

Optimal Stochastic Impulse Control

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Joint work with **Sören Christensen** (University of Hamburg) and **Frank Seifried** (University of Trier).

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Outline

- (1) Impulse Control Problems and Quasi-Variational Inequalities
- (2) Superharmonic Functions and the Stochastic Perron Method
- (3) Applications and Work in Progress



The General Impulse Control Problem



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Consider an \mathbb{R}^n -valued **system** $X = X_{t,x}^\Lambda$

$$\begin{aligned} X(t) &= x, \\ dX(u) &= \mu(X(u))du + \sigma(X(u)) dW(u), \end{aligned}$$



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Consider an \mathbb{R}^n -valued **system** $X = X_{t,x}^\Lambda$ controlled by an **impulse control** $\Lambda = \{(\tau_k, \Delta_k)\}_{k \in \mathbb{N}}$ as follows:

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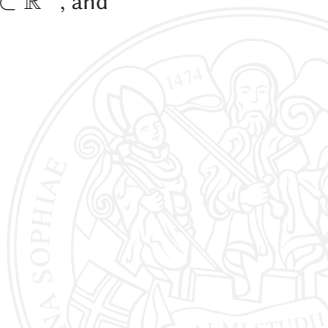
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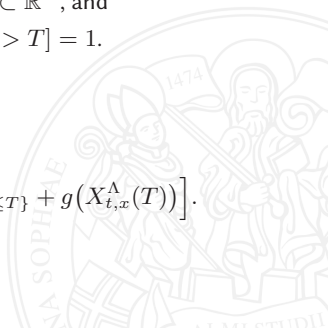
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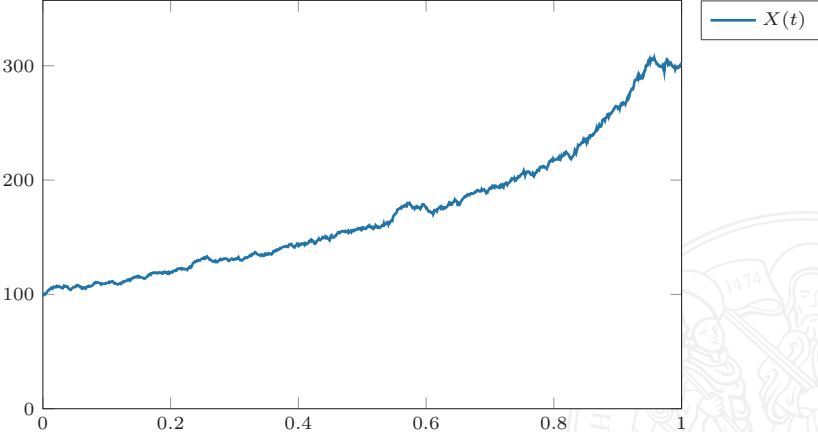
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The **objective** is to maximize

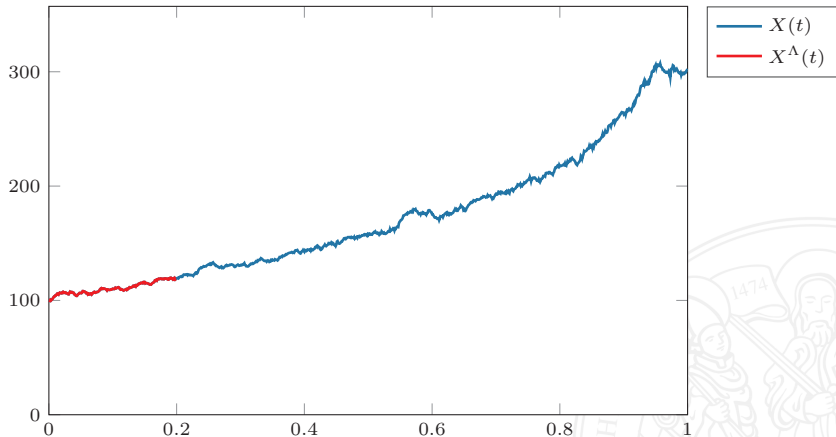
$$\mathcal{V}(t, x) = \sup_{\Lambda \in \mathcal{A}(t, x)} \mathbb{E} \left[\sum_{k \in \mathbb{N}} K(X_{t,x}^\Lambda(\tau_k-), \Delta_k) \mathbf{1}_{\{\tau_k \leq T\}} + g(X_{t,x}^\Lambda(T)) \right].$$



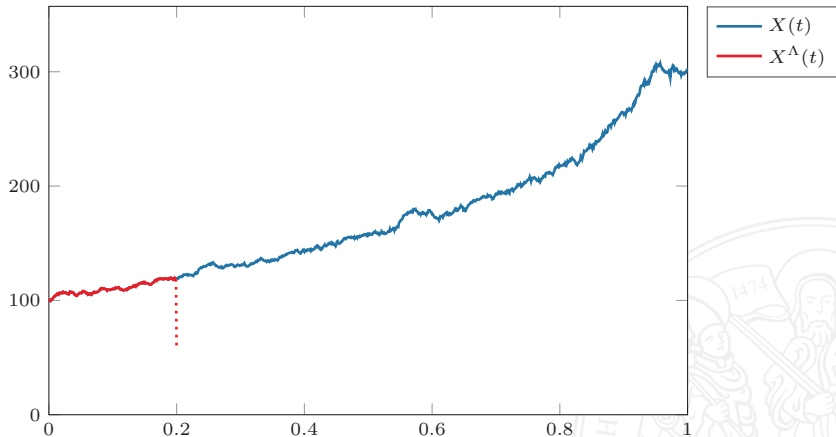
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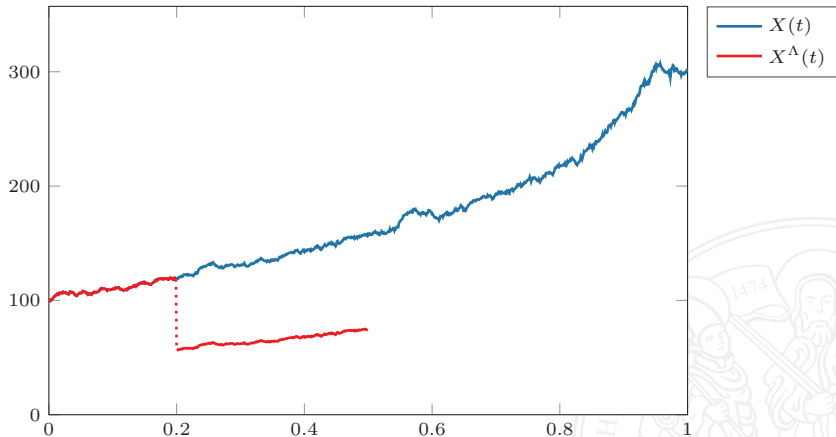
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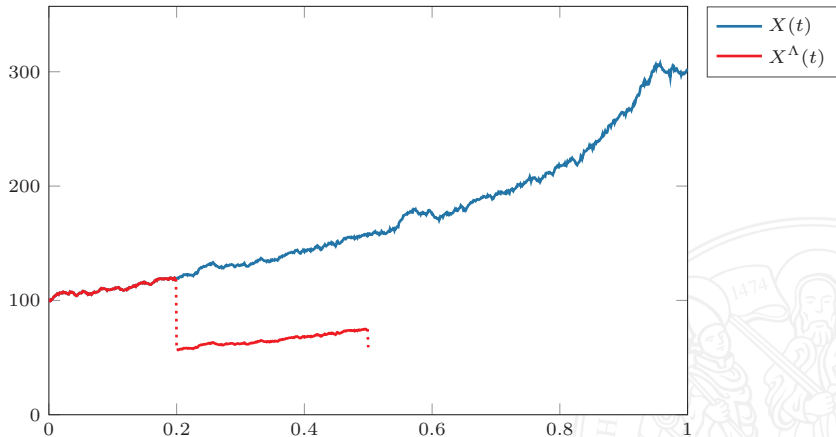
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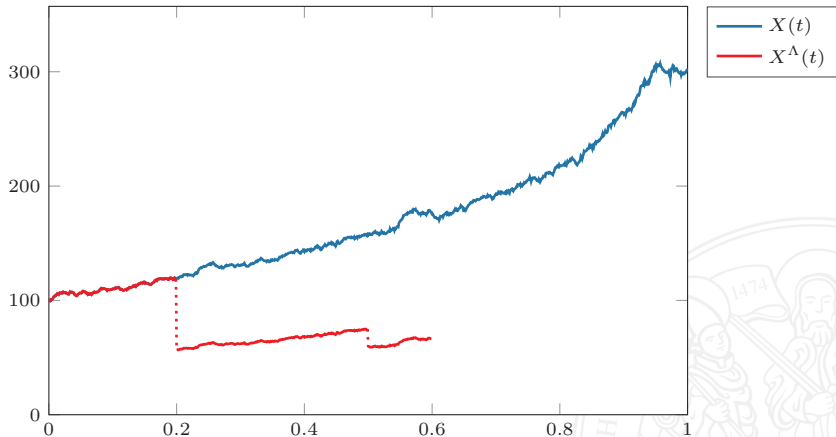
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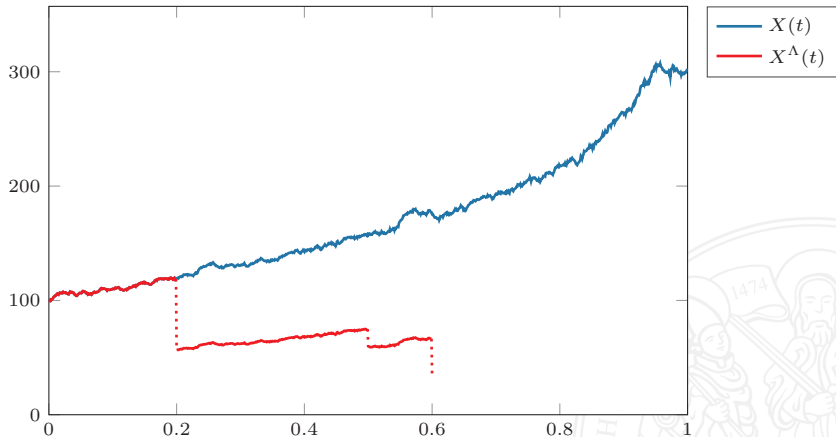
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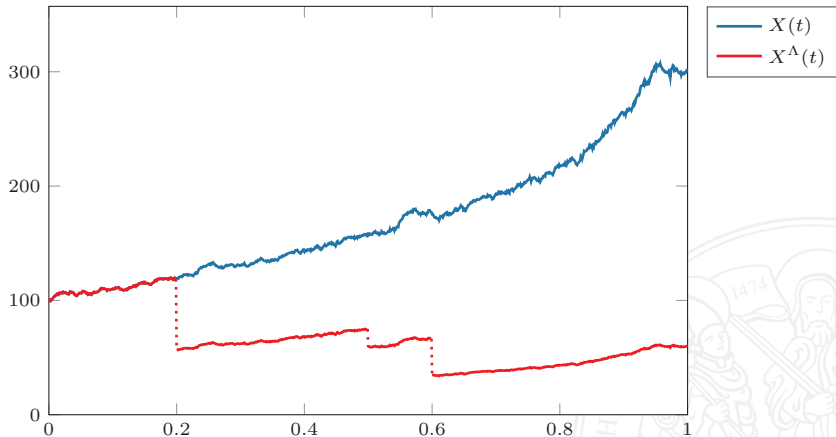
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The Quasi-Variational Inequalities



The Maximum Operator

Suppose that the system is in state x at time t

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$$(t, \Gamma(x, \Delta)) + K(x, \Delta)$$



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Suppose that the system is in state x at time t and the controller is **forced to make an impulse** and behaves optimally afterwards.

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Suppose that the system is in state x at time t and the controller is **forced to make an impulse** and behaves optimally afterwards. The best immediate impulse is

$$\sup_{\Delta \in Z(x)} \mathcal{V}(t, \Gamma(x, \Delta)) + K(x, \Delta),$$

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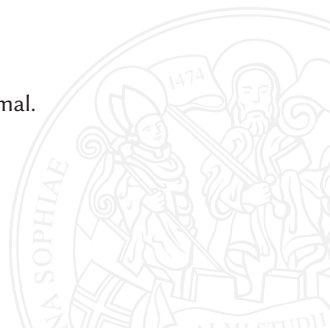
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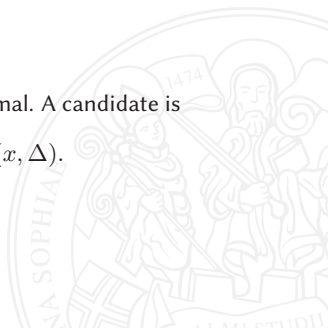
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$$\Delta^* = \arg \max_{\Delta \in Z(x)} \mathcal{V}(t, \Gamma(x, \Delta)) + K(x, \Delta).$$



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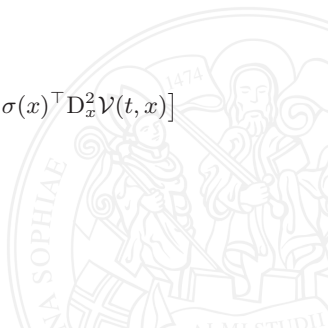
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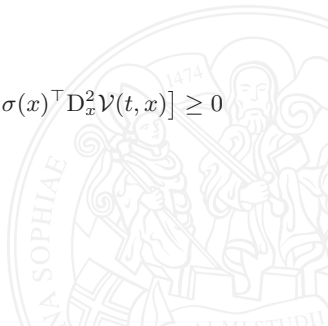
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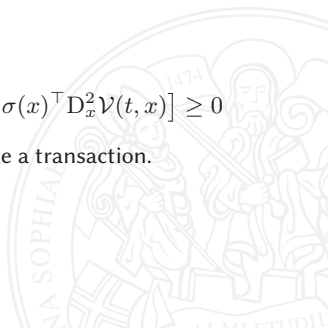
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and equality holds **if and only if** it is not optimal to make a transaction.



The Quasi-Variational Inequalities

Thus, we have argued that \mathcal{V} should solve the **quasi-variational inequalities** (QVIs)

$$\min\{\mathcal{L}\mathcal{V}(t, x), \mathcal{V}(t, x) - \mathcal{M}\mathcal{V}(t, x)\} = 0 \quad (t, x) \in [0, T) \times \mathbb{R}^n.$$



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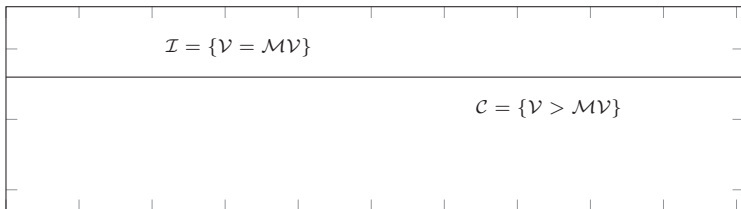
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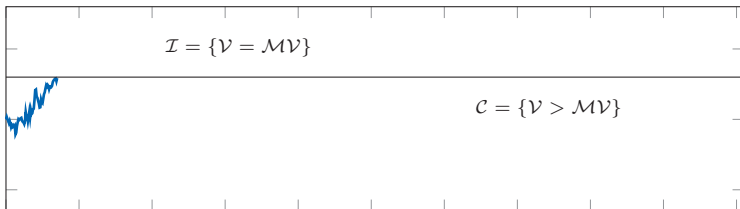
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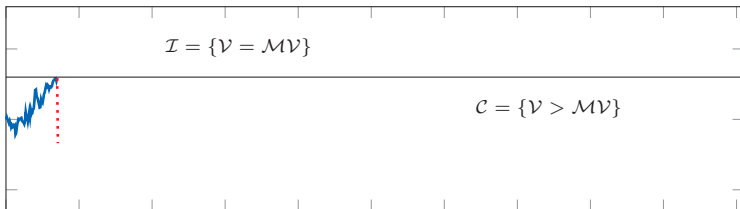
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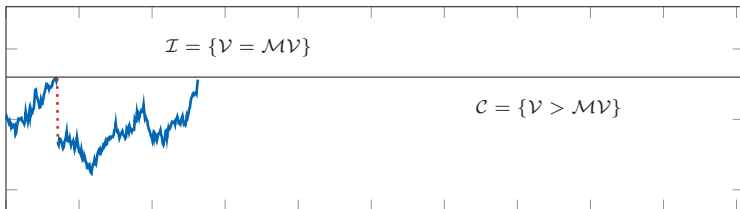
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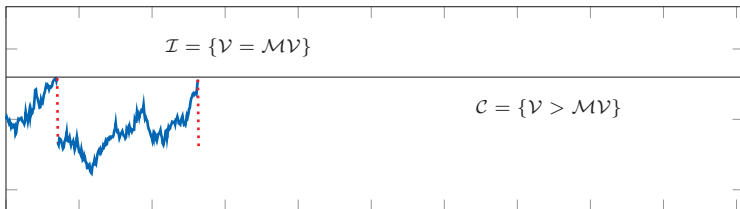
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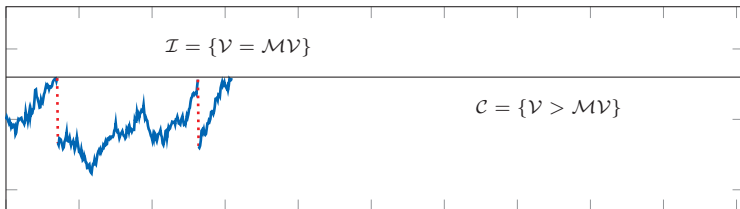
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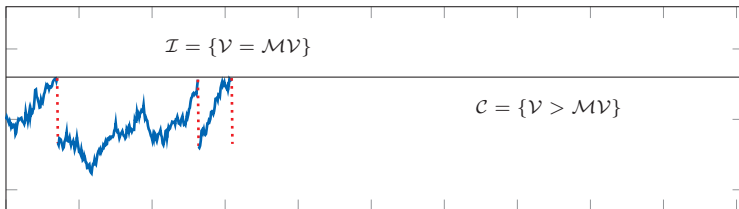
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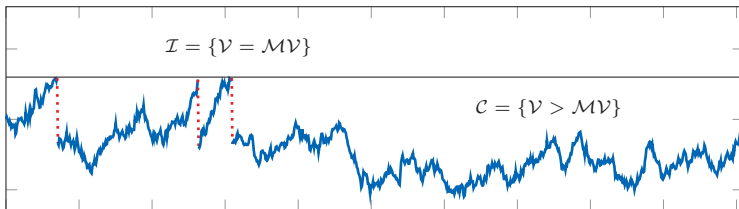
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- (i) Guess the set $\mathcal{C} \triangleq \{\mathcal{V} > \mathcal{M}\mathcal{V}\}$.
- (ii) Solve $\mathcal{L}\mathcal{V}(t, x) = 0$ on \mathcal{C} .



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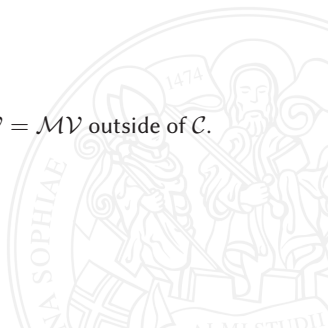
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- (i) Guess the set $\mathcal{C} \triangleq \{\mathcal{V} > \mathcal{M}\mathcal{V}\}$.
- (ii) Solve $\mathcal{L}\mathcal{V}(t, x) = 0$ on \mathcal{C} .
- (iii) Extend the solution to all of $[0, T] \times \mathbb{R}^n$ by setting $\mathcal{V} = \mathcal{M}\mathcal{V}$ outside of \mathcal{C} .



So, to solve the impulse control problem, we need to solve the QVIs

$$\begin{aligned} \min\{\mathcal{L}\mathcal{V}(t, x), \mathcal{V}(t, x) - \mathcal{M}\mathcal{V}(t, x)\} &= 0, & (t, x) &\in [0, T) \times \mathbb{R}^n, \\ \mathcal{V}(T, x) &= g(x), & x &\in \mathbb{R}^n. \end{aligned}$$

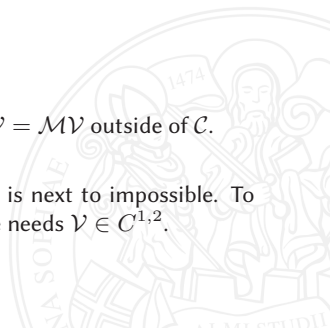
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Problem: This rarely works. In higher dimensions, this is next to impossible. To verify that the solution of the QVIs coincides with \mathcal{V} , one needs $\mathcal{V} \in C^{1,2}$.



Superharmonic Functions and Stochastic Perron



Superharmonic Functions

We say that an integrable function $h : [0, T] \times \mathbb{R}^n$ is **superharmonic** if

$$h(t, x) \geq \mathbb{E} \left[h(\tau, X_{t,x}(\tau)) \right]$$

for any choice of (t, x) and τ .



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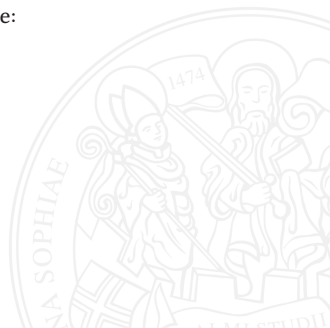
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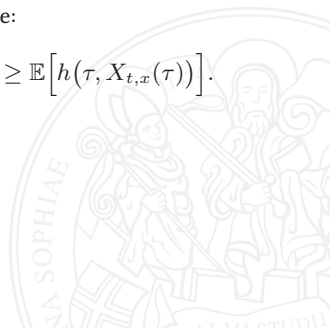
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That is h is **superharmonic**.



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Let \mathbb{H} be the set of all $h : [0, T] \times \mathbb{R}^n \rightarrow \mathbb{R}$ such that

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Theorem (B., Christensen, Seifried (2016))

Suppose that \mathcal{V} is continuous and $\mathcal{V} \in \mathbb{H}$. Under some technical assumptions, the control Λ^* defined in terms of

$$\{\mathcal{V} > \mathcal{M}\mathcal{V}\} \quad \text{and} \quad \{\mathcal{V} = \mathcal{M}\mathcal{V}\}$$

is **optimal** if it is admissible.

The Stochastic Perron Method

The stochastic Perron method:

- (1) Show that $\mathbb{V} \triangleq \inf_{h \in \mathbb{H}} h \geq \mathcal{V}$ is an upper semi-continuous (viscosity) subsolution of the QVIs:

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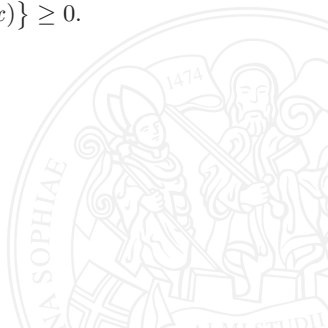
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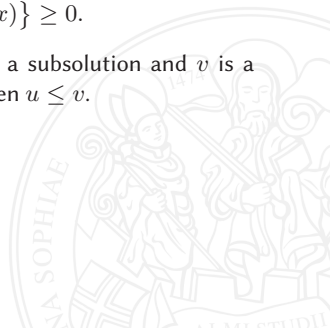
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$$\mathfrak{V} = \mathcal{V} = \mathbb{V} \quad \text{is continuous.}$$

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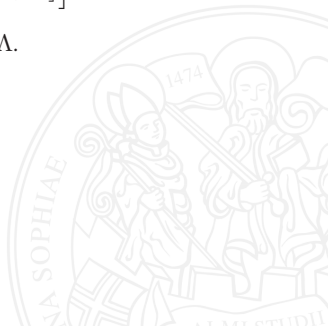
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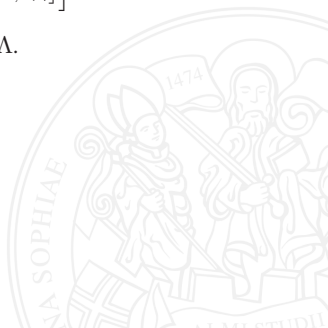
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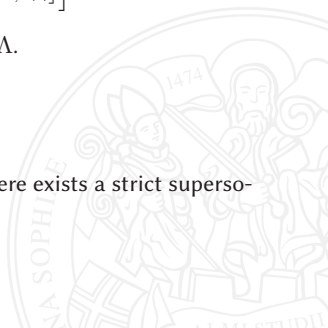
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Remark: The last two assumptions are e.g. satisfied if there exists a strict supersolution of the QVIs which grows faster at infinity than \mathcal{V} .



Applications and Work in Progress



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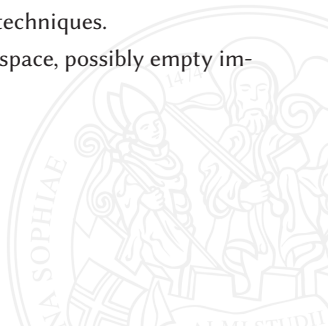
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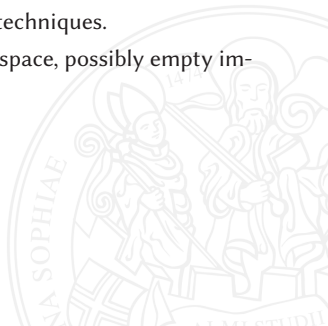
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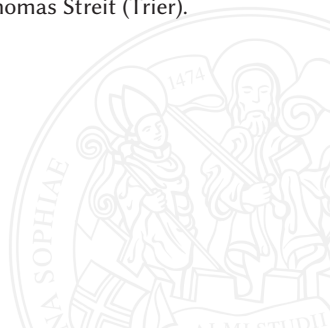
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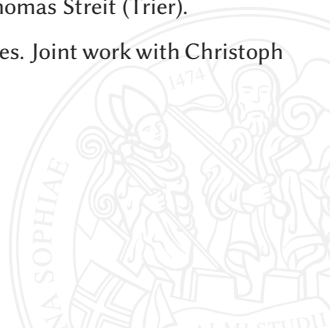
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Thanks for your attention!

Belak, Christensen, Seifried (2016):

A General Verification Result for Stochastic Impulse Control Problems

To appear in *SIAM Journal on Control and Optimization*

Belak, Christensen (2016):

Utility Maximization in a Factor Model with Constant and Proportional Costs

Available at: www.belak.ch/publications/

