Simulating Financial Markets using MASON Framework*

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Abstract. AAA

Key words: Agent-based Modeling, Computational Social Science, Financial Markets

- 1 Motivation and Objectives
- 2 Platform Architecture
- 3 Overview of Implemented Models

3.1 Cont Model

The Cont model Cont [2006] is the simplest implemented. All traders (agents) follow the same behavioral rules. They are heterogeneous in the sense that they are given independently assigned (subjective) volatility thresholds $\theta_i(t)$. Each period, all agents recieve a common signal, which can be interpreted as public information or "news", in the form of IID Gaussian random variables ϵ_t . Each agent i responds to this signal by selling if $\epsilon_t < -\theta_i(t)$, buying if $\epsilon_t > \theta_i(t)$, and otherwise sitting out the period. The market then determines the excess demand and arrives at the market clearing price by means of a market impact function. Lastly, with probability s, agents update their threshold to match the absolute value of the return rate for the current period. Somewhat surprisingly, even this very simple "zero intelligence" model of agent behavior yields fairly realistic market movements, as presented on Figure 3.

TODO Note:cyclicality.

3.2 Farmer Model

The Farmer model Doyne Farmer et al. [2003] has two types of agents, which trade a single asset by means of a continuous double auction similar to those normally used on stock markets. "Patient" traders place orders place limit orders,

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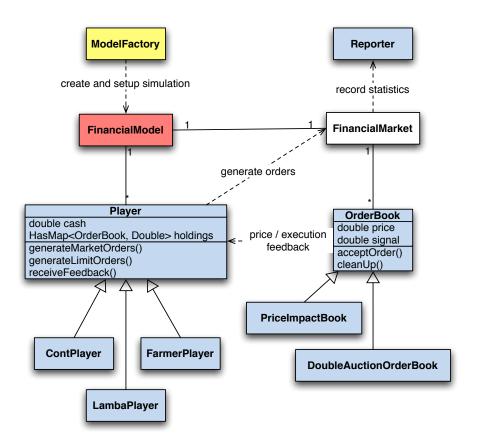


Fig. 1. High-level UML class diagram of the main components and relations in the FinancialMarketModel, including the main attributes of Players and Order-Books. Agent classes (light blue) inherit from the MASON Steppable interface while the master class is implementation of MASON's SimState.

which specify both a price and quantity they wish to buy or sell. Limit orders do not execute immediately, and may be canceled at any time. "Impatient" traders place market orders, specifying only a quantity, which execute immediately at the current best price, assuming there is sufficient liquidity provided by outstanding limit orders. This model also has simple behavioral logic for traders. Behavior is regulated by:

– Patient agents place limit orders of constant size at a Poisson rate per unit time, and with a randomly generated price. Orders are equally likely to be either buy or sell orders. Buy limit orders are placed uniformly anywhere in the semiinfinite interval below the current ask price $(-\infty , where <math>p$ is the logarithm of the price; similarly for sell orders. A uniform distribution

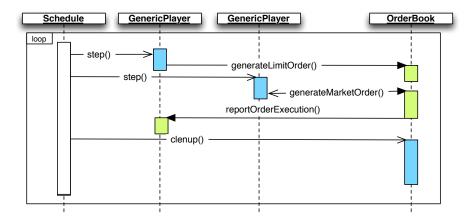


Fig. 2. High-level UML sequence diagram of the interactions between main object of the FinancialMarketModel.

in log prices is equivalent to an exponential distribution in prices. In addition, outstanding limit orders are cancelled at a Poisson rate per unit time. All of these processes are independent.

- Impatient agents place market orders of constant size at a Poisson rate per unit time.
- A double auction order book. This manages the pending limit orders and facilitates trades between market and limit orders. It also provides aggregate market information such as return rate per unit time and bid-ask spreads.

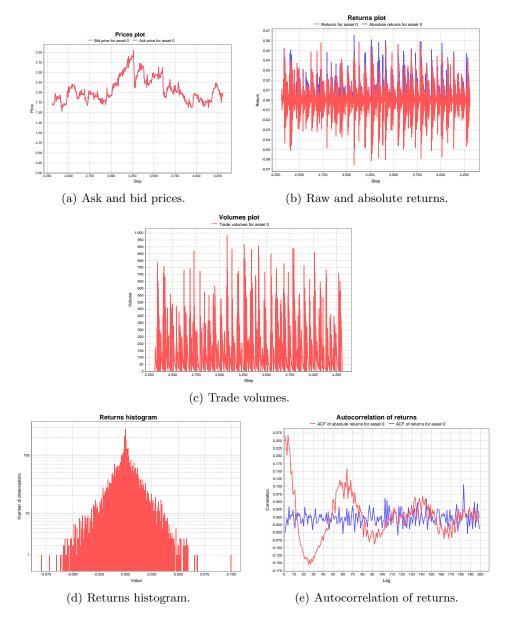
This model explains a large portion of the spread and price diffusion of actual stocks given an order flow rate, which essentially means that the double auction structure has a large impact on the nature of market movements. The model also yields market behavior qualitatively close to actual markets in many respects, see Figure 4.

TODO Note: log vs. non-log prices TODO Note: effect of price granularity

3.3 Farmer-Cont Model

We constructed a model combining aspects of the Farmer model with aspects of the Cont model. The result is a model which is still essentially zero intelligence, but which exhibits empirically positive characteristics of both models. In this model there are two types of agents, which interact via the continuous double auction mechanism, as in Farmer. In fact, the Patient traders, which place limit orders and thus provide liquidity, remain the same. Impatient traders are modeled after Cont. They have heterogeneous volatility thresholds, and place market orders depending upon how public information in the form of a normally distributed random variable compares with these thresholds. TODO Show results. TODO Note: liquidity issues

4 R. Axtell et al.



 ${\bf Fig.\,3.}$ Examples of outputs and statistics from a single run of the Financial Model simulation for default Cont's parametrization.

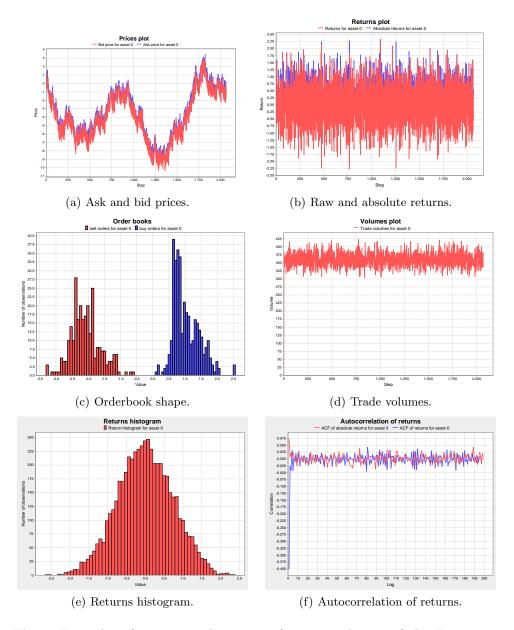


Fig. 4. Examples of outputs and statistics from a single run of the FinancialModel simulation for default Farmer's parametrization.

6 R. Axtell et al.

4 Verification of Correctness

TODO: show results side by side with original papers' figures

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