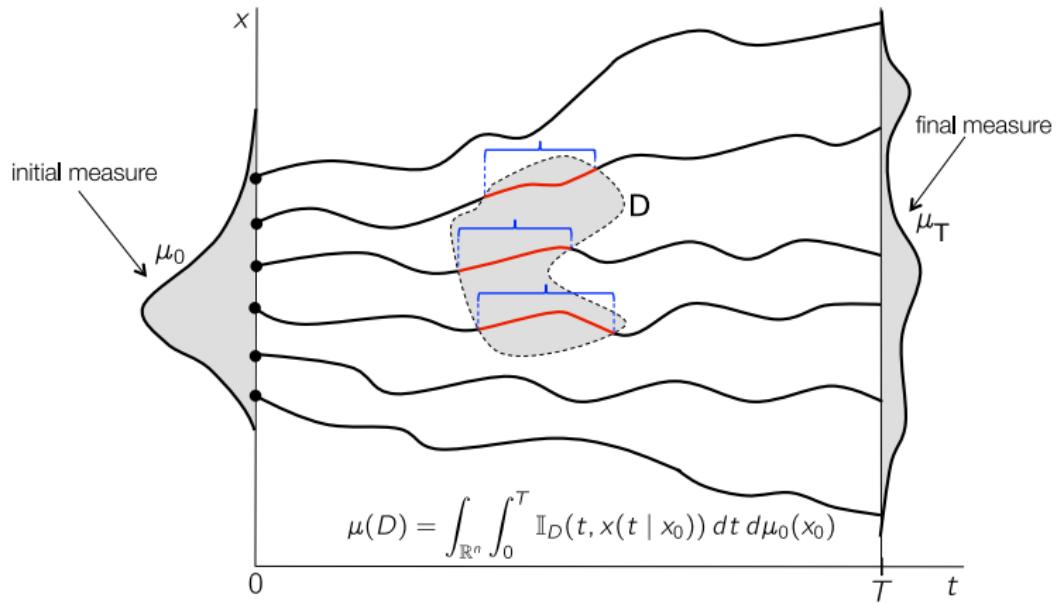
$$X_0 = \{x : r_k^0(x) \geq 0, k = 0, \dots, n_0\}, \quad X_T = \{(t, x) : r_k^T(t, x) \geq 0, k = 0, \dots, n_T\}$$

**Objective:** Compute  $\min_{u(\cdot)} T$  when  $f, p_k, q_k, r_k^0, r_k^T$  are polynomials and the above sets are compacts.

**Result:** [J. B. Lasserre, D. Henrion, C. Prieur, E. Trélat, 2008]  
Converging monotone nondecreasing sequence of lower bounds of  $t_f$ .







$$\int_0^T v(t, x(t)) dt = \int_0^T \int_X \int_U v(t, x) d\mu(t, x, u), \quad v \in \mathcal{C}^0([0, T] \times X)$$

## Liouville's equation

Linear equation linking the measures  $\mu_0, \mu$  and  $\mu_T$ .

$$\int_{X_T} v(T, x) d\mu_T(x) - \int_{X_0} v(0, x) d\mu_0(x) = \int_{[0, T] \times Q \times U} \frac{\partial v}{\partial t} + \nabla_x \cdot \mathbf{f}(x, u) d\mu(t, x, u)$$

for all test functions  $v \in \mathcal{C}^1([0, T] \times X)$ .

Optimization over system trajectories

$\Leftrightarrow$

Optimization over measures satisfying Liouville equation.

- **Relaxed controls:**  $u(t)$  is replaced for each  $t$  by a probability measure  $\omega_t(u)$  supported on  $U$ .
- **Relaxed problem:**

$$T_R = \min_{\omega} T$$

$$\text{s.t. } \dot{x}(t) = \int_U f(x(t), u) d\omega_t(u)$$

$$x(0) \in X_0, \quad x(t) \in X, \quad x(T) \in X_T$$

- **Relaxed controls:**  $u(t)$  is replaced for each  $t$  by a probability measure  $\omega_t(u)$  supported on  $U$ .
- **Relaxed problem:**

$$T_R = \min_{\omega} T$$

$$\text{s.t. } \dot{x}(t) = \int_U f(x(t), u) d\omega_t(u)$$

$$x(0) \in X_0, \quad x(t) \in X, \quad x(T) \in X_T$$

- **Linear Problem on measures:**

$$d\mu(t, x, u) = dt d\delta_{x(t)}(x) d\omega_t(u) \in \mathcal{M}_+([0, T] \times X \times U)$$

$$T_{LP} = \min_{\mu, \mu_T, \mu_0} \int d\mu_T$$

$$\begin{aligned} \text{s.t. } & \int \left( \frac{\partial v}{\partial t} + \frac{\partial v}{\partial x} f(x, u) \right) d\mu \\ &= \int v(\cdot, x_T) d\mu_T - \int v(0, x_0) d\mu_0, \quad \forall v \in \mathbb{R}[t, x], \end{aligned}$$

$$\mu \in \mathcal{M}_+([0, T] \times X \times U), \quad \mu_T \in \mathcal{M}_+(X_T), \quad \mu_0 \in \mathcal{M}_+(X_0)$$

**Notation:**  $\alpha = (\alpha_1, \dots, \alpha_p) \in \mathbb{N}^p$ ,  $z = (z_1, \dots, z_p) \in \mathbb{R}^p$ . We denote by  $z^\alpha$  the monomial  $z_1^{\alpha_1} \dots z_p^{\alpha_p}$  and by  $\mathbb{N}_d^p$  the set  $\{\alpha \in \mathbb{N}^p, |\alpha|_1 = \sum_{i=1}^p \alpha_i \leq d\}$ .

**Moment of order  $\alpha$  for a measure  $\nu \in \mathcal{M}_+(Z)$ :**  $y_\alpha^\nu = \int z^\alpha d\nu(z)$ .

**Riesz linear functional:**  $l_{y^\nu} : \mathbb{R}[z] \rightarrow \mathbb{R}$  s.t.  $l_{y^\nu}(z^\alpha) = y_\alpha^\nu$ .

**Moment Matrix:**  $M_d(y^\nu)[i,j] = y_{i+j}^\nu, \forall i,j \in \mathbb{N}_d^p$ .

**Notation:**  $\alpha = (\alpha_1, \dots, \alpha_p) \in \mathbb{N}^p$ ,  $z = (z_1, \dots, z_p) \in \mathbb{R}^p$ . We denote by  $z^\alpha$  the monomial  $z_1^{\alpha_1} \dots z_p^{\alpha_p}$  and by  $\mathbb{N}_d^p$  the set  $\{\alpha \in \mathbb{N}^p, |\alpha|_1 = \sum_{i=1}^p \alpha_i \leq d\}$ .

**Moment of order  $\alpha$  for a measure  $\nu \in \mathcal{M}_+(Z)$ :**  $y_\alpha^\nu = \int z^\alpha d\nu(z)$ .

**Riesz linear functional:**  $l_{y^\nu} : \mathbb{R}[z] \rightarrow \mathbb{R}$  s.t.  $l_{y^\nu}(z^\alpha) = y_\alpha^\nu$ .

**Moment Matrix:**  $M_d(y^\nu)[i, j] = y_{i+j}^\nu, \forall i, j \in \mathbb{N}_d^p$ .

**Proposition (Putinar, 1993)**

Let  $Z = \{z \in \mathbb{R}^p \mid g_k(z) \geq 0, k = 1, \dots, n_Z\}$ . The sequence  $(y_\alpha)_\alpha$  has a representing measure  $\nu \in \mathcal{M}_+(Z)$  if and only if

$$M_d(y) \succeq 0, \quad M_d(g_k y) \succeq 0, \quad \forall d \in \mathbb{N}, \forall k = 1, \dots, n_Z.$$

## Moment Semidefinite Programming Problem:

$$T_{SDP} = \min_{y^\mu, y^{\mu_T}} I_{y^{\mu_T}}(1)$$

$$I_{y^\mu} \left( \frac{\partial v}{\partial t} + \frac{\partial v}{\partial x} f(x, u) \right) = I_{y^{\mu_T}}(v(\cdot, x_T)) - I_{y^{\mu_0}}(v(0, x_0)), \forall v \in \mathbb{R}[t, x],$$

$$M_d(y^\mu) \succeq 0, \quad M_d(g_i y^\mu) \succeq 0, \quad \forall i, \quad \forall d \in \mathbb{N},$$

$$M_d(y^{\mu_0}) \succeq 0, \quad M_d(g_i^0 y^{\mu_T}) \succeq 0, \quad \forall i \quad \forall d \in \mathbb{N}$$

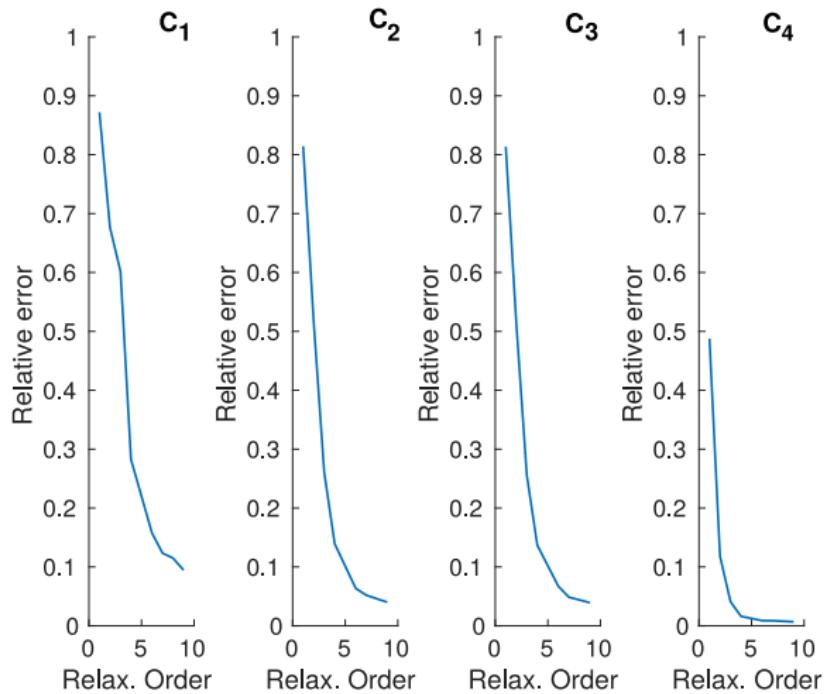
$$M_d(y^{\mu_T}) \succeq 0, \quad M_d(g_i^T y^{\mu_T}) \succeq 0, \quad \forall i \quad \forall d \in \mathbb{N}$$

where  $g_i, g_i^0$  and  $g_i^T$  are polynomials defining the sets  $[0, T] \times X \times U$ ,  $X_0$  and  $X_T$  respectively.

By truncating the sequences  $(y^\mu)$ ,  $(y^{\mu_T})$  up to moments of length  $r$  (relaxation order), we have a **hierarchy of Semidefinite Programming Problems** and the lower bounds  $T_{sdp}^1, \dots, T_{sdp}^r, \dots$  of these problems satisfy:

$$\textcolor{red}{t_f} = T_{LP} = T_{SDP} \geq \dots \geq T_{sdp}^{r+1} \geq T_{sdp}^r \geq \dots \geq T_{sdp}^1.$$

# Numerical results on the saturation problem



## Perspectives

- Generalization to an ensemble of pair of spins where Bloch equations are coupled and Inhomogeneities on the control field are taken into account.
- Contrast problem where we have two species to discriminate. Saturation of the first spin while the norm of the second spin is maximized.