

An introduction to optimal control problem

The use of Pontryagin maximum principle

Jérôme Lohéac

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ERC NUMERIWAVES – Course

Generalities

The aim of this course is to give basis to solve analytically or numerically optimal control problems.

In full generality, we consider a system governed by the dynamic:

$$\dot{x} = f(x, u). \quad (1)$$

with $x \in \mathbb{R}^n$ is the state variable and $u \in \mathbb{R}^m$ is the control variable. The control problems is:

Problem (Control problem)

Given $T > 0$, $x^i, x^f \in \mathbb{R}^n$ does it exists $u : [0, T] \rightarrow \mathbb{R}^m$ such that systems (1) steers x^i to x^f in time T .

For optimal control problem, we consider a cost function:

$$J(x, u) = \int_0^T f_0(x(t), u(t)) dt + g(x(T), u(T))$$

and the aim is to find a control u which realize the control problem and minimize J .

References (books)

- E. Trélat [Contrôle optimal : théorie et applications](#)
- J.-M. Coron [Control and Nonlinearity](#)
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- A. A. Agrachev and Y. L. Sachkov [Control theory from the geometric viewpoint](#)
- L. S. Pontryagin [The maximum principle in optimal control](#)
- A. D. Ioffe and V. M. Tihomirov [Theory of extremal problems](#)
- J.-L. Lions [Contrôlabilité exacte, perturbations et stabilisation de systèmes distribués](#)
- M. Tucsnak and G. Weiss [Observation and control for operator semi-groups](#)
- H. O. Fattorini [Infinite-dimensional optimization and control theory](#)

- 1 Controllability results
- 2 Optimal control problems
- 3 Numerical methods

- 1 Controllability results
 - Linear control problems
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- 2 Optimal control problems
- 3 Numerical methods

Linear control problems I

Autonomous systems

We consider the system:

$$\dot{x} = Ax + Bu \quad x(0) = x^i. \quad (2)$$

with $A \in \mathcal{M}_n(\mathbb{R})$ and $B \in \mathcal{M}_{n,m}(\mathbb{R})$.

Proving the controllability of the system (2) is equivalent as proving the observability of the adjoint system:

$$\dot{\varphi} = -A^* \varphi \quad \varphi(T) = \varphi^f, \quad (3)$$

with the observation operator B^* .

That is to say, proving that there exists a constant $C > 0$ such that:

$$\|\varphi^f\|_{\mathbb{R}^n}^2 \leq C \int_0^T \|B^* \varphi(t)\|_{\mathbb{R}^m}^2 dt \quad (\varphi^f \in \mathbb{R}^n).$$

Linear control problems II

Autonomous systems

Theorem (Kalman rank condition)

If

$$\text{rank} [B \ AB \ \dots \ A^{n-1}B] = n,$$

then for every $T > 0$ and every $x^i, x^f \in \mathbb{R}^n$, there exists $u \in L^2(0, T; \mathbb{R}^m)$ such that system (2) steers x^i to x^f in time T . In addition, the control of minimal L^2 -norm is given by:

$$u(t) = B^* \varphi(t)$$

with φ the solution of the adjoint problem (3) with final condition $\hat{\varphi}^f$ and where $\hat{\varphi}^f$ realise the minimum of:

$$J(\varphi^f) = \frac{1}{2} \int_0^T \|B^* \varphi(t)\|_{\mathbb{R}^m}^2 dt + \langle x^i, \varphi(0) \rangle_{\mathbb{R}^n} - \langle x^f, \varphi^f \rangle_{\mathbb{R}^n}$$

(where here φ is the solution of the adjoint problem (3) with final condition φ^f).

Linear control problems III

Autonomous systems

Example

$$\ddot{x} = -x + u$$

Linear control problems

Non-autonomous systems

We consider here the system:

$$\dot{x} = A(t)x + B(t)u \quad (4)$$

and we assume that A and B are smooth.

Theorem (Kalman rank condition for non-autonomous linear systems)

Let us define for every $i \in \mathbb{N}$ the matrices $B_i \in \mathcal{M}_{n,m}(\mathbb{R})$ by:

$$B_0(t) = B(t), \quad B_i(t) = \dot{B}_{i-1}(t) - A(t)B_{i-1}(t) \quad (t \in [0, T]).$$

If there exists $\bar{t} \in [0, T]$ so that:

$$\text{Span} \{B_i(\bar{t})u, u \in \mathbb{R}^m, i \in \mathbb{N}\} = \mathbb{R}^n$$

then the system (4) is controllable.

Nonlinear control problems I

Linear test

We consider here the general system

$$\dot{x} = f(x, u). \quad (5)$$

Let assume that $f \in C^1(\mathbb{R}^n \times \mathbb{R}^m, \mathbb{R}^n)$. Let us consider a trajectory (\bar{x}, \bar{u}) , i.e. $\dot{\bar{x}} = f(\bar{x}, \bar{u})$, the linearizing along this trajectory leads to the linear system:

$$\dot{x} = \partial_x f(\bar{x}, \bar{u}) x + \partial_u f(\bar{x}, \bar{u}) u. \quad (5')$$

Nonlinear control problems II

Linear test

Theorem

If system (5') is controllable, then the system (5) is locally controllable along the trajectory (\bar{x}, \bar{u}) .

That is to say:

For every $\varepsilon > 0$, there exists $\eta > 0$ such that for every $x^i, x^f \in \mathbb{R}^n$ with

$$\|\bar{x}(0) - x^i\|_{\mathbb{R}^n}, \|\bar{x}(T) - x^f\|_{\mathbb{R}^n} < \eta,$$

there exists a trajectory (x, u) of (5) so that

$$x(0) = x^i, \quad x(T) = x^f \quad \text{and} \quad \sup_{t \in [0, T]} \|u(t) - \bar{u}(t)\|_{\mathbb{R}^m} \leq \varepsilon.$$

Nonlinear control problems III

Linear test

Example (Spring)

$$\ddot{x} = -x - x^3 + u$$

Example (Brockett integrator)

$$\dot{x}_1 = u_1$$

$$\dot{x}_2 = u_2$$

$$\dot{x}_3 = x_1 u_2 - x_2 u_1$$

Nonlinear control problems I

Systems without drift

Let us now consider the control system:

$$\dot{x} = \sum_{i=1}^m f_i(x) u_i, \quad (6)$$

where the f_i are smooth vector fields on \mathbb{R}^n .

To these vector fields, we associated the Lie algebra $\text{Lie}\{f_1, \dots, f_m\}$ generated by the iterated Lie brackets:

$$[f_i, f_j](x) = \partial_x f_j f_i(x) - \partial_x f_i f_j(x) \quad (x \in \mathbb{R}).$$

Nonlinear control problems II

Systems without drift

Theorem (Chow)

Let us assume:

$$\text{rank Lie}\{f_1, \dots, f_m\}(x) = n \quad (x \in \mathbb{R}^n).$$

Then,

- 1 If the set of admissible controls $\Omega \subset \mathbb{R}^m$ is nonempty and contains 0 in its interior, then there exists $T > 0$ such that the system (6) is controllable.
- 2 If the admissible set Ω is \mathbb{R}^m , for every $T > 0$, the system (6) is controllable.

Example (Brockett integrator)

$$\begin{aligned}\dot{x}_1 &= u_1 \\ \dot{x}_2 &= u_2 \\ \dot{x}_3 &= x_1 u_2 - x_2 u_1\end{aligned}$$

Nonlinear control problems I

Drift systems

Let us now consider the control system with drift:

$$\dot{x} = f_0(x) + \sum_{i=1}^m f_i(x)u_i, \quad (7)$$

where the f_i are smooth vector fields on \mathbb{R}^n and we assume that $f(0) = 0$ (i.e. $(0,0)$ is an equilibrium point).

For such a system, it is also natural to consider the Lie algebra generated by $\{f_0, \dots, f_m\}$.

But the Lie algebra rank condition is not enough to obtain controllability,

Example

$$\begin{aligned}\dot{x}_1 &= x_2^2 \\ \dot{x}_2 &= u\end{aligned}$$

Nonlinear control problems II

Drift systems

To end up with this problem, we define the set $\text{Br} \{f_0, \dots, f_m\}$ of formal iterated Lie brackets generated by $\{f_0, \dots, f_m\}$. And for every $h \in \text{Br}$, we define $\delta_i(h)$ the number of times that f_i appears in h .

Let us also define

$$\sigma(h) = \sum_{\pi \in \text{Per}_m} \pi(h),$$

with Per_m the group of permutation of $\{0, \dots, m\}$ such that for every $\pi \in \text{Per}_m$, $\pi(0) = 0$.

For $\theta \in [0, \infty]$, we will say that the control system (7) satisfies the **Sussmann condition** $S(\theta)$ if

- ① it satisfies the Lie algebra rank condition at 0,
- ② for every $h \in \text{Br}$ with $\delta_0(h)$ odd and $\delta_i(h)$ even for every $i \in \{1, \dots, m\}$, $\sigma(h)(0)$ is in the span of the $g(0)$'s such that

$$g \in \text{Br} \quad \text{and} \quad \begin{cases} \theta \delta_0(g) + \sum_{i=1}^m \delta_i(g) < \theta \delta_0(h) + \sum_{i=1}^m \delta_i(h) & \text{if } \theta \in [0, \infty), \\ \delta_0(g) < \delta_0(h) & \text{if } \theta = \infty. \end{cases}$$

Nonlinear control problems III

Drift systems

Theorem (Sussmann)

If for some $\theta \in [0, 1]$, the control system (7) satisfies the condition $S(\theta)$, then it is small time locally controllable.

That is to say:

For every $\tau > 0$, there exists $\eta > 0$ such that for every $x^i, x^f \in \mathbb{R}^n$, with

$$\|x^i\|, \|x^f\| < \eta,$$

there exists $u : [0, \tau] \rightarrow \mathbb{R}^m$ such that the trajectory of (7) satisfies:

$$x(0) = x^i, \quad x(\tau) = x^f \quad \text{and} \quad \|u(t)\| \leq \tau \quad (t \in [0, \tau]).$$

Example

$$\dot{x}_1 = x_2^3, \quad \dot{x}_2 = u$$

Nonlinear control problems IV

Drift systems

We can extend this result to more general problems. Consider the general control system:

$$\dot{x} = f(x, u), \quad (8)$$

with $f(0, 0) = 0$.

To (8), we associate:

$$\dot{x} = f(x, y), \quad \dot{y} = u. \quad (8')$$

Theorem

If for some $\theta \in [0, 1]$, the control system (8) (i.e. (8')) satisfies the condition $S(\theta)$, then it small time locally controllable.

- 1 Controllability results
- 2 Optimal control problems
 - Linear control problems
 - General control problems
- 3 Numerical methods

Linear control problems I

Linear-quadratic theory

We consider the system:

$$\dot{x} = A(t)x + B(t)u, \quad x(0) = x^i \quad (9)$$

and given $T > 0$, we want to minimize the cost:

$$J(u) = x(T)^\top Qx(T) + \int_0^T (x(t)^\top W(t)x(t) + u(t)^\top U(t)u(t)) dt, \quad (10)$$

with x the solution of (9)

In the above, the matrices Q and $W(t)$ are symmetric nonnegative and $U(t)$ is positive. In addition, we assume that A, B, W and U are L^∞ in time.

Linear control problems II

Linear-quadratic theory

Theorem (Existence)

Assume there exists $\alpha > 0$ such that:

$$\int_0^T u^\top(t) U(t) u(t) dt \geq \alpha \int_0^T u^\top(t) u(t) dt \quad (u \in L^2([0, T], \mathbb{R}^m)),$$

then there exists a unique control $u \in L^2([0, T]; \mathbb{R}^n)$ minimizing J .

Linear control problems III

Linear-quadratic theory

Theorem (Optimality condition)

The trajectory x associated to the optimal control u is optimal if and only if there exists an adjoint vector p such that:

$$\dot{p} = -A(t)^\top p + W(t)x(t), \quad p(T) = -Qx(T).$$

Moreover, the optimal control u is:

$$u(t) = U(t)^{-1}B(t)^\top p(t).$$

Example

$$\dot{x} = u, \quad x(0) = x^i$$

with the cost $J(u) = \int_0^T (x(t)^2 + u(t)^2) dt$.

Linear control problems I

Time optimality

We consider the system:

$$\dot{x} = A(t)x + B(t)u, \quad x(0) = x^i \quad (11)$$

Given $x^f \in \mathbb{R}^n$, the aim is to find the minimal time $T > 0$ such that there exists a control $u : [0, T] \rightarrow \mathbb{R}^m$ steering x^i to x^f in time T and such that:

$$u(t) \in \Omega \quad (t \in [0, T]),$$

with Ω a compact set of \mathbb{R}^m .

Theorem (Existence)

If there exists a time $T > 0$ such that x^i can be steered to x^f with a control u with $u(t) \in \Omega$ then, the minimal time exists.

Linear control problems II

Time optimality

Theorem (Optimality condition)

Assume that the previous existence theorem hold and let us define T the minimal time. Then the control $u : [0, T] \rightarrow \mathbb{R}^m$ is a control in time T if and only if there exists $p(t)$ a non trivial solution of:

$$\dot{p} = -A(t)^\top p,$$

such that u satisfies:

$$p(t)^\top B(t)u(t) = \max_{v \in \Omega} p(t)^\top B(t)v \quad (t \in [0, T] \text{ a.e.}).$$

Linear control problems III

Time optimality

Let us focus on the autonomous control system:

$$\dot{x} = Ax + Bu, \quad (12)$$

with $A \in \mathbb{R}^{n \times n}$ and $B \in \mathbb{R}^{n \times 1}$.

Theorem (Bang-Bang property)

Assume that the pair (A, B) satisfies the Kalman rank condition and that $\Omega = [-1, 1]$ then,

- 1 If all the eigenvalues of A are real, then the time optimal control has at most $n - 1$ commutations;
- 2 If A has a non real eigenvalue, then for every $N \in \mathbb{N}$, there exists points x^i and x^f such that the time optimal control has at least N commutations.

Example

$$\ddot{x} = -x + u$$

Existence results I

General systems

Let us consider the control system:

$$\dot{x} = f(t, x, u), \quad (13)$$

with a control u such that $u(t) \in \Omega$, with Ω a compact set of \mathbb{R}^m . The aim is to join a set $M^i \subset \mathbb{R}^n$ to a set $M^f \subset \mathbb{R}^n$ by minimizing the cost: with the cost:

$$J(u) = \int_0^{t(u)} f_0(t, x(t), u(t)) dt + g(t(u), x(t(u))).$$

Existence results II

General systems

Theorem

- Assume that M^i is accessible from M^f , i.e. for every $x^f \in M^f$, there exists $x^i \in M^i$, $u : \mathbb{R}_+ \rightarrow \mathbb{R}^m$ with $u(t) \in \Omega$ such that there exists $t = t(u) > 0$ for which the solution of (13) with initial condition $x(0) = x^i$ satisfies $x(t(u)) = x^f$.
- Assume in addition that for all of these control u , there exists a constant $b > 0$ (independent of u) such that:

$$t(u) + \|x(t)\| \leq b \quad (t \in [0, t(u)]).$$

- Assume also that for every $s \in \mathbb{R}$ and $y \in \mathbb{R}^n$, the set

$$\left\{ \begin{pmatrix} f(s, y, v) \\ f_0(s, x, v) + \gamma \end{pmatrix}, v \in \Omega, \gamma \geq 0 \right\} \quad \text{is convex.}$$

Then there exists a control u , an initial condition $x^i \in M^i$ and a time $t(u) \geq (u)$ such that u steers x^i to x^f in time $t(u)$ and u minimize J .

Existence results III

General systems

Remark

The boundedness property for u (u with values in a compact set Ω) and for x can be replaced by some growth conditions.

See for instance Harlt, Sethi and Vickson, [A survey of the maximum principle for optimal control problems with state constraints](#), SIAM review, 1995.

Pontryagin maximum principle I

General result

Theorem (Pontryagin maximum principle)

If u is optimal, then there exists an application $p : [0, T] \rightarrow \mathbb{R}^n$ and $p_0 \leq 0$ such that:

- ① $(p(\cdot), p_0)$ is nontrivial;
- ② for almost every $t \in [0, T]$, we have:

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

$$\dot{p} = -\partial_x H(t, x, p, p_0, u)$$

with:

$$H(t, x, p, p_0, u) = \langle f(t, x, u), p \rangle_{\mathbb{R}^n} + p_0 f_0(t, x, u)$$

and the control u is such that:

$$\max_{v \in \Omega} H(t, x(t), p(t), p_0, v) = H(t, x(t), p(t), p_0, u(t)) \quad (t \in [0, T] \text{ a.e.}).$$

Pontryagin maximum principle II

General result

Remark

One can also notice that:

$$\frac{d}{dt} H(t, x(t), p(t), p_0, u(t)) = \partial_t H(t, x(t), p(t), p_0, u(t)).$$

and hence if f and f_0 are independent of t , then $H(t, x(t), p(t), p_0, u(t))$ is constant.

Pontryagin maximum principle I

Transversality conditions

Theorem (Transversality conditions)

In addition,

- 1 If the final time T is not fixed, then,

$$\max_{v \in \Omega} H(T, x(T), p(T), p_0, v) = -p_0 \partial_t g(T, x(T)).$$

- 2 If $x(0)$ and $x(T)$ are not fixed but satisfies $(x(0), x(T)) \in \mathcal{M}$ with \mathcal{M} a sub-manifold of $\mathbb{R}^n \times \mathbb{R}^n$, then,

$$(p(0), p(T) - p_0 \partial_x g(T, x(T))) \perp T_{(x(0), x(T))} \mathcal{M}.$$

Pontryagin maximum principle II

Transversality conditions

Example

- 1 If $x(0)$ is free, then, $p(0) = 0$;
- 2 If we want periodic trajectories, i.e. $x(0) = x(T)$ and if $g = 0$, then, $p(0) = p(T)$.

Pontryagin maximum principle I

Applications

Example (Linear control problem)

$$\dot{x} = Ax + Bu$$

- T fixed, $x(0) = x^i$, $x(T) = x^f$, $\Omega = \mathbb{R}^m$, $g = 0$, $f_0(t, x, u) = \frac{1}{2} \|u\|^2$;
- T fixed, $x(0) = x^i$, $x(T) \in \mathbb{R}^n$, $\Omega = \mathbb{R}^m$, $g(t, x) = x^\top Qx$,
 $f_0(t, x, u) = x^\top Wx + u^\top Uu$;
- T free with $T > 0$, $x(0) = x^i$, $x(T) = x^f$, $\Omega = B_{\mathbb{R}^m}(0, 1)$, $g = 0$,
 $f_0(t, x, u) = 1$.

Pontryagin maximum principle II

Applications

Example (Brockett integrator)

$$\begin{aligned}\dot{x} &= u \\ \dot{y} &= \langle Mx, u \rangle\end{aligned}\quad (x(0), y(0)) = (0, 0)$$

Minimize T such that $(x(T), y(T)) = (0, y^f)$, with $\Omega = B_{\mathbb{R}^m}(0, 1)$.
Assuming $M^\top \neq M$.

Example (Zermelo's problem)

$$\begin{aligned}\dot{x} &= v \cos(u) + c(y) \\ \dot{y} &= v \sin(u)\end{aligned}\quad (x(0), y(0)) = (0, 0).$$

- $T > 0$ free, $y(T) = 1$, $x(T) \in \mathbb{R}$, $f = 0$, $g(t, x) = x^2$.
Assuming $c(y) > v$ for every y ;
- Minimize T such that $y(T) = 1$ and $x(T) \in \mathbb{R}$.

- 1 Controllability results
- 2 Optimal control problems
- 3 Numerical methods**
 - Indirect methods
 - Direct methods

Shooting methods I

We remind that the optimality system is:

$$\dot{x} = \partial_p H(t, x, p, p_0, u) \quad (14a)$$

$$\dot{p} = -\partial_x H(t, x, p, p_0, u) \quad (14b)$$

where:

$$H(t, x(t), p(t), p_0, u(t)) = \max_{v \in \Omega} H(t, x(t), p(t), p_0, v) \quad (15)$$

In general, p_0 can be chosen to -1 if it is not possible, this means that the admissible trajectories are independent of the cost function.

Let us assume that $u(t)$ is uniquely determined by (15), i.e., $u = u(t, x, p)$. Thus, writing $z = (x, p)^\top$, (14) can be written as:

$$\dot{z} = F(t, z).$$

Shooting methods II

We chose an initial condition $z(0) = z^i$ and we look for z^i such that transversality conditions are fulfilled. The transversality condition can be expressed as the zeros of a function G , defined by $G(z^i) = R(z^i, z(T; z^i))$.

The aim is to find z^i such that:

$$G(z^i) = 0.$$

this can be done for instance with a Newton method.

Remark

If T is free, we can add T in the state variable, with the equation $\dot{T} = 0$ and rescale the problem on $[0, 1]$.

Remark

This procedure can be parallelized. Instead of having an unknown z^i we can take a subdivision $0 = t_0 < \dots < t_N = T$ and take as unknown, $z(t_i)$. In addition to the transversality condition, we will have the continuity of z at points t_1, \dots, t_{N-1} .

Shooting methods

Newton method

We recall the Newton method.

Given $G : \mathbb{R}^d \rightarrow \mathbb{R}^d$ the aim is to find $z \in \mathbb{R}^d$ such that $G(z) = 0$. If z_k is close to z , then we have:

$$0 = G(z) = G(z_k) + dG(z_k) \cdot (z - z_k) + o_0(z - z_k).$$

Then we chose

$$z_{k+1} = z_k + d_k,$$

with $d_k \in \mathbb{R}^d$ solution of:

$$G(z_k) + dG(z_k) \cdot d_k = 0.$$

Full discretization I

For this method, we tackle directly the optimization problem:

$$\begin{array}{l} \min \quad J(T, x, u) \\ \left| \begin{array}{l} \dot{x} = f(t_i, x, u) \\ u(t) \in \Omega \\ x(0) = ?, \quad x(T) = ? \end{array} \right. \end{array} \quad (16)$$

More precisely, we will discretize it, i.e., given a subdivision

$0 = t_0 < \dots < t_N = T$, we consider the unknowns $x_i \simeq x(t_i)$ and $u_i \simeq u(t_i)$. And we write $X = (x_0, \dots, x_N)$ and $U = (u_0, \dots, u_N)$.

We also consider a quadrature formula for J and $\dot{x} = f(x, u)$, i.e.

$J(T, x, u) \simeq \tilde{J}(T, X, U)$ and $\dot{x} = f(t, x, u)$ becomes constraints of the type $c_i(X, U) = 0$ for every $i \in \{1, \dots, N\}$.

Full discretization II

Thus the minimisation problem (16') becomes:

$$\begin{array}{l} \min \quad \tilde{J}(T, X, U) \\ \left| \begin{array}{l} c_i(X, U) = 0 \quad i = 1, \dots, N \\ u_i \in \Omega \quad i = 0, \dots, N \\ x_0 = ?, \quad x_N = ? \end{array} \right. \end{array} \quad (16')$$

which is a finite dimensional minimisation problem with constraints.

This problem can be solved with a penalty method or with a sequential quadratic programming method.

Remark

Other methods are based on Halminton-Jacobi equation.

We refer to Bardi and Capuzzo-Dolcetta, [Optimal control and viscosity solutions of Hamilton-Jacobi-Bellman equation](#), Birkhäuser, 1997.

Thanks for you attention