Properties of Minimizers that are not also Relaxed Minimizers

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Original Problem

Consider the Optimal Control Problem:

$$(P) \begin{cases} \text{minimize } g(x(0), x(1)) \\ \text{over the } x \in W^{1,1}([0,1]; \mathbb{R}^n) \text{ s.t.} \\ \dot{x}(t) \in F(t, x(t)) \text{ a.e.} \\ (x(0), x(1)) \in C \end{cases}$$

- (H1): $C \subset \mathbb{R}^n \times \mathbb{R}^n$ is closed and $g : \mathbb{R}^n \times \mathbb{R}^n \to \mathbb{R}^n$ is locally Lipschitz continuous;
- (H2): F(⋅, ⋅) takes values closed sets and F(⋅, x) is measurable for each x;

• For a nominal trajectory $\bar{x}(.)$

(H3): There exist $k(.), c(.) \in L^1$, and $\varepsilon > 0$ s.t.: I) $F(t,x) \subset F(t,x') + k(t)|x - x'|B$ II) $F(t,x) \subset c(t)B$ for all $x, x' \in \overline{x}(t) + \varepsilon B$, a.e. $t \in [0, 1]$.

Define the maximized Hamiltonian:

$$H(t, p, x) = \max_{v \in F(t, x)} (p \cdot v)$$

Local minimizers

An *F*-trajectory $\bar{x}(.)$ is a **strong** local minimizer for (*P*) if, for $\varepsilon > 0$,

$$g(\bar{x}(0), \bar{x}(1)) \leq g(x(0), x(1))$$

for every

$$||x(.)-\bar{x}(.)||_{L^{\infty}} \leq \varepsilon.$$

 $\bar{x}(.)$ is a **weak** local minimizer if

$$g(\bar{x}(0),\bar{x}(1)) \leq g(x(0),x(1))$$

for every

$$||x(.)-\bar{x}(.)||_{W^{1,1}} \leq \varepsilon.$$

Existence of a Minimizer

Theorem: Suppose hypotheses (H1), (H2) and (H3) and assume, furthermore, the following conditions:

- if we write $C = C_0 \times C_1$, then or C_0 either C_1 is bounded;
- the multifunction $x \mapsto F(t, x)$ has convex value a.e. $t \in [0, 1]$.

Then (P) has a minimizer.

Relaxed Control Problem

Let's introduce the relaxed problem:

(R)
$$\begin{cases} \text{minimize } g(x(0), x(1)) \\ \text{over the } x \in W^{1,1}([0,1]; \mathbb{R}^n) \text{ s.t.} \\ \dot{x}(t) \in \operatorname{co} F(t, x(t)) \text{ a.e.} \\ (x(0), x(1)) \in C \end{cases}$$

Relaxation Theorem: Assume the hypotheses (H1) - (H3) and take any co*F*-trajectory x and a $\delta > 0$. Then there exists an *F*-trajectory y such that satisfies y(0) = x(0) and

$$\max_{t\in[0,1]}|y(t)-x(t)|<\delta.$$

Reachable Set Interpretation

Define the original Reachable set

 $\mathcal{R} := \{ (x(0), x(1)) : \dot{x}(t) \in F(t, x(t)) \text{ a.e. } t \in [0, 1] \},\$

and the relaxed Reachable set

 $S := \{ (x(0), x(1)) : \dot{x}(t) \in coF(t, x(t)) \text{ a.e. } t \in [0, 1] \}.$

Then it turns out that

$$\bar{\mathcal{R}} = \mathcal{S}.$$

Why is Relaxation useful?

A standard relaxation procedure concerns:

- **Step 1:** Solve the relaxed problem (which always has a solution).
- Step 2: Approximate the solution to the relaxed problem as closely as required to obtain a sub-optimal solution to the original problem.

Infimum Gap

The relaxation procedure works if we assume that:

 $\inf\{P\} = \inf\{R\}.$

But, in certain pathological situations, it could happen that

$$\inf\{P\} > \inf\{R\}.$$
 (1)

These situations are problematic because we cannot obtain a suboptimal minimizers for (P). Furthermore, we encounter also a breakdown of the **Dynamic Pro**gramming method and HJB approach.



Infimum Gap: An Example

Consider the problem:

$$(E) \begin{cases} \text{Minimize} - x_1(1) \\ \text{over } x(.) = (x_1(.), x_2(.), x_3(.)) \text{ satisfying} \\ \dot{x}_1(t) = 0 \\ \dot{x}_2(t) = x_1(t)u(t) \\ \dot{x}_3(t) = |x_2(t)|^2 \\ u(t) \in U(t) \equiv \{-1\} \cup \{+1\} \\ x_2(0) = x_3(0) = x_3(1) = 0 . \end{cases}$$

It turns out that $\bar{x}(.) = (0,0,0)$ is an original strong local minimizer but is **not** also a relaxed minimizer. Indeed, every trajectory $\tilde{x}(.) = (\alpha, 0, 0)$ is a relaxed strong local minimizer for the bound

$$||x(.)||_{L^{\infty}} \leq \alpha$$

Abnormality of Multipliers

A trajectory $\bar{x}(.)$ is **abnormal** if there exists a couple $(\lambda, p(.))$ which satisfies some set of Necessary Conditions with $\lambda = 0$.

We relate the "infimum gap" (1) to abnormality of multipliers.

Assume "infimum gap" (1). Then

(Type A): Any **original** minimizers for (P) is abnormal.

(Type B): Any **relaxed** trajectory which satisfies (1) is abnormal.

Some Results

Type A Theorem (Palladino-Vinter), 2013 Let $\bar{x}(\cdot)$ be a strong local minimizing *F*-trajectory for

(*P*). Assume (*H*1), (*H*2) and (*H*3). Suppose that $\bar{x}(.)$ is **not** also a minimizer for (*R*).

Then there exists $p(\cdot) \in W^{1,1}([0,1]; \mathbb{R}^n)$ such that:

i) $((\lambda = 0), p(\cdot)) \neq (0, 0);$

 $ii) \ (-\dot{p}(t), \dot{ar{x}}(t)) \in \mathsf{CO}\partial_{x,p}H(t, ar{x}(t), p(t))$ a.e.;

iii) $(p(0), -p(1)) \in (\lambda = 0) \partial g(\bar{x}(0), \bar{x}(1)) + N_C(\bar{x}(0), \bar{x}(1)).$

Type B Theorem (Palladino-Vinter), 2013 Let $\bar{x}(\cdot)$ be a feasible co*F*-trajectory. Assume (*H*1), (*H*2) and (*H*3). Suppose also

(H4) : there exist $\varepsilon > 0$ and $\delta > 0$ such that

 $g(x(0), x(1)) > g(\bar{x}(0), \bar{x}(1)) + \delta$

for every feasible F-trajectory $x(\cdot)$ such that

$$||x(\cdot) - \bar{x}(\cdot)||_{\infty} \leq \varepsilon.$$

Then there exists $p(\cdot) \in W^{1,1}([0,1];\mathbb{R}^n)$ such that i) $((\lambda = 0), p(\cdot)) \neq (0,0);$

$$ii)~(-\dot{p}(t),\dot{ar{x}}(t))\in\mathsf{CO}\partial_{x,p}H(t,ar{x}(t),p(t))$$
 a.e.;

iii) $(p(0), -p(1)) \in (\lambda = 0) \partial g(\bar{x}(0), \bar{x}(1)) + N_C(\bar{x}(0), \bar{x}(1)).$



Type A Theorem

Type B Theorem

Sketch of the Proof for Type B Th.

- Use the Relaxation Theorem to define an original (non feasible) sequence $x_i \to \overline{x}$ in L^{∞} .
- Construct a sequence of perturbed problems (respect to x_i) and use the Stegall's Variational Principle.
- Pass to the limit in the Clarke's Hamiltonian N. C.

Type B Th. \Longrightarrow Type A Th.

Remarks

If we look at the **contrapositive** statement of the Type A Theorem, it follows that normality of the minimizer implies the equality

 $\inf\{P\} = \inf\{R\}.$

Hence, **normality** of a local minimizer is a sufficient condition in order to carry out relaxation procedure of numerical schemes for control problems.

Theory easily adapts to allow **state constraints** of the form $h(t, x(t)) \leq 0$ for every $t \in [0, 1]$.

Research Questions

 Are the theorems true if we consider weak neighborhood as

$$||x(.) - \bar{x}(.)||_{W^{1,1}} \le \varepsilon$$
?

• Are they valid also for other N. C. involving different adjoint equations like

$$\dot{p}(t) \in \operatorname{CO}\{q : (q, p(t)) \in N_{\operatorname{Gr}F(t,.)}(\bar{x}(t), \dot{\bar{x}}(t))\}$$

or

$$-\dot{p}(t) \in \operatorname{co}\{q : (q, \dot{\overline{x}}(t)) \in \partial_{x,p}H(t, \overline{x}(t), p(t))\}$$
?

Concerning the above open questions, **Ioffe** claimed:

'Suppose that $\bar{x}(.)$ is a local $W^{1,1}$ minimizer which is normal with respect to

 $\dot{p}(t) \in \operatorname{co}\{q : (q, p(t)) \in N_{\operatorname{Gr} F(t,.)}(\bar{x}(t), \dot{\bar{x}}(t))\}$

Then $\bar{x}(.)$ is also a $W^{1,1}$ -relaxed minimizer'.

(A. D. Ioffe - Euler Lagrange and Hamiltonian Formalisms in Dynamic Optimization, 1997)' This claim, as stated, is incorrect. This is illustrated by the following counter-example . . .

In fact, our work reveals, to correct Ioffe's assertion, we need to consider **normality w.r.t. the Hamiltonian inclusion**, and consider **local optimality w.r.t. the** L^{∞} **norm**.

Counterexample

(R. Vinter), 2013

Consider the probem:

$$(E) \begin{cases} \text{Minimize } \left(\frac{1}{2}x_{1}(1) - x_{2}(1)\right) \\ \text{over abs. cont. } x(.) = (x_{1}(.), x_{2}(.), x_{3}(.)) \text{ s.t.} \\ \left[\frac{\dot{x}_{1}(t)}{\dot{x}_{2}(t)}\right] \in \left[\begin{array}{c} 0 \\ x_{1}(t) \\ 1 \end{array}\right] \cup \left[\begin{array}{c} 0 \\ x_{1}(t) \\ -1 \end{array}\right] \cup \left[\begin{array}{c} 0 \\ \min\{x_{1}(t), 0\} \\ 0 \end{array}\right] \\ (x_{1}(0), x_{2}(0), x_{3}(0)) \in [0, \infty) \times \{0\} \times \{0\}, \\ (x_{1}(1), x_{2}(1), x_{3}(1)) \in \mathbb{R} \times \mathbb{R} \times \mathbb{R} \end{cases}$$



It turns out that:

- Since (E) is free endpoint constraint, then the solutions are **normal**;
- $\bar{x}(.) = (0,0,0)$ is a weak local minimizer for (E) with $||x(.)||_{W^{1,1}} \le \varepsilon < 1/2$;
- The co*F*-trajectory $\tilde{x}(.)$, determined by $\tilde{x}(0) = (\varepsilon/2\sqrt{3}, 0, 0)$ and $\dot{\tilde{x}}(t) = (0, \varepsilon/2\sqrt{3}, 0)$ is feasible and yields a cost **less** than 0.

Then $\bar{x}(.) = (0, 0, 0)$ is **not** also a relaxed minimizer.

To Summarize...

Normality for (P) with Fully Convexified Hamiltonian N.C. $\Rightarrow \inf\{P\} = \inf\{R\}$ in the L^{∞} norm

Normality for (P) with Partially Convexified Hamiltonian N.C. $\Rightarrow \inf\{P\} = \inf\{R\}$ in the W^{1,1} norm

Normality for (P) with Extended Euler Lagrange N.C. $\Rightarrow \inf\{P\} = \inf\{R\}$ in the W^{1,1} norm

Open Research Question

Consider the problem:

$$(P) \begin{cases} \text{minimize } g(x(0), x(1)) \\ \text{over the absolutely continuos arcs } x(\cdot) \\ \text{and the measurable control functions } u(\cdot) \text{ s.t.} \\ \dot{x}(t) = f(t, x(t), u(t)) \quad \text{a.e. } t \in [0, 1] \\ u(t) \in U(t) \quad \text{a.e. } t \in [0, 1] \\ (x(0), x(1)) \in C \end{cases}$$

Warga proved a Type B Theorem for a Nonsmooth Maximum principle.

Is the Type A Theorem also true?

If we assume the dynamic has the structure

f(t, x, u) := A(t, x) + B(t, x)u

with

$$det(B(t,x)B^T(t,x)) \neq 0,$$

for every x, a.e. $t \in [0, 1]$, then a **Type A theorem** is valid for the Nonsmooth P.M.P. with adjoint equation

$$-\dot{p}(t)\in \mathsf{CO}\;\partial_x h(t,ar{x}(t),ar{u}(t),p(t))$$
 a.e.

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Thank you!