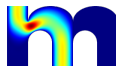


On the Existence of Optimal Worst-Case Robust Controls

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Uncertain Processes, Trier

June 5, 2009

- 1 Motivation
- 2 Problems without Robust State Constraints
- 3 Problems with Robust State Constraints

$$\begin{aligned}
 & \min_{\substack{u \in L^2(\Gamma) \\ y \in H^1(\Omega)}} && \frac{1}{2} \|y - y_d\|_{L^2(\Omega)}^2 + \frac{\gamma}{2} \|u\|_{L^2(\Gamma)}^2 \\
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 & && \partial_\nu y = p(u - y) && \text{on } \Gamma \\
 & \text{s.t.} && u_a \leq u \leq u_b && \text{a.e. on } \Gamma
 \end{aligned}$$

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 & \min_{u \in L^2(\Gamma)} \quad \sup_{\substack{p \in P_{\text{ad}} \\ y \in H^1(\Omega)}} \quad \frac{1}{2} \|y - y_d\|_{L^2(\Omega)}^2 + \frac{\gamma}{2} \|u\|_{L^2(\Gamma)}^2 \\
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Reduced Formulation

$$\min_{u \in U_{\text{ad}}} \quad \sup_{p \in P_{\text{ad}}} \quad \frac{1}{2} \|S(u, p) - y_d\|_{L^2(\Omega)}^2 + \frac{\gamma}{2} \|u\|_{L^2(\Gamma)}^2$$

where $S(u, p) \in H^1(\Omega)$ is the unique weak solution of the PDE

$$\min_{\substack{u \in L^2(\Omega) \\ y \in H^1(\Omega)}} \quad \frac{1}{2} \|y - y_d\|_{L^2(\Omega)}^2 + \frac{\gamma}{2} \|u + p\|_{L^2(\Omega)}^2$$

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Does there exist a worst-case robust optimal control?

Approximation problem $(A_0 + p A_1) u \approx b$

- **Nominal Problem**

$$\min_{u \in \mathbb{R}^n} \|A_0 u - b\|_2^2$$

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- **Stochastic Robust Problem**

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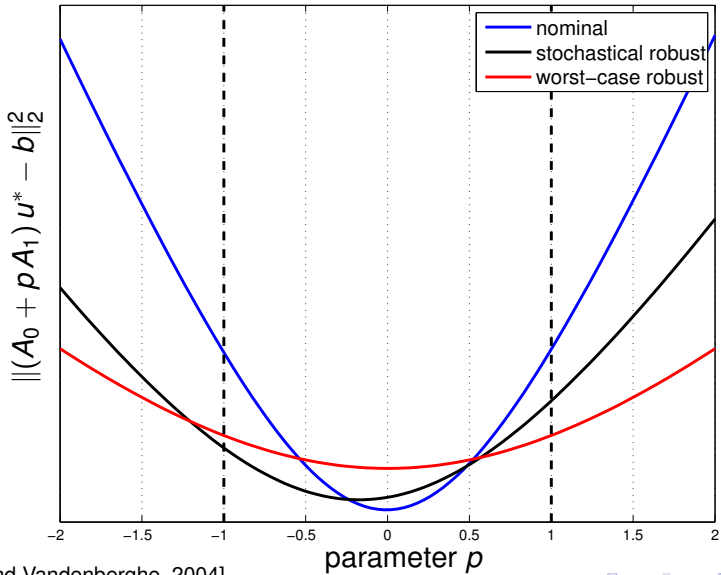
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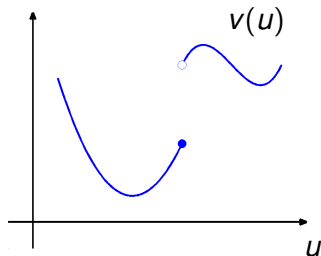
- **Worst-case Robust Problem**

$$\min_{u \in \mathbb{R}^n} \max_{p \in [-1, 1]} \| (A_0 + p A_1) u - b \|_2^2$$



[Boyd and Vandenberghe, 2004]

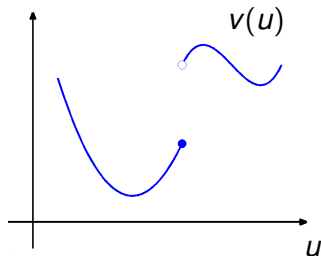
$$\min_{u \in U_{\text{ad}}} v(u) \quad (*)$$



Theorem

Let $\dim U_{\text{ad}}$ be finite. $(*)$ has a global minimizer, if

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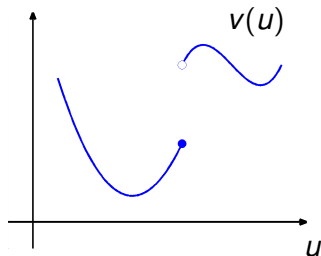


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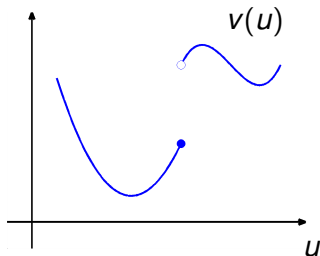


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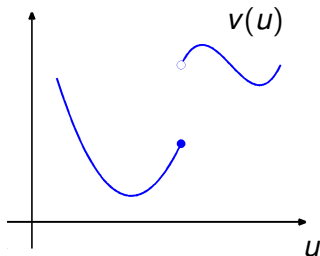


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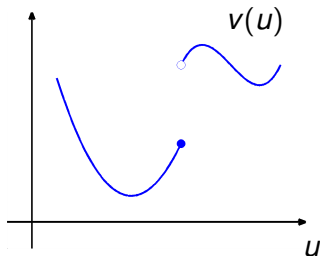


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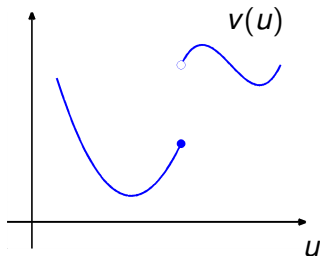


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Let $\dim U_{\text{ad}}$ be *infinite*. (*) has a global minimizer, if

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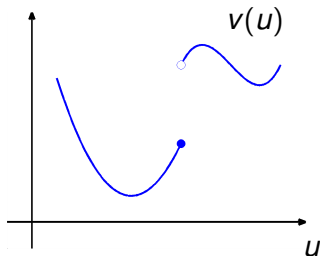


Theorem

Let $\dim U_{\text{ad}}$ be *infinite*. (*) has a global minimizer, if

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Lemma

The Optimal Value Function

$$v(u) := \sup_{p \in P_{\text{ad}}} f(u, p)$$

is weakly lower semicontinuous if $u \mapsto f(u, p)$ is weakly lower semicontinuous for all $p \in P_{\text{ad}}$.

$$\min_{u \in U_{\text{ad}}} \sup_{p \in P_{\text{ad}}} \frac{1}{2} \|S(u, p) - y_d\|_{L^2(\Omega)}^2 + \frac{\gamma}{2} \|u\|_{L^2(\Omega)}^2$$

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There exists a worst-case robust optimal control.

Proof.



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 $v(u) = \sup_{p \in P_{\text{ad}}} f(u, p)$ is weakly lower semicontinuous



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Infinite-Dimensional Semi-Infinite Program

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Infinite-Dimensional Semi-Infinite Program

Feasible Set

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Lemma

The intersection of weakly closed sets is weakly closed.

Lemma

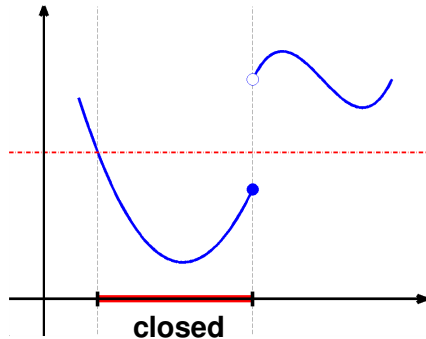
If

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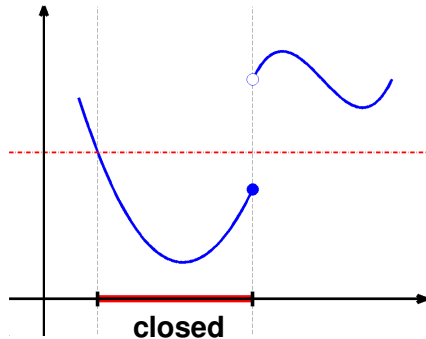
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Theorem

There exists a worst-case robust optimal control $u^* \in U_{\text{ad}}$.

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Infinite-Dimensional Semi-Infinite Program

Example (Pointwise State Constraints)

$$G : C(\bar{\Omega}) \rightarrow C(\bar{\Omega}), \quad y \mapsto y - \psi \quad \text{and} \quad K = C^+(\bar{\Omega})$$

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Infinite-Dimensional Semi-Infinite Program

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Theorem

There exists a worst-case robust optimal control $u^ \in U_{\text{ad}}$, if G is continuous*

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- 3 Necessary optimality conditions \rightsquigarrow MPCC

Thank You!