

The Geometry of Model Uncertainty

Mathias Beiglböck

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1) background – model uncertainty and optimal transport

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- 2) particular aspect: Skorokhod embedding

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basic object: **stock price** (random process)

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all call prices known $\iff S_t \sim_{\mathbb{P}} \mu_t$

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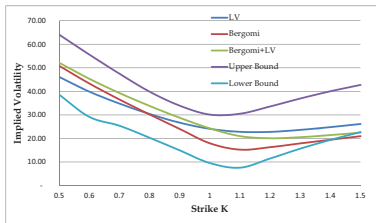
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lower/upper prices versus (local) Bergomi and LV models

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B., Henry-Labordere, Penkner / Galichon, Touzi ('13)

→ *transport approach*

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call prices known



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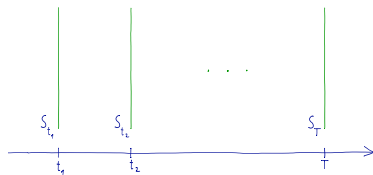
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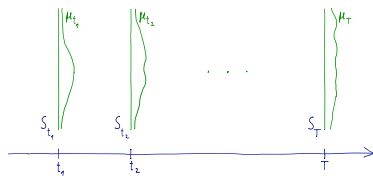
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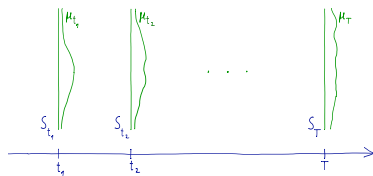
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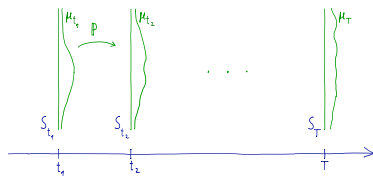
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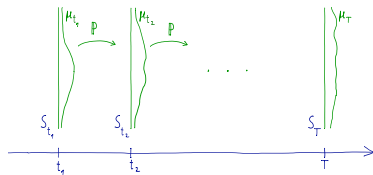
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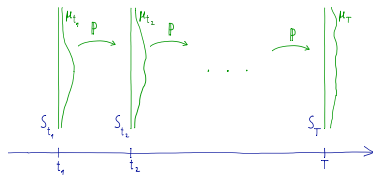
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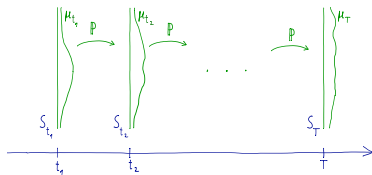
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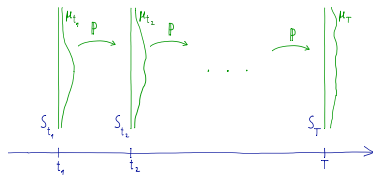
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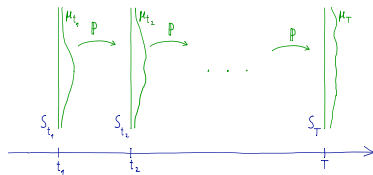
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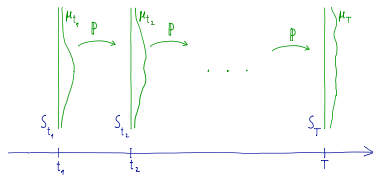
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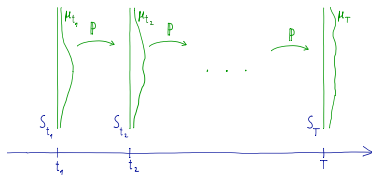
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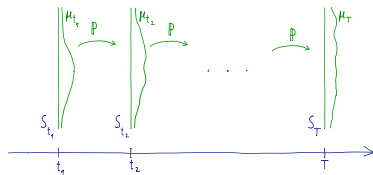
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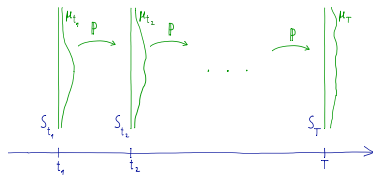
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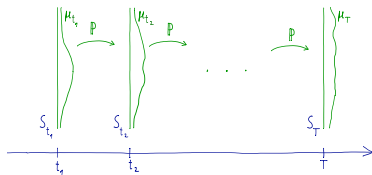
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geometric description of extreme models

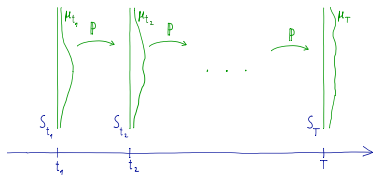
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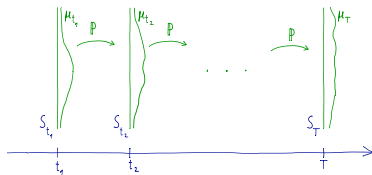
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cont. time – *geometry* of Skorokhod embedding

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- Rost '71
- Azema-Yor '79
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- | | | |
|-----------------|------------------|-----------------|
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| ○ Rost '71 | ○ Hobson '98 | <i>surveys:</i> |
| ○ Azema-Yor '79 | ○ Madan-Yor '02 | ○ Obloj '04 |
| ○ Bass '83 | ○ Cox-Hobson '07 | ○ Hobson '11 |
| ○ Vallois I '83 | ○ Eldan '15 | |
| ○ Perkins '85 | ○ ... | |

Skorokhod embedding problem (SEP)

given: μ , $\int x^2 d\mu < \infty$, $B \dots$ BM with $B_0 = \int x d\mu$

task: determine τ s.t. $B_\tau \sim \mu$, $\mathbb{E}\tau < \infty$

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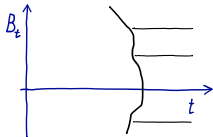
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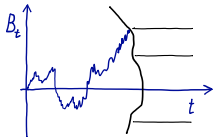


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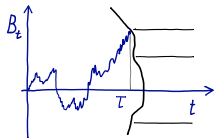


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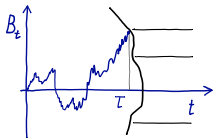


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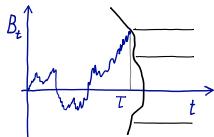


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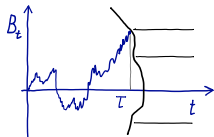
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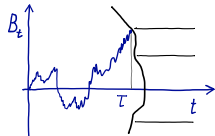
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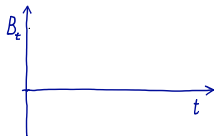
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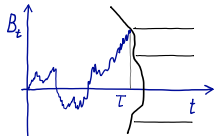


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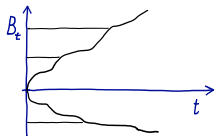
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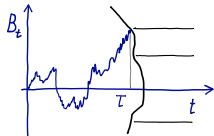


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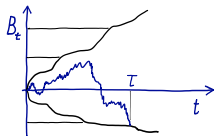
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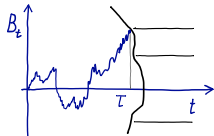


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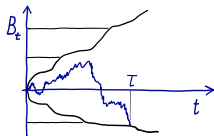
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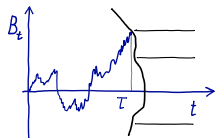
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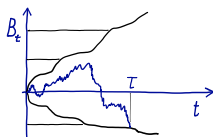
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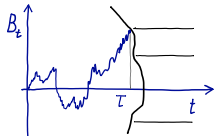
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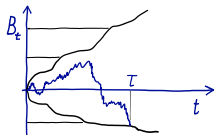
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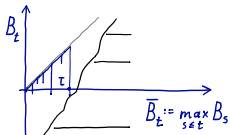
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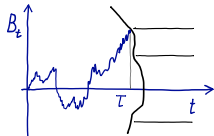


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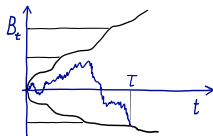
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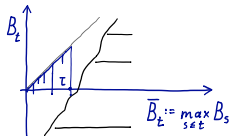
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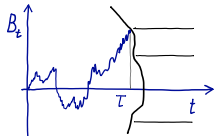
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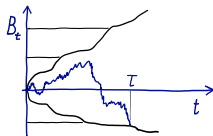
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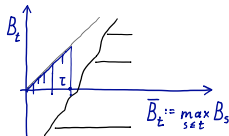
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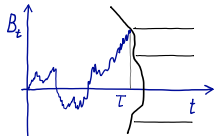
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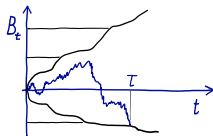
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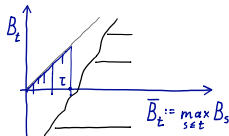
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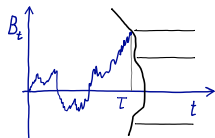
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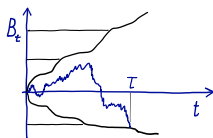
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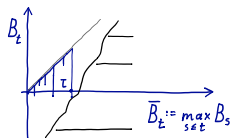
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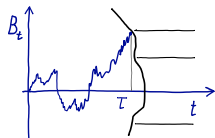
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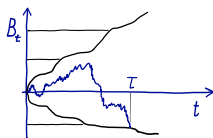
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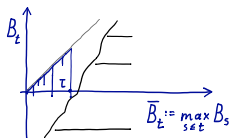
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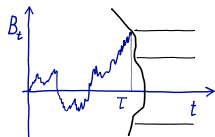
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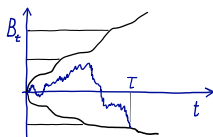
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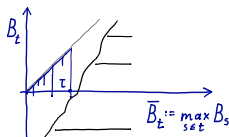
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optimal embeddings: basis for *all* known extremal models

Transport Approach to Skorokhod Embedding [BCH15]

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Φ Borel, τ optimal \implies support of τ is pathwise optimal
 \exists monotone Γ s.t. $\mathbb{P}((B_s)_{s \leq \tau} \in \Gamma) = 1$

Γ monotone $:\iff \Gamma$ cannot be improved by pathwise modifications


Thm [Root69]:

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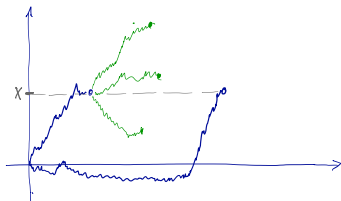
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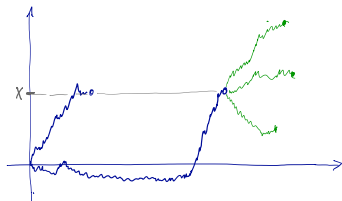
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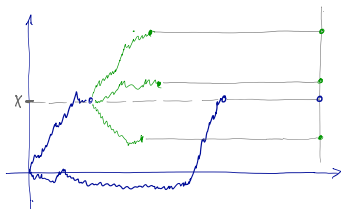


VS

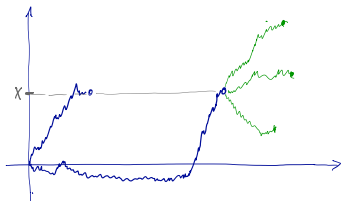



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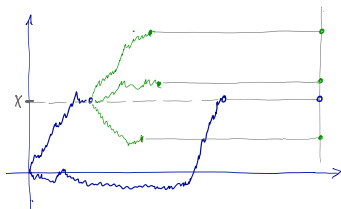


VS

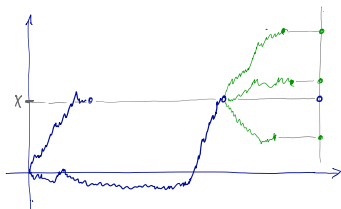



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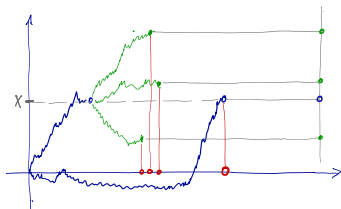


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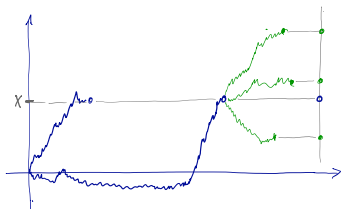



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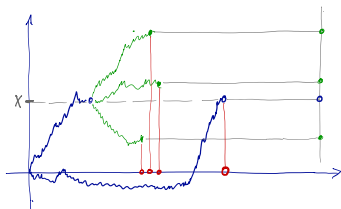


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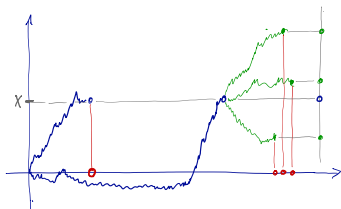



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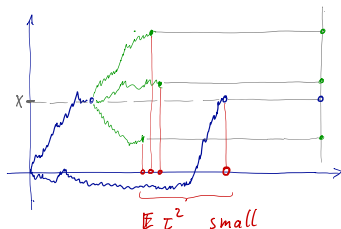


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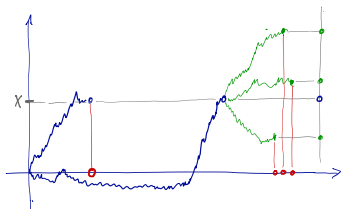



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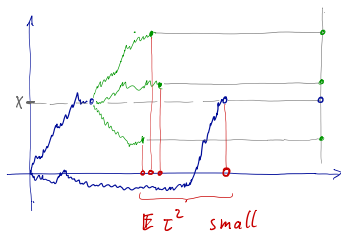


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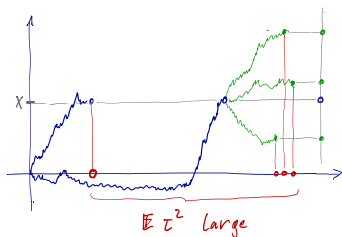



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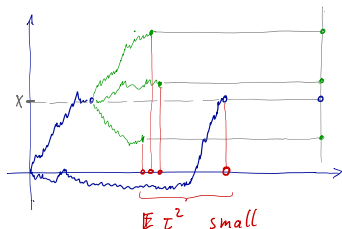


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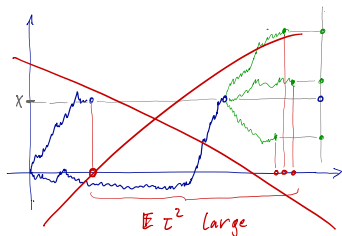



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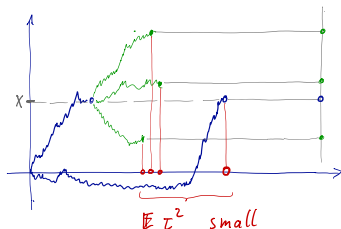


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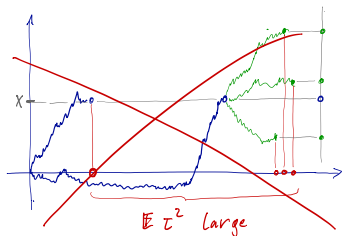


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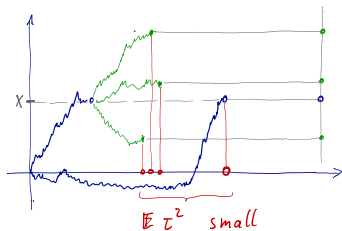
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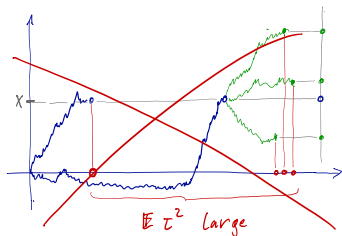
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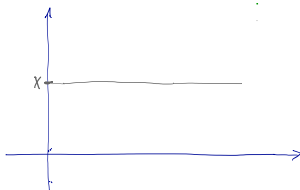
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


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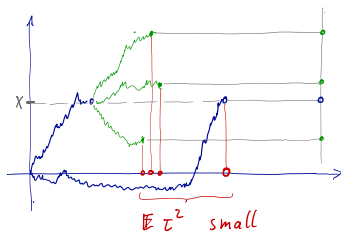


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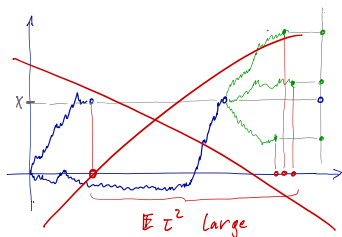


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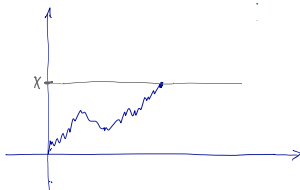
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


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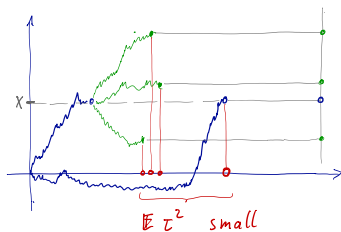


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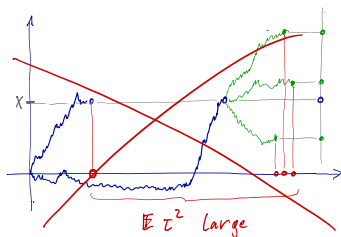


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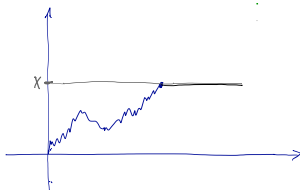
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


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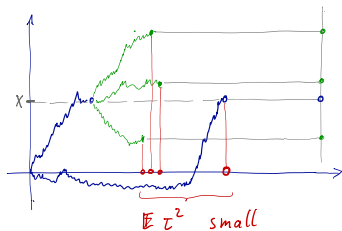


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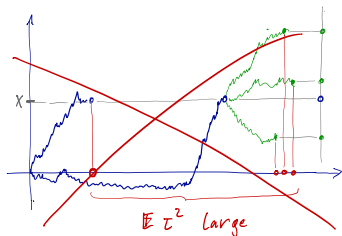


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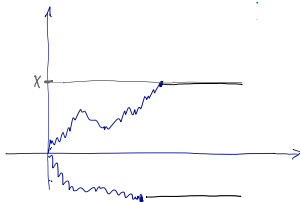
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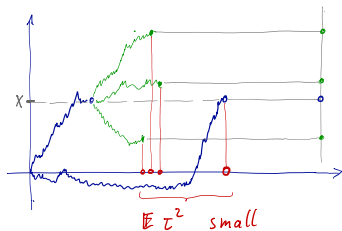


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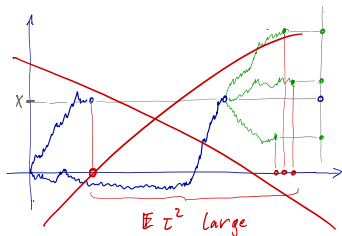


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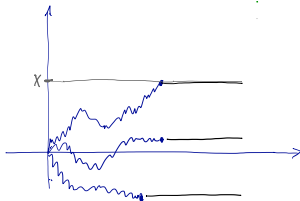
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


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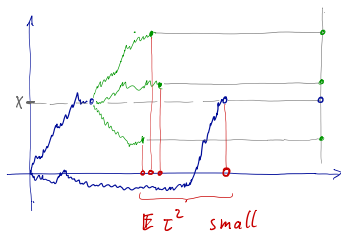


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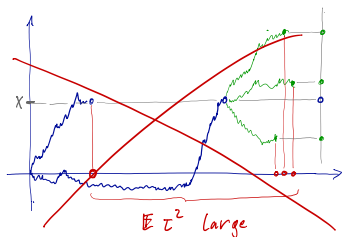


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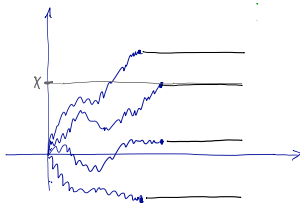
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


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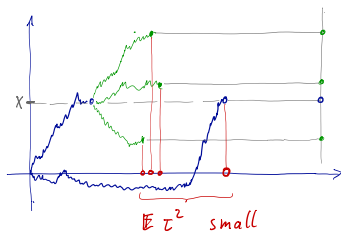


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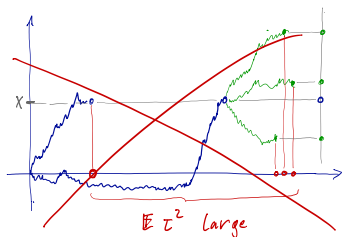


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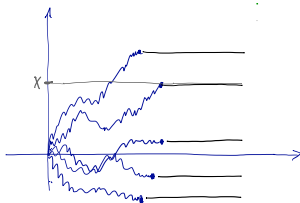
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


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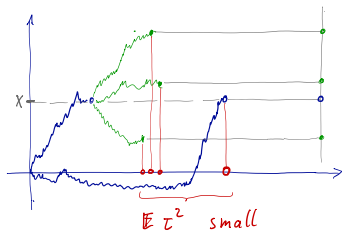


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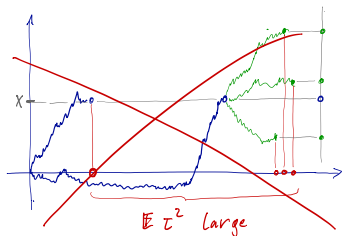


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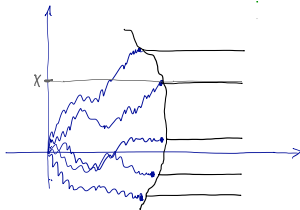
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


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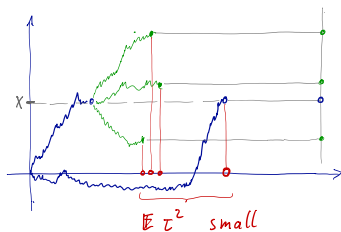


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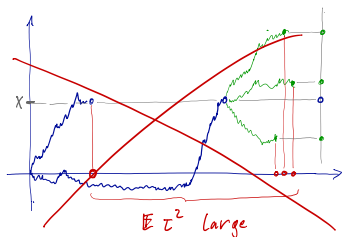


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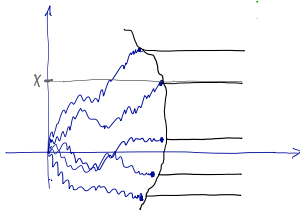
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transport approach [BCH16]:

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transport approach [BCH16]: all optimal emb. extend in full generality

Model Uncertainty and Skorokhod Embedding:

- Hobson, Robust hedging of the lookback option, *Finance and Stochastics* '98
- Obloj, The Skorokhod embedding problem and its offspring, *Probab. Surv.*, '04
- Hobson, The Skorokhod embedding problem and model-independent bounds for option prices, *Springer, Berlin*, '11

Connection to (Martingale) Optimal Transport:

- B/Henry-Labordere/Penkner, Model-independent Bounds for Option Prices: A Mass Transport Approach, *Finance and Stochastics* '13
- Galichon/Henry-Laborder/Touzi, A stochastic control approach to no-arbitrage bounds given marginals '14, *AAP*
- Dolinsky/Soner, Martingale optimal transport in continuous time, *PTRF* '14
- Acciaio/B/Penkner/Schachermayer, A model-free version of the FTAP, *Math. Fin.* '16

Monotonicity Principle:

- B/Julliet, Optimal transport under marginal martingale constraints, *Ann. Prob.* '16
- B/Nutz/Touzi, Complete Duality for MOT on the Line, *Ann. Prob.*, to appear
- B/Cox/Huesmann, Optimal Transport and Skorokhod embedding, *Inventiones Math.*, to appear