

# Optimal liquidation against a Markovian limit order book

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Objectives

Limit order book model

Optimisation

Numerical results

Outlook

# Modelling objectives

- ▶ Framework to investigate optimal multiperiod liquidation strategies against a limit order book

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- ▶ Detailed but tractable stochastic model of transaction costs
  - ▶ Spread
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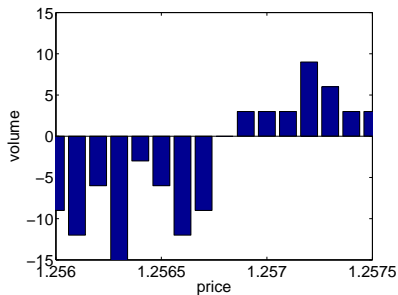
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- ▶ Allowance for risk
- ▶ Opportunity costs of delayed trading

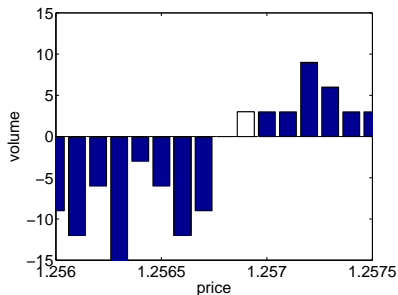
# Order book model

Following [SFGK03], we specify a Markov process  $S_t$  on the large state space of order books  $S$ . Need transition rates for...



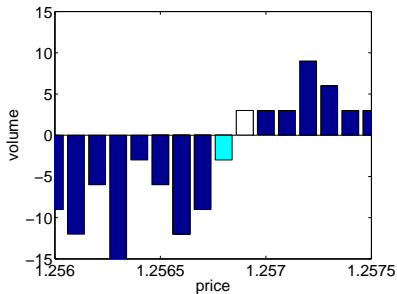
# Order book model

Market order arrival: e.g. fixed Poisson rate  $\mu$  for buys and sells



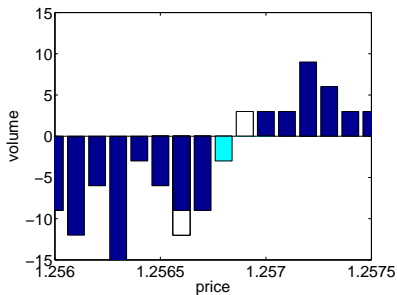
# Order book model

Limit order arrival: e.g. fixed Poisson rate  $\lambda$  per tick below best ask for buy orders, v.v. for sell orders



# Order book model

Cancellation of existing limit orders: e.g. outstanding limit order die at a rate  $\chi$



## Set-up of optimisation

- ▶ Objective (cf. [AC00]) is to sell  $V_0$  units of an asset to maximise risk-adjusted average sale price, using market orders only

$$\sup_{\tau_1 \leq \dots \leq \tau_{V_0} < T} \mathbb{E}(U(\sum_{i=1}^{V_0} p^{bid}(S_{\tau_i}))) \quad (1)$$

where  $S_t$  follows the above dynamics, except at times  $\tau_1, \dots, \tau_{V_0}$ , when a market buy order is removed;  $\tau_i$  adapted to full observation filtration generated by  $S_t$ .

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- ▶ Solution will map  $(s, v, c, T - t) \in \mathcal{S} \times \{1, \dots, V_0\} \times \mathbb{Z} \times \mathbb{R} \mapsto \{\text{sell 1 unit}, \dots, \text{sell } v \text{ units}, \text{wait}\}$

# Bellman equation

- ▶ Defining the value function  $\phi$  for the problem in the usual way, we have the following dynamic programming equation

$$\phi(s, v, c, T-t) = \max \begin{cases} \phi(TMS(s), v-1, c + p^{\text{bid}}(s), T-t) \\ \mathbb{E}(\phi(S_{t+\delta t}, v, c, T-(t+\delta t)) | S_t = s) \end{cases} \quad (2)$$

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- ▶ Direct solution difficult due to curse of dimensionality

## Reduction of the problem

- ▶ Put  $U(c) = -e^{-\gamma c}$ . Then the solution can be written in terms of the order book shape and not its location via  $\phi(s, v, c, T - t) = -e^{-\gamma c + v p^{\text{bid}}(s)} \bar{\phi}(\bar{s}, v, T - t)$ .

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- ▶ Note that  $\phi(\cdot, v, \cdot, \cdot)$  depends explicitly on  $\phi(\cdot, v - 1, \cdot, \cdot)$ , so we can solve sequentially for increasing  $v$ .
- ▶ For each  $v$ , it remains to find optimal strategy mapping  $\bar{s} \in \bar{\mathcal{S}} \mapsto \{\text{sell, wait}\}$

## Numerical schemes

- ▶ Design low-dimensional approximation architecture for  $\bar{\phi}$

$$\bar{\phi}(\bar{s}) \cong \psi(\beta, \bar{s}) \quad (4)$$

using e.g. state aggregation, radial basis functions etc.  
Involves defining a distance on  $\bar{\mathcal{S}}$  which captures relevant similarity of order book shapes.

## Numerical schemes

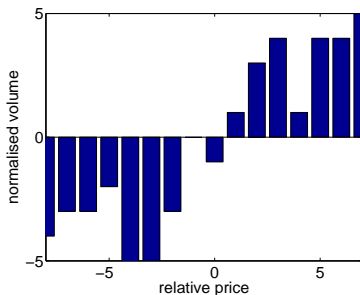
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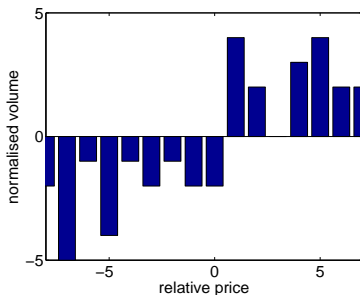
- ▶ Tune  $\beta$  so that  $\psi$  approximately satisfies Bellman equation over  $\bar{\mathcal{S}}$ , using e.g. approximate value iteration, Q-learning ([TR96], [Ber05])

## Example results - preliminary



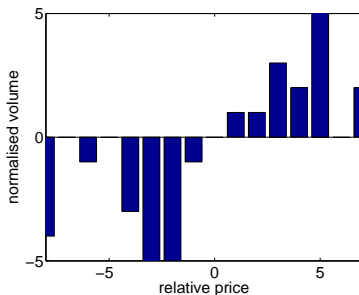
In sell region when  $\nu = 1$  but not when  $\nu = 2$

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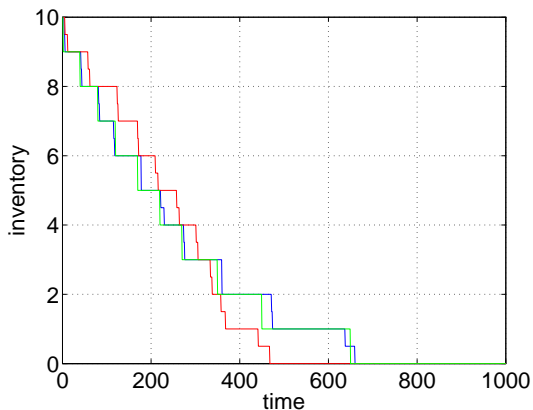
In sell region when  $v = 2$  (in fact sell 2 units) but not when  $v = 1$

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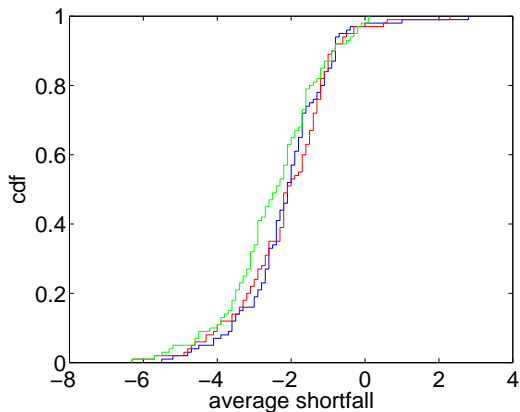


Not in sell region for  $v = 1$  or  $v = 2$

## Example results - preliminary







## Example results - preliminary



# Outlook

- ▶ Improvement and analysis of numerical schemes
- ▶ Upper bounds for value function using e.g. duality
- ▶ Allow our agent to use limit sell orders
- ▶ Expand state space to capture salient non-Markovian features of limit order book process
- ▶ Partial observation of limit order book state, with application to FX market

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