# Second-order optimality conditions for state-constrained optimal control problems

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#### An academic example

- Framework
- 2 General unconstrained problems
  - Optimality conditions
  - Well-posedness

#### 3 State constraints

- Framework
- (Alternative) optimality condition

## ④ Sensitivity

- Sensitivity: framework
- Main result

Framework

## Data of the academic example

(*P*) Min 
$$\int_0^1 \left(\frac{1}{2}u^2(t) + g(t)y(t)\right) dt$$
  
s.t.  $\dot{y}(t) = u(t), \quad y(0) = y(1) = 0, \quad y(t) \ge h$ 

with

$$g(t) := (c - \sin(\alpha t))g_0, \qquad c > 0, \ \alpha > 0.$$

Time viewed as second state variable ( $\dot{\tau} = 1$ )  $\mu = (h - h_0)/(h_1 - h_0)$  homotopy parameter;  $h_0 = \min \bar{y}(t)$ , where  $\bar{y}$  is the solution of unconstrained problem  $h_1 = h$  target value; numerical values are

$$g_0 := 10, \qquad \alpha = 10\pi, \qquad c = 0.1, \qquad h_1 = -0.001.$$

Framework

#### Unconstrained problem: optimal state



k = 0

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Framework

Neigborhood of limiting problem: when  $\mu > 0$  is small

For  $\mu > 0$  the state constraint is active (convex problem)

The contact set could be then for small  $\mu > 0$ :

- One point
- A small interval
- A non connected set

Your guess ?

Framework

Neigborhood of limiting problem: when  $\mu > 0$  is small

For  $\mu > 0$  the state constraint is active (convex problem)

- Structural result: the contact set is an interval
- Quantitative result: first-order expansion of value of extreme points of that interval !

Next: numerical results using a shooting algorithm that will be presented later.

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An academic example General unconstrained problems State constrainty Sensitivity Numerical results ||



k = 1

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k = 2

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An academic example General unconstrained problems State constraints Sensitivity Numerical results IV



k = 3

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#### Numerical results V



k = 4

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## Numerical results VI



k = 5

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Optimality conditions Well-posedness

Setting: general unconstrained problem

- State equation:  $y(t) \in \mathbb{R}^n$ ,  $u(t) \in \mathbb{R}^m$ 
  - $\dot{y}(t) = f(u(t), y(t))$  p.p.  $t \in [0, T], \quad y(0) = y_0$  (1)
- Cost function: integral + final term

$$J(u,y) = \int_0^T \ell(u(t), y(t)) dt + \phi(y(T)).$$
 (2)

• Optimal control problem

$$\operatorname{Min}_{(u,y)} J(u,y) \quad \text{s.t.} (1). \tag{P}$$

•  $C^{\infty}$ , Lipschitz data f,  $\ell$ ,  $\phi$ .

Optimality conditions Well-posedness

#### Functional spaces and costate equation

- Control space:  $\mathcal{U} := L^{\infty}(0, T; \mathbb{R}^m)$
- State and costate space  $\mathcal{Y} := W^{1,\infty}(0, T; \mathbb{R}^n), \quad \mathcal{P} := W^{1,\infty}(0, T; \mathbb{R}^{n*})$ where  $W^{1,\infty}(0, T; \mathbb{R}^n) = \{ y \in L^{\infty}(0, T; \mathbb{R}^n); \ \dot{y} \in L^{\infty}(0, T; \mathbb{R}^n) \}.$
- Hamiltonian  $H(u, y, p) := \ell(u, y) + pf(u, y)$
- Costate equation

$$-\dot{p}(t) = H_y(u(t), y(t), p(t)) \text{ p.p. } t \in [0, T],$$
  
 $p(T) = \phi'(y(T)).$ 

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Optimality conditions Well-posedness

## Pontryaguin's principle

- S(P) Solution set of (P)
- Pontryaguin's Minimum principle (PMP):  $H(u(t), y(t), p(t)) = \min_{v} H(v, y(t), p(t))$  a.a. t
- Weak PMP:

$$H_u(u(t),y(t),p(t))=0$$
 a.a.  $t$ 

Theorem: If  $u \in S(P)$ , and y and p are the associated state and costate, then it satisfies the PMP.

Consequence: weak PMP and also:

 $H_{uu} := H_{uu}(u(t), y(t), p(t))$  is semidefinite positive.

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Optimality conditions Well-posedness

## Elimination of control

• Assume: (A1) Strong Legendre condition (W/SLC)

 $H_{uu}(u(t), y(t), p(t)) \succeq \alpha I_d, \quad \text{ for some } \alpha > 0$ 

• By IFT: weak PMP locally equivalent to:

 $u(t) = \Upsilon(y(t), p(t))$ 

with  $\Upsilon$  of class  $C^{\infty}$ 

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**Optimality conditions** Well-posedness

## Shooting mapping

• TPBVP Two Point Boundary Value Problem

$$\dot{y} = f(\Upsilon(y, p), y)$$
 p.p. [0, T],  $y(0) = y_0$ 

$$-\dot{p} = H_y(\Upsilon(y,p),y,p)$$
 p.p.  $[0,T], p(T) = \phi_y(y(T)).$ 

• Shooting function  $\mathbb{R}^{n*} \mapsto \mathbb{R}^{n*} : p_0 \mapsto p(T) - \phi_y(y(T))$ , where (y, p) solution of the Cauchy problem

$$\dot{y} = f(\Upsilon(y, p), y) \quad \text{p.p. } t \in [0, T], \quad y(0) = y_0 -\dot{p} = H_y(\Upsilon(y, p), y, p) \quad \text{p.p. } [0, T], \quad p(0) = p_0.$$

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Optimality conditions Well-posedness

# Definition

- Shooting mapping well posed at solution point  $p_0$  if it has an invertible Jacobian at this point. Then:
- By IFT: well-posedness under small perturbation
- Newton's method converges locally quadratically

• Question: Is it satisfied under weak hypotheses ?

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Optimality conditions Well-posedness

## Tangent quadratic problem

$$\min_{v} J'(u)v + \frac{1}{2}J''(u)(v,v)$$
 (TQP)

Quadratic Growth condition QGC: for some arepsilon > 0 and lpha > 0

$$J(u+v) \geq J(u) + \alpha \|v-u\|_2^2 \quad \text{if } \|v-u\|_{\infty} < \varepsilon.$$

- Second Order Necessary Condition (SONC): v = 0 solution of (TQP) (interpretation)
- Second Order Sufficient Condition (SOSC): v = 0 unique solution of (TQP) and Strong Legendre Condition (definition)

Theorem: SOSC equivalent to QGC. These conditions imply that the shooting algorithm is well-posed.

Framework (Alternative) optimality condition

## State constrained problem: data

• State constraint:

$$g_i(y(t)) \leq 0, \quad t \in [0, T], \ i = 1, \dots, r.$$
 (3)

• Same cost function: integral + final term

$$J(u, y) = \int_0^T \ell(u(t), y(t)) dt + \phi(y(T)).$$
 (4)

• Optimal control problem

$$\min_{(u,y)} J(u,y) \quad \text{s.t. (1) and (3).}$$
(P)

•  $C^{\infty}$ , Lipschitz data f,  $\ell$ ,  $\phi$ , g.

Framework (Alternative) optimality condition

#### Constraint structure

• Contact set:  $\{t \in [0, T] ; g(y(t)) = 0\}.$ 



boundary arc  $[\tau_{en}, \tau_{ex}]$  (isolated) touch point  $\{\tau_{to}\}$ • Question: If known structure: number, ordering of boundary arcs and touch points; Then can we design a shooting algorithm ? Will it be well-posed ?

Framework (Alternative) optimality condition

## Junction points

- Set of junction points: closure of end-points of interior arcs
- Regular junction point: end-point of two arcs, of three types:
- Entry, exit points: end-points of a boundary arc
- Touch point: isolated contact points

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Framework (Alternative) optimality condition

## Homotopy

- : Constraint structure generally unknown
- Possible approach: start from an unconstrained perturbed problem and compute a path with endpoints the solutions of the perturbed and original problem.
- Example:  $g^{\mu}(y) = g(y) (1 \mu)K$ , K "large".
- Motivates the local study of structural changes.
- Work by Oberle, Gergaud, Caillau, Martinon ...

Framework (Alternative) optimality condition

#### Multipliers are measures

- Lagrange multiplier  $\eta \in M(0, T)$
- Lagrangian function

$$L(u,\eta) := J(u) + \int_0^T g(y_u(t)) \mathrm{d}\eta(t)$$

Slater qualification condition:  $G(u) = g(y_u)$ 

G(u)+G'(u)v<0 on [0,T], for some  $v\in\mathcal{U}.$ 

• Complementarity conditions

$$N(u) := \left\{ \eta \in M(0, T)_+; \quad \int_0^T g(y_u(t)) = 0 \right\}.$$

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Framework (Alternative) optimality condition

## Costate equation

• Costate equation

$$\begin{aligned} -\mathrm{d} \boldsymbol{p}(t) &= H_y(\boldsymbol{u}(t), \boldsymbol{y}(t), \boldsymbol{p}(t)) \mathrm{d} t + \mathrm{d} \boldsymbol{\eta}(t) \boldsymbol{g}'(\boldsymbol{y}(t)), \quad \text{p.p. } t \\ \boldsymbol{p}(T) &= \phi'(\boldsymbol{y}(T)). \end{aligned}$$

- Weak Pontryaguin principle (WPMP)  $H_u(u(t), y(t), p(t)) = 0$  for a.a.  $t; \quad \eta \in N(u).$
- Then: call  $\eta$  a Lagrange multiplier; denote  $\eta \in \Lambda(u)$ .

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Framework (Alternative) optimality condition

## Pontryaguin's principle

• S(P) Solution set of (P)

• Minimum principle (PMP):  $H(u(t), y(t), p(t)) = \underset{v}{\min} H(v, y(t), p(t))$  a.a. t for some  $\eta \in \Lambda(u)$ .

Theorem: Let  $u \in S(P)$  be qualified, y associated state. Then (i) The set  $\Lambda(u)$  is non empty and bounded. (ii) There exists  $\eta \in N(u)$  for which the PMP holds.

Consequence: For the  $(p, \eta)$  satisfying the PMP, we have  $H_{uu} := H_{uu}(u(t), y(t), p(t))$  semidefinite positive.

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Framework (Alternative) optimality condition

Order of the state constraint

• Total derivative of a scalar state constraint:

$$g^{(1)}(u,y) := g'(y)f(u,y).$$

While result does not depend on u, we can continue:

$$g^{(i+1)}(u,y) := g^{(i)}(y)f(u,y).$$

Constraint order: q smallest number such that

$$g_u^{(q)}(u,y)\neq 0$$

Well-posed constraint order: when

$$g_u^{(q)}(u,y) \neq 0,$$
 for all  $(u,y)$ 

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Framework (Alternative) optimality condition

## Algebraic variables

- Two algebraic variables
  - u,  $\dot{\eta}$  (density, if it exists)
- Algebraic relations: interior arc

$$H_u(u(t), y(t), p(t)) = 0; \quad \dot{\eta} = 0.$$

• Algebraic relations: boundary arc

$$H_u(u(t), y(t), p(t)) = 0; \quad g^{(q)}(u, y) = 0.$$

Not well-posed in the latter case:  $\dot{\eta}$  does not appear.

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Framework (Alternative) optimality condition

#### First step of the alternative formulation I

#### • Costate equation

$$\begin{aligned} -\mathrm{d}\boldsymbol{p}(t) &= H_y(\boldsymbol{u}(t),\boldsymbol{y}(t),\boldsymbol{p}(t))\mathrm{d}t + \mathrm{d}\boldsymbol{\eta}(t)\boldsymbol{g}'(\boldsymbol{y}(t)), \quad \text{p.p. } t\\ \boldsymbol{p}(T) &= \phi'(\boldsymbol{y}(T)). \end{aligned}$$

• Write costate dynamics as:

$$-\mathrm{d}(p+\eta g'(y)) = [H_y(u,y,p)-\eta g''(y)f(u,y)]\mathrm{d}t$$

• First alternative costate and multiplier:  $p^1 = p + \eta g'(y); \quad \eta^1 = -\eta$ 

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Framework (Alternative) optimality condition

## First step of the alternative formulation II

• The alternative costate  $p^1$  has bounded derivatives. It is solution of the differential equation

$$\begin{aligned} -\dot{p}^1 &= \ell_y(u, y) + pf_y(u, y) + \eta^1 g''(y) f(u, y) \\ &= \ell_y(u, y) + p^1 f_y(u, y) + \eta^1 [g'(y) f_y(u, y) + g''(y) f(u, y)] \end{aligned}$$

• The bracket on r.h.s. is a partial derivative w.r.t. y:

$$g_{y}^{(1)}(u, y) = g'(y)f(u, y)$$
  

$$g_{y}^{(1)}(u, y) = g'(y)f_{y}(u, y) + g''(y)f(u, y).$$

• We recognize a Hamiltonian system!

Framework (Alternative) optimality condition

Alternative costate equation

• First alternative Hamiltonian

$$H^1(u, y, p^1, \eta^1) := \ell(u, y) + p^1 f(u, y) + \eta^1 g^{(1)}$$

• Alternative costate equation

$$-\dot{p}^1 = H^1_y(u, y, p^1, \eta^1); \quad p^1(T) = cst + \phi'(y(T)).$$

• Alternative Pontryaguin's principle: since

$$H^{1}(u, y, p^{1}, \eta^{1}) := \ell(u, y) + (p^{1} + \eta^{1}g'(y))f(u, y) = H(u, y, p),$$

Weak/strong Pontryaguin's principle is invariant, e. g.:

$$H^1_u(u, y, p^1, \eta^1) = 0$$

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Framework (Alternative) optimality condition

## First alternative algebraic relations

• Boundary arcs (e.g. when all constraints active): obtain

$$g^{(q)}(u,y) = 0;$$
  $H_u(u,y,p^1) + \eta^1 g_u^{(1)} = 0.$ 

• Case of scalar control, scalar first-order state constraint: Elimination of algebraic variables holds !

$$u = \Psi^{q}(y); \quad \eta^{1} = -H_{y}(u, y, p^{1})/g_{u}^{(1)}$$

• Unconstrained arcs

$$u = \Psi(y, p^1, \eta^1); \quad \dot{\eta}^1 = 0.$$

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#### General first-order constraints

• Boundary arcs, all constraints active: obtain

$$\begin{aligned} & H_u(u, y, p^1) + \eta^1 g_u^{(1)}(u, y) = 0; \quad g^{(1)}(u, y) = 0. \\ & \text{Jacobian:} \begin{pmatrix} H_{uu}^1 & (g_u^{(1)})^\top \\ g_u^{(1)} & 0 \end{pmatrix} \end{aligned}$$

• Invertible iff

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$$g^{(1)}_u(u,y)$$
 onto.;  $H^1_{uu}=H_{uu}$  invertible on  $\operatorname{Ker} g^{(1)}_u$ 

• So under weak hypotheses we can eliminate algebraic variables, even in the "vector case"

Framework (Alternative) optimality condition

## References for alternative formulation

- Bryson Denham, Dreyfus (1963): provided the idea
- Maurer (1979), unpublished: rigorous derivation
- Several related works by Maurer and Malanowski
- Ref. FB and A. Hermant, INRIA Rep. 6199, 2007 Equivalence with PMP, general vector case

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Framework (Alternative) optimality condition

## Continuity of control I

- Hyp  $H_{uu}(\cdot, y(t), p(t))$  unif. invertible
- $u^-$ ,  $u^+$  values just before, after time  $\tau$ .
- Jump of multiplier at time  $\tau$ :

$$[p(\tau)] = -\nu g'(y(\tau)); \quad \nu := -[\eta(\tau)]$$

• Is *u* continuous ? assume *H* strongly convex w.r.t. *u* 

$$\Delta := H_u(u^-, y, p^+) - H_u(u^-, y, p^-) = -\nu g'(y(\tau))f_u(u^-, y).$$

- u continuous iff  $\nu g'(y(\tau))f_u(u^-, y) = 0$ .
- Holds if all constraints of order > 1. What about order 1 ?

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Framework (Alternative) optimality condition

## Continuity of control II

• Contribution of first-order terms: take  $\ell = 0$ 

$$0 = H_u(u^+, y, p^+) - H_u(u^-, y, p^-) = \int_0^1 [H_{uu}()[u] + [p]f_u()] dt$$

• Since  $H_{uu}()$  is uniformly positive:

$$\begin{aligned} \alpha |[u]|^2 &\leq \int_0^1 H_{uu}()([u], [u]) \mathrm{d}t \\ &= \nu g'(y) \int_0^1 f_u()[u] \mathrm{d}t = \nu g'(y)[f] \\ &= \nu [g^{(1)}] \leq 0 \end{aligned}$$

therefore u is continuous.

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Framework (Alternative) optimality condition

Contribution of mixed state-control constraint

• Mixed state-control constraint

$$c(u,y) \leq 0.$$

• Similar computations give:

$$\begin{aligned} \alpha |[u]|^2 &\leq \int_0^1 H_{uu}()([u], [u]) \mathrm{d}t \\ &= \nu g'(y) \int_0^1 f_u()[u] \mathrm{d}t - [\lambda] \int_0^1 c_u(u, y)[u] \mathrm{d}t \\ &= \nu g'(y) [f] - [\lambda] [c(u, y)] \\ &= \nu [g^{(1)}] - [\lambda] [c(u, y)] \leq 0 \end{aligned}$$

therefore again *u* is continuous.

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Framework (Alternative) optimality condition

## Smoothness of control at junction points

- Scalar state constraint of order 1 or 2: *u* continuous.
- Scalar state constraint of order q ≥ 3: q − 2 continuous derivatives (q − 1 if q is odd).
- Ref: Jakobson et al., 1971; Maurer, 1979.
- vector case much more involved, see FB and A. Hermant, 2007.
- No example of "generic" regular junction known when  $q \ge 3$ .

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Sensitivity: framework Main result

## Sensitivity: Framework

$$(P^{\mu}) \qquad \min_{\substack{(u,y)\in\mathcal{U}\times\mathcal{Y}\\ y\in\mathcal{U}\times\mathcal{Y}}} \int_{0}^{T} \ell^{\mu}(u(t), y(t)) dt + \phi^{\mu}(y(T))$$
  
s.c.  $\dot{y}(t) = f^{\mu}(u(t), y(t)) \text{ p.p. } [0, T] ; y(0) = y_{0}^{\mu}$   
 $g^{\mu}(y(t)) \leq 0 \text{ on } [0, T].$ 

- Rem. : scalar control  $u(t) \in \mathbb{R}$ .
- $\mu$  : perturbation parameter
- Hyp (A0) smooth data:  $C^{\infty}$ , Lipschitz (A1)  $g^{\mu_0}(y_0^{\mu_0}) < 0$ .

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Hypotheses |

Sensitivity: framework Main result

 $(\bar{u}, \bar{y})$  solution for  $\mu = \mu_0$ , with multipliers  $(\bar{p}, \bar{\eta})$ . (A2)  $H^{\mu_0}(\cdot, \bar{y}(t), \bar{p}(t))$  uniformly strongly convex (A3) (Order 1 constraint) for all t:

 $|g_u^{(1)}(u(t), y(t))| \geq \gamma > 0.$ 

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Hypotheses II

(A4)  $(\bar{u}, \bar{y})$  has a finite number of regular junctions. (A5) Strict complementarity on boundary arcs:

 $rac{\mathrm{d}ar\eta(t)}{\mathrm{d}\,t} \geq eta > 0, \quad ext{ on interior boundary arcs.}$ 

Sensitivity: framework

(A6) For all touch point (isolated contact point)  $\tau$ ,

$$\frac{\mathrm{d}^2}{\mathrm{d}t^2}g(\bar{y}(t))|_{t=\tau}<0.$$

Sensitivity: framework Main result

## Notion of quadratic growth condition

We say that the Quadratic Growth Condition (QGC) holds) holds if, for all  $C^2$ -perturbation  $(P^{\mu})$  of  $(P^{\mu_0})$ , there exists a neighborhood  $(V_u, V_{\mu})$  of  $(\bar{u}, \mu_0)$ , such that for  $\mu \in V_{\mu}$ , there exists a unique local solution  $(u^{\mu}, y^{\mu})$  of  $(\mathcal{P}^{\mu})$  with  $u^{\mu} \in V_u$ satisfying the QGC  $\exists c, r > 0$  such that

$$J^{\mu}(u, y) \ge J^{\mu}(u^{\mu}, y^{\mu}) + c(\|u - u^{\mu}\|_{2}^{2} + \|y - y^{\mu}\|_{1,2}^{2}),$$
  
$$\forall (u, y) \text{ feasible for } (P^{\mu}), \ \|u - \bar{u}\|_{\infty} + \|y - \bar{y}\|_{1,\infty} < r.$$

Sensitivity: framework Main result

#### Main result: statement

#### Theorem

Let  $(\bar{u}, \bar{y}) = (u^{\mu_0}, y^{\mu_0})$  local solution of  $(P^{\mu_0})$  satisfying (A1)-(A6). The the following statements are equivalent:

(i) The QGC holds

(ii) The following second-order sufficient condition is satisfied: The tangent linear-quadratic problem (defined later) has v = 0 as unique solution.

Under these conditions: local uniqueness of local solutions in U. Also: Boundary arcs are stable,

Touch point remain so, vanish or become boundary arcs.

Sensitivity: framework Main result

# Main result (continued)

#### Theorem (End of statement)

... If (i) or (ii) is satisfied, then  $\mu \mapsto (u^{\mu}, y^{\mu}, p^{\mu}, \eta^{\mu})$  is locally Lipschitz in

$$\mathcal{U} \times \mathcal{Y} \times L^{\infty}(0, T; \mathbb{R}^{n*}) \times L^{\infty}(0, T; \mathbb{R})$$

and directionally differentiable in

 $L^{\mathbf{r}}(0,T) \times W^{1,\mathbf{r}}(0,T;\mathbb{R}^n) \times L^{\mathbf{r}}(0,T;\mathbb{R}^{n*}) \times L^{\mathbf{r}}(0,T)$ 

for all  $1 \leq r < \infty$ . The directional derivative in direction d is the unique solution of a certain linear quadratic problem  $(P_d)$ .

Sensitivity: framework Main result

## The linear quadratic problem

Space of linearized control and states

 $\mathcal{V} := L^2(0, T) \supset \mathcal{U}; \qquad \qquad \mathcal{Z} := H^1(0, T; \mathbb{R}^n) \supset \mathcal{Y}.$ 

 $\textit{d} = \mu - \mu_{0}$  : "given" direction of perturbation.

$$(\mathcal{P}_{d}) \qquad \min_{(v,z)\in\mathcal{V}\times\mathcal{Z}} \quad \frac{1}{2} \{ \int_{0}^{T} D_{(u,y,\mu)^{2}}^{2} H^{\mu_{0}}(\bar{u},\bar{y},\bar{p})(v,z,d)^{2} \mathrm{d}t \\ + D^{2}\phi^{\mu_{0}}(\bar{y}(T))(z(T),d)^{2} + \int_{0}^{T} D^{2}g^{\mu_{0}}(\bar{y},\mu_{0})(z,d)^{2} \mathrm{d}\bar{\eta}(t) \}$$

s.c. 
$$\dot{z}(t) = Df^{\mu_0}(\bar{u}, \bar{y})(v, z, d)$$
 sur  $[0, T]$ ,  $z(0) = Dy_0^{\mu_0} d$   
 $Dg^{\mu_0}(\bar{y})(z, d) = 0$  on boundary arcs of  $(\bar{u}, \bar{y})$   
 $Dg^{\mu_0}(\bar{y}(\tau))(z(\tau), d) \leq 0$ ,  $\forall \tau$  isolated contact point of  $(\bar{u}, \bar{y})$ .

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## Algorithmic consequences

- If no isolated touch point: Newton's method well-defined (with the "shooting parameters, see paper)
- Convergent homotopy algorithm taking into account transitions
- Touch point viewd as zero lenght boundary arc
- Backtracking over  $\mu$  if Newton's method non convergent.

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Expression of linearization of entry times

Linearize

$$\hat{g}^{(1)}(\bar{u}(\bar{t}^{en}),\bar{y}(\bar{t}^{en}),\mu_0)=0$$

Denote by v, z,  $\sigma^{en}$  the directional derivative of control, state, entry point w.r.t. a variation of  $\mu$  in direction d, then

$$\sigma^{en} = -\frac{D\hat{g}^{(1)}(\bar{u}(\bar{t}^{en}), \bar{y}(\bar{t}^{en}), \mu_0)(v(\bar{t}^{en-}), z(\bar{t}^{en}), d)}{\frac{\mathrm{d}}{\mathrm{d}t}g^{(1)}(\bar{u}, \bar{y})|_{t=\bar{t}^{en-}}}$$

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# Challenges

What happens when:

- A boundary arc splits into two ?
- Two boundary arcs split into one ?
- second-order derivative at a touch point is zero ?

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# References

- J.F. B. and A.Hermant, *Stability and Sensitivity Analysis for Optimal Control Problems with a First-order State Constraint*. INRIA report RR-5889. ESAIM:COCV, to appear.
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Sensitivity: framework Main result

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