

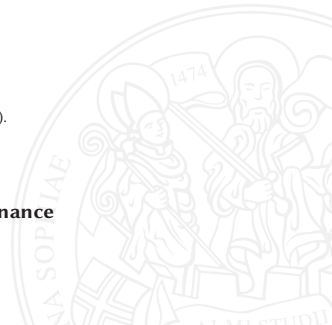
Utility Maximization with Constant and Proportional Costs

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Joint work with **Sören Christensen** (Göteborg).

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September 14, 2016



Market Model and Problem Formulation



The Market Model and Transaction Costs

The **financial market** consists of n assets $P = (P^1, P^2, \dots, P^n)$ with prices

$$\begin{aligned}dP(t) &= \text{diag}(P(t)) [\mu(Y(t))dt + \sigma(Y(t))dW(t)], \\dY(t) &= \alpha(Y(t))dt + \beta(Y(t))dW(t).\end{aligned}$$



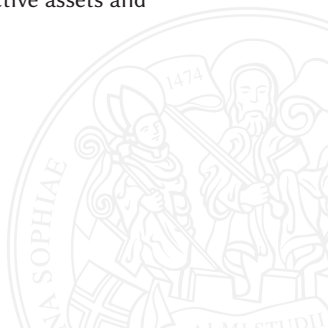
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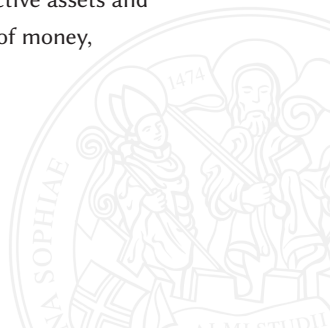
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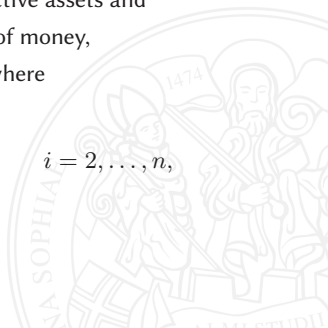
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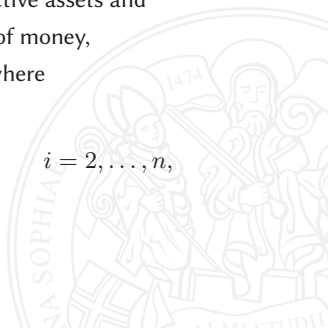
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where $\gamma_i \in (0, 1)$ and $K_i > 0$ for $i = 2, \dots, n$.



Solvency and Trading Strategies

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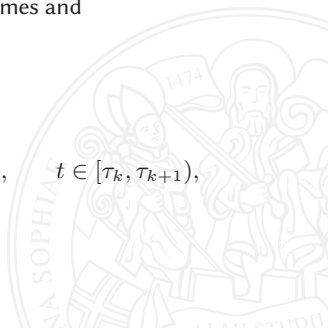
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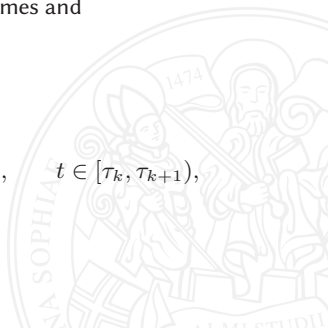
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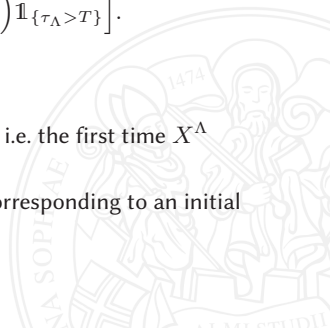
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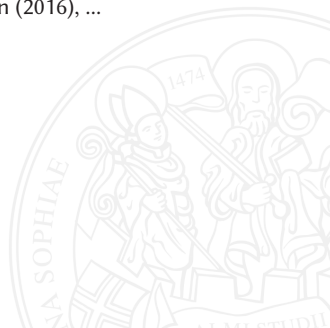
In the above,

- τ_Λ denotes the **bankruptcy time** of the strategy Λ , i.e. the first time X^Λ leaves the solvency region \mathcal{S} , and
- $\mathcal{A}(t, x, y)$ denotes the set of admissible strategies corresponding to an initial state of (t, x, y) .



In the existing literature, there are three different approaches:

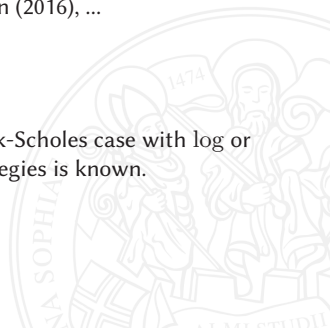
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Note: To the best of our knowledge, not even in the Black-Scholes case with log or power utility, a rigorous existence result for optimal strategies is known.



The Candidate Optimal Strategy



The Quasi-Variational Inequalities

The Bellman principle suggests that \mathcal{V} can be linked to the **quasi-variational inequalities** (QVIs)

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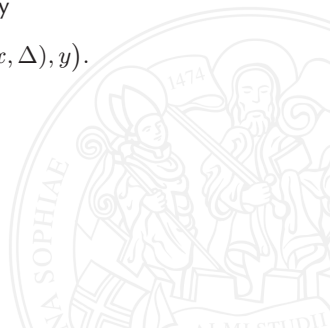
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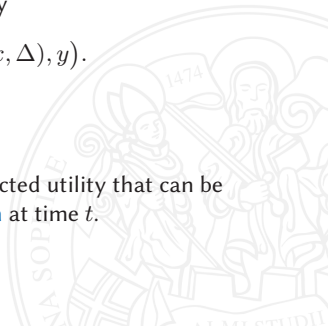
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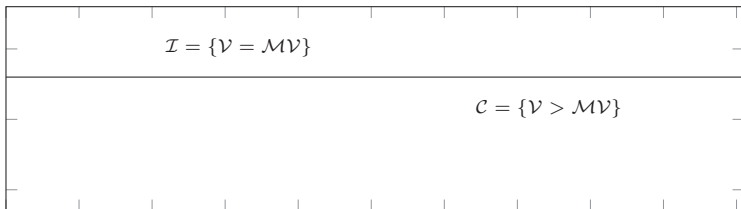
Note: $\mathcal{M}\mathcal{V}(t, x, y)$ can be thought of as the highest expected utility that can be achieved if the investor is **forced to make a transaction** at time t .



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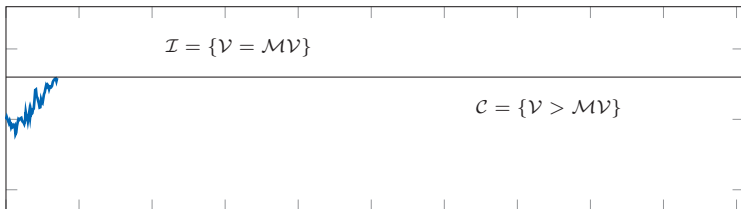
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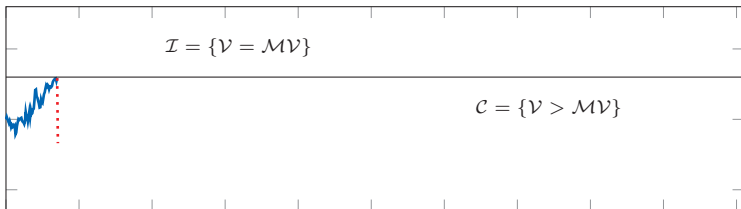
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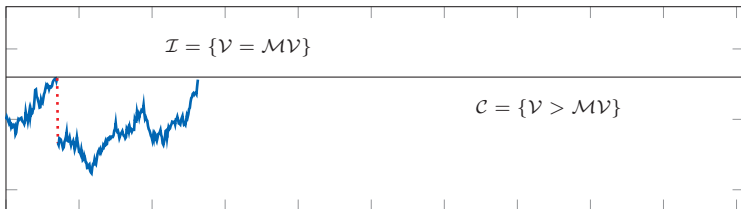
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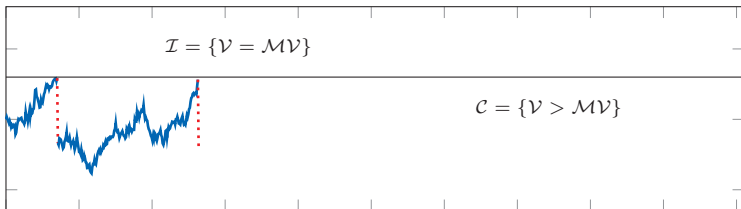
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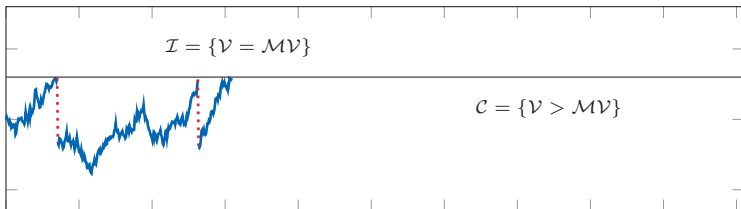
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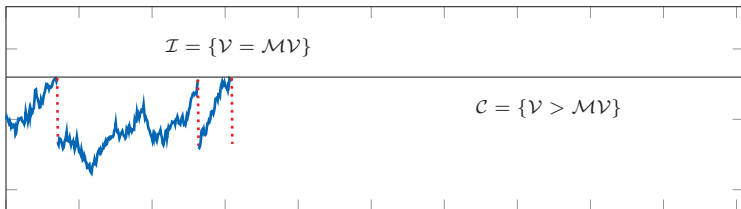
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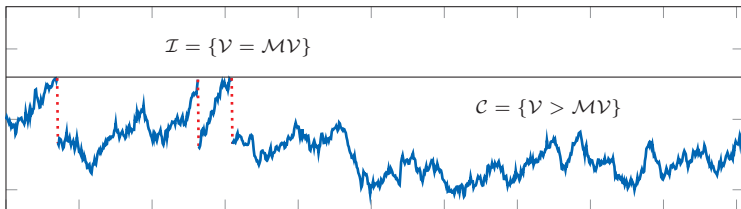
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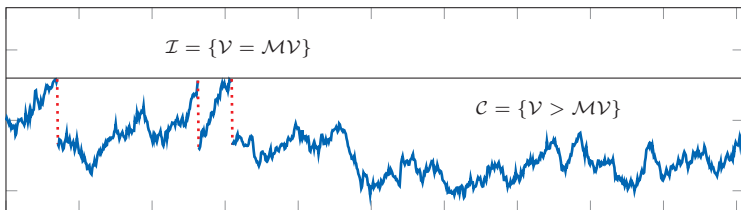
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Problem: To verify optimality we need sufficient regularity to apply Itô's formula, but it seems unlikely that the QVIs admit a $C^{1,2}$ -solution.

Verification via Superharmonic Functions



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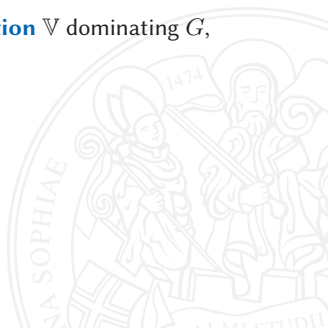
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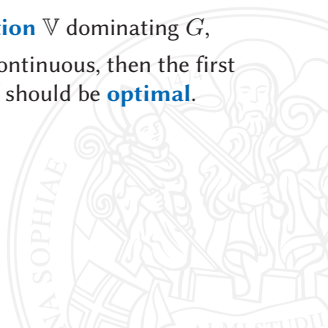
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Remark: $G = \mathcal{M}\mathcal{V}$ is upper semi-continuous if $\mathcal{V} = \mathbb{V}$ is upper semi-continuous. That is, to verify optimality, continuity of \mathbb{V} should suffice!

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Verification Theorem (Belak and Christensen (2016))

Suppose that \mathbb{V} is continuous. Then $\mathbb{V} = \mathcal{V}$ and the candidate optimal control Λ^* defined in terms of the sets

$$\mathcal{C} = \{\mathbb{V} > \mathcal{M}\mathbb{V}\} \quad \text{and} \quad \mathcal{I} = \{\mathbb{V} = \mathcal{M}\mathbb{V}\}.$$

is admissible and optimal.

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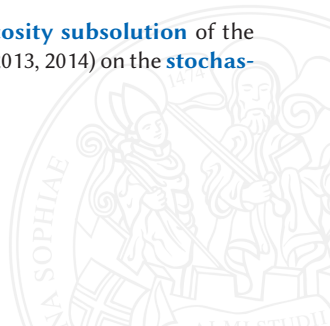
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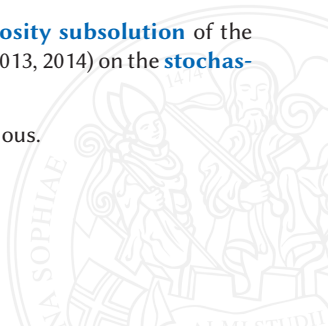
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- (4) By **viscosity comparison**: $\mathbb{V} \leq \mathbb{V}_*$, i.e. \mathbb{V} is continuous.



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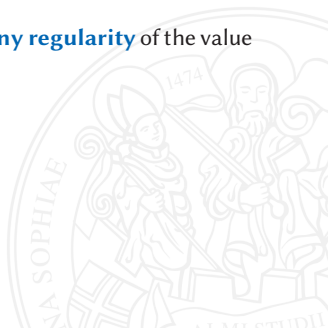
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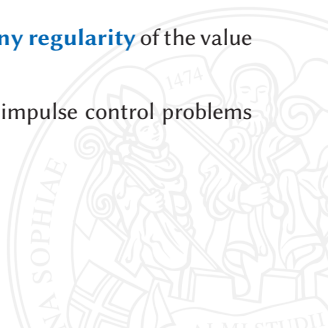
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- The approach **generalizes** to a big class of general impulse control problems (see Belak, Christensen, and Seifried (2016)).



Thanks for your attention!

Belak and Christensen (2016):

Utility Maximization in a Factor Model with Constant and Proportional Costs

Belak, Christensen, and Seifried (2016):

A General Verification Result for Stochastic Impulse Control Problems

Available at: www.belak.ch/publications/

