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Simulation of Trading in an artificial Stock Market

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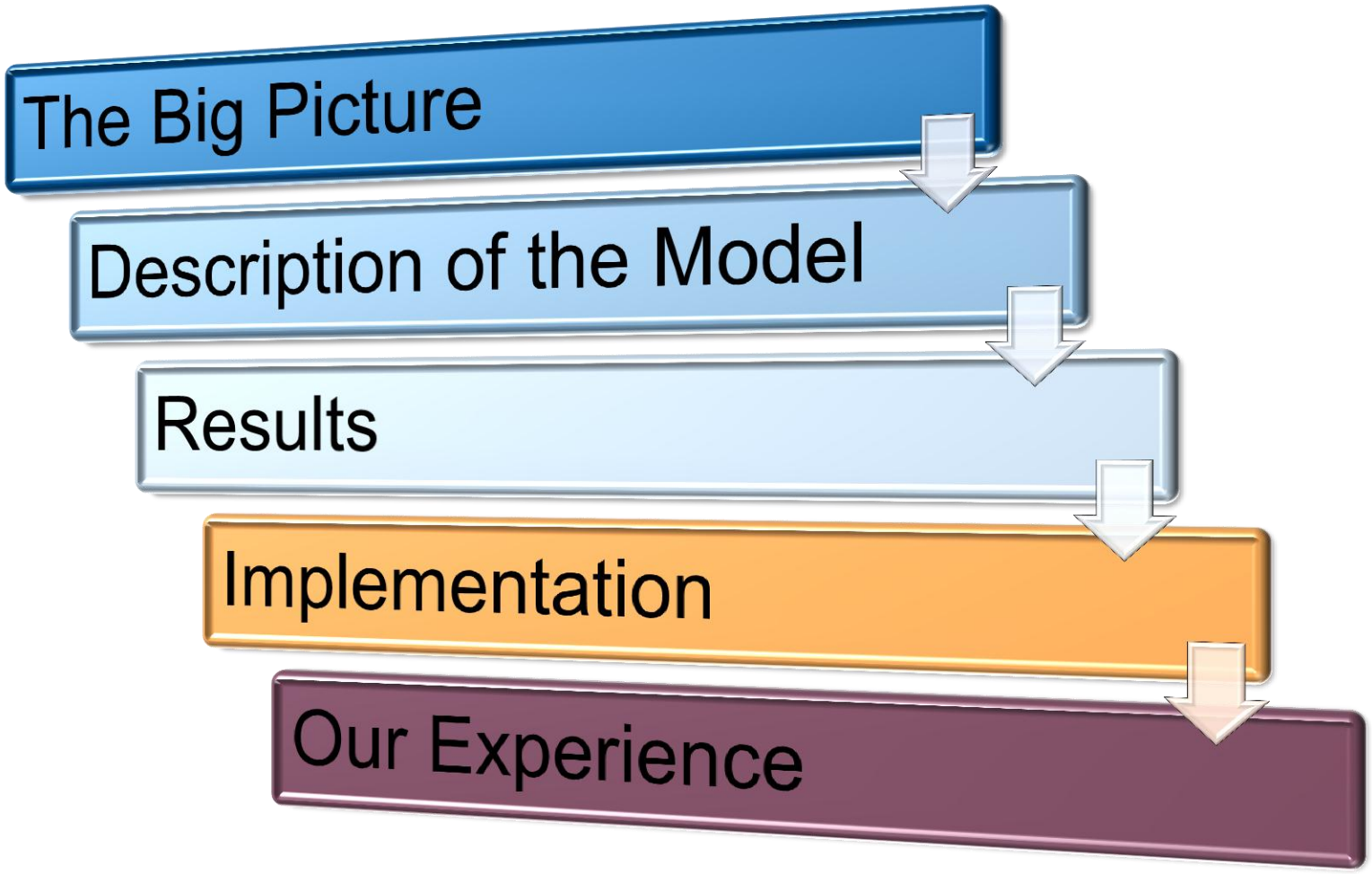
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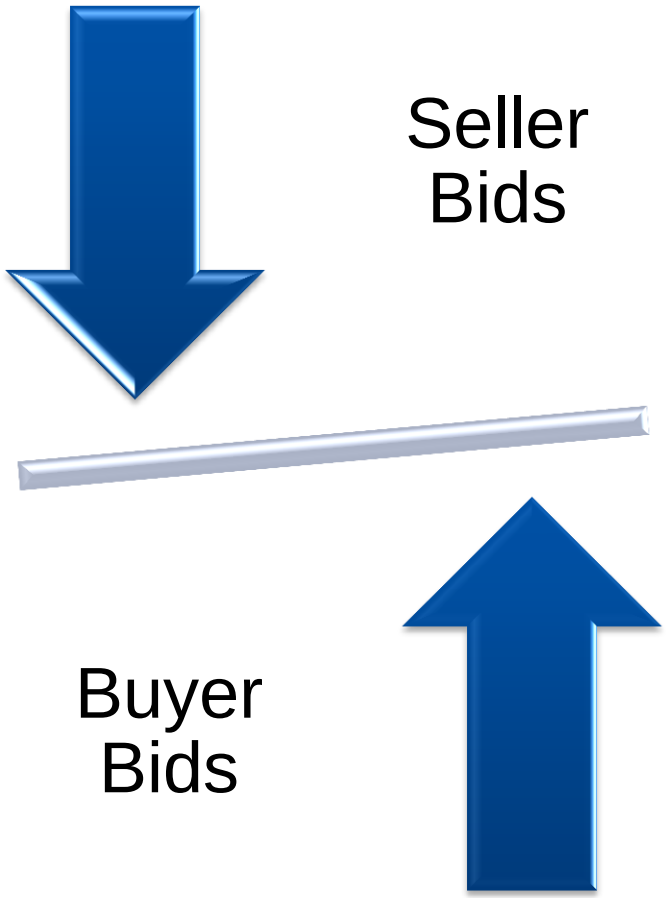
The Big Picture – Why do we want a Model

- The financial system is flawed
- A financial crisis has very broad repercussions on the general prosperity of the population
- The stock market can be used as a general indicator of the state of an economy
- The goal : to foresee/prevent financially unfavorable situations by modelling trading on the stock market

Fundamental Research Questions

The Model (1)

- One publically traded firm
- Based on paper (Raberto, 2005)
- Agent-based simulation : Traders (agents) interact in the market (playground)
- Fixed number of traders each of which have a specific amount of shares and liquidities which evolve over time
- Limit order book (double auction) mechanism : sellers and buyers bid seperately; price overlap of bids causes a transaction

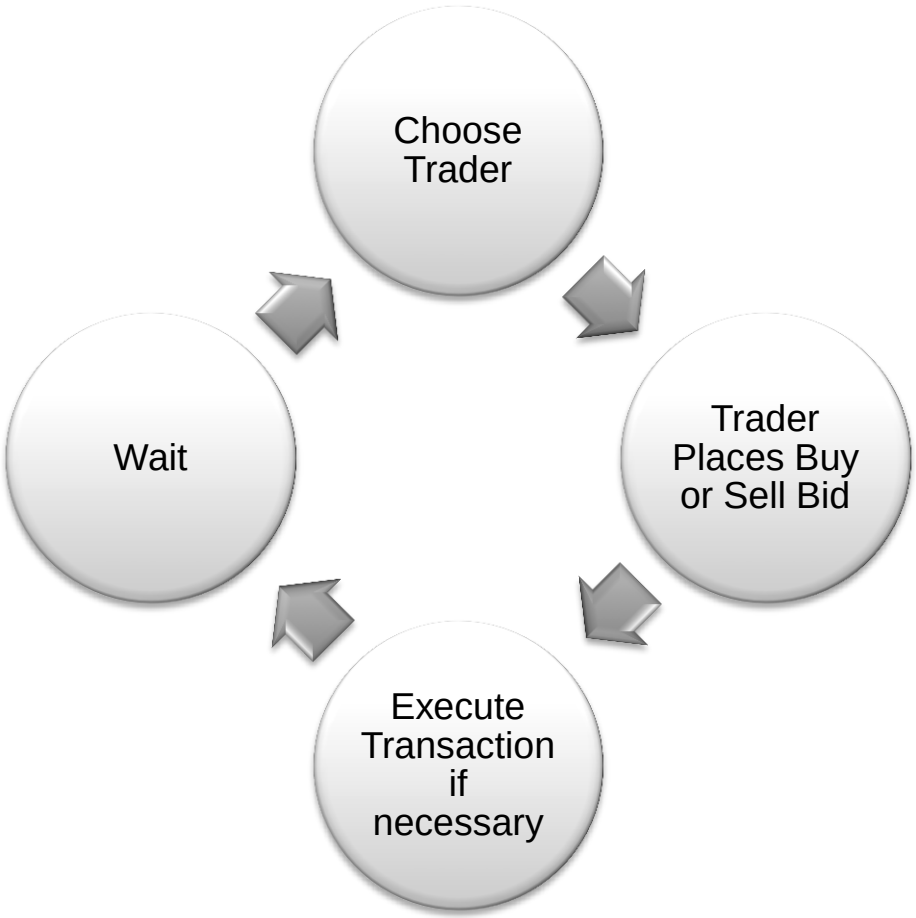


The Model (2)

- Time is separated into days
- At the end of a day the market closes : all remaining bids are cleared from the books
- Input : initial bid prices (IPO parameters), number of traders, distribution parameters for random variables, etc...
- Output : Average transaction price over time, trader assets over time, value of firm over time, etc...

Source: ece.cmu.edu

Sequence of Events – When the market is open



- The waiting time follows an exponential distribution
- A trader is chosen based on a uniform distribution
- The trader will place either a buy or sell order with a predetermined probability
- The price of the order follows a Gaussian distribution with the price of the most competitive valid order as the mean
- The number of shares ordered is uniformly distributed between 1 and the trader's maximum

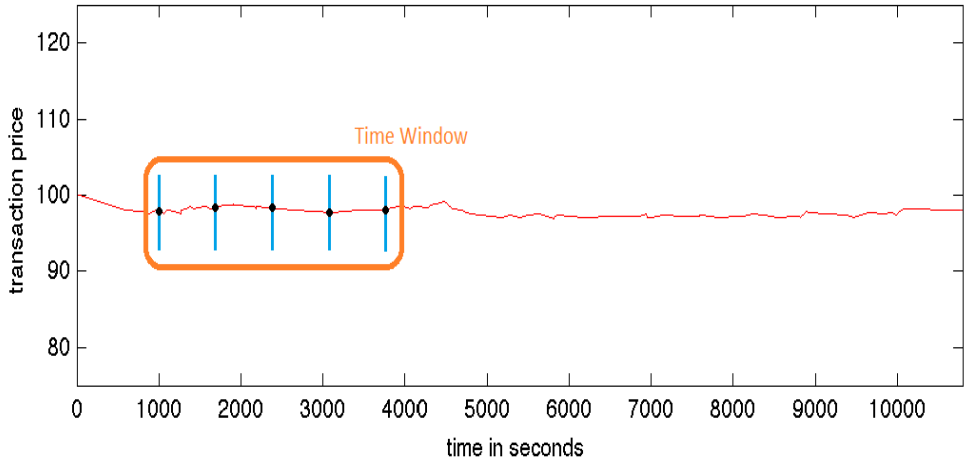
Measuring Market Volatility

- Rate of Return (ROR) : ratio of money gained or lost on an investment relative to the amount of money invested (%)
- Logarithmic returns :

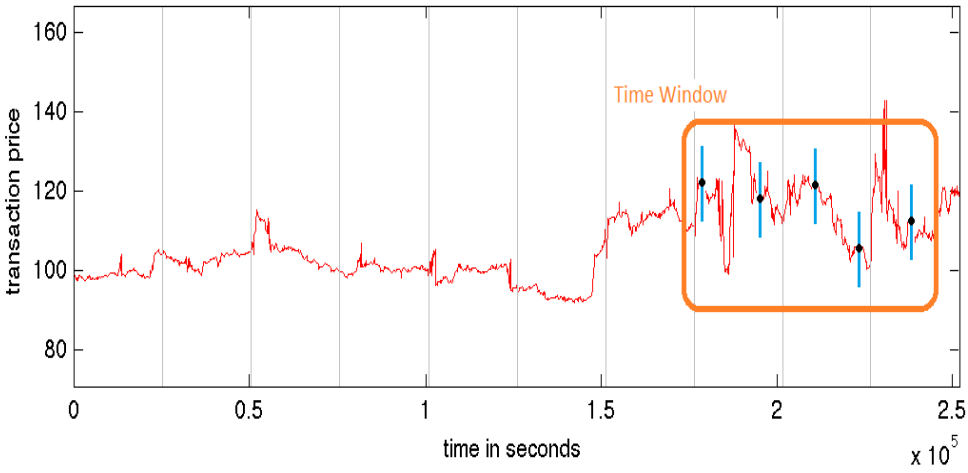
$$r_{log} = 100 \cdot \ln\left[\frac{V_{final}}{V_{initial}}\right]$$

- Single-period vs. Multiperiod
- Take standard deviation of multiple single-period return values (σ_T)
- Stable market : small deviations
- Volatile market : large deviations

Stable Market :



Volatile Market :



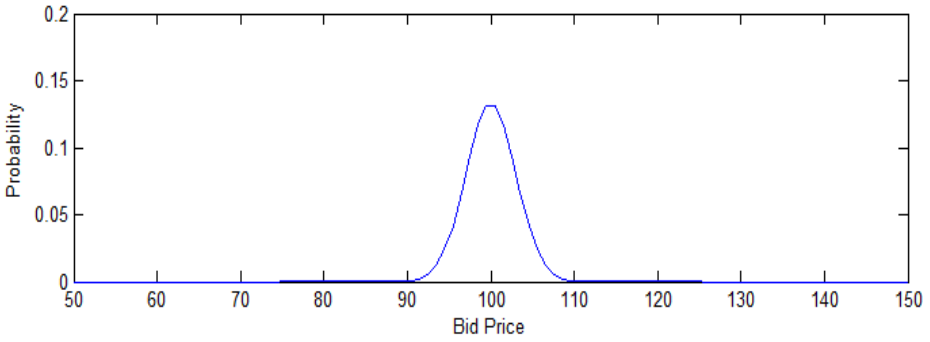
Taking past Market Volatility into Account

- Volatility feedback occurs solely in the prices of new bids
- The principle : if the transaction price has fluctuated in the past traders are less sure of the true value of the stock
- As a result, prices tend to deviate more from the latest bid values
- We can achieve this by modifying the variance of the Gaussian distribution that the new bid prices are based on :

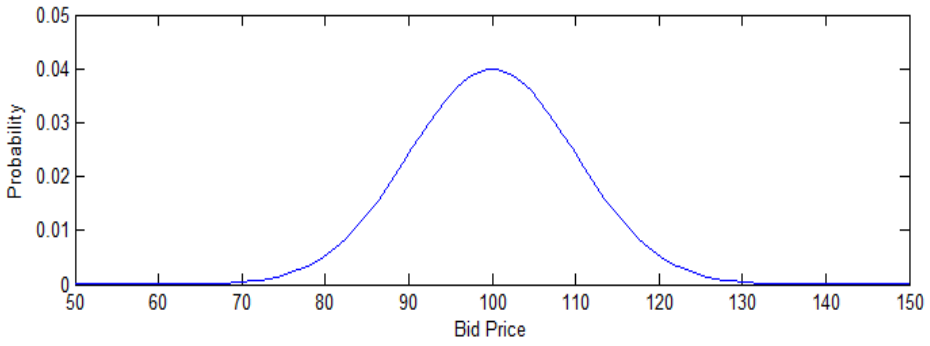
$$\sigma_{new} = k \cdot \sigma_T \, , k = const.$$

$$\sigma = \alpha \cdot \sigma_{new} + (1 - \alpha) \cdot \sigma \, , \alpha \in \,]0,1]$$

Low Volatility (Small Variance) :

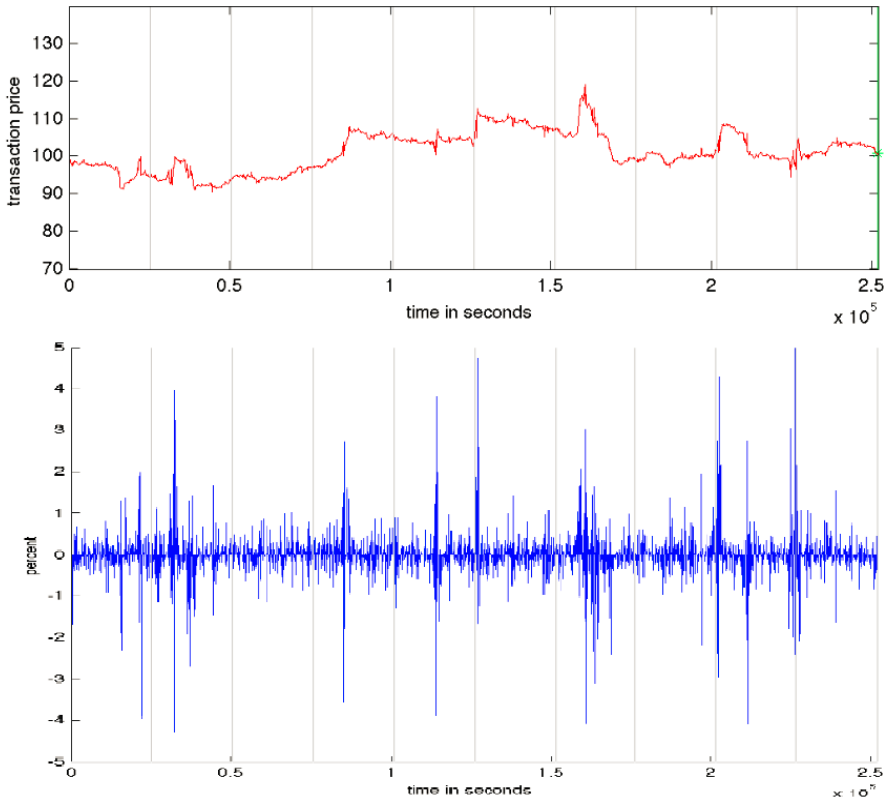


High Volatility (Large Variance) :

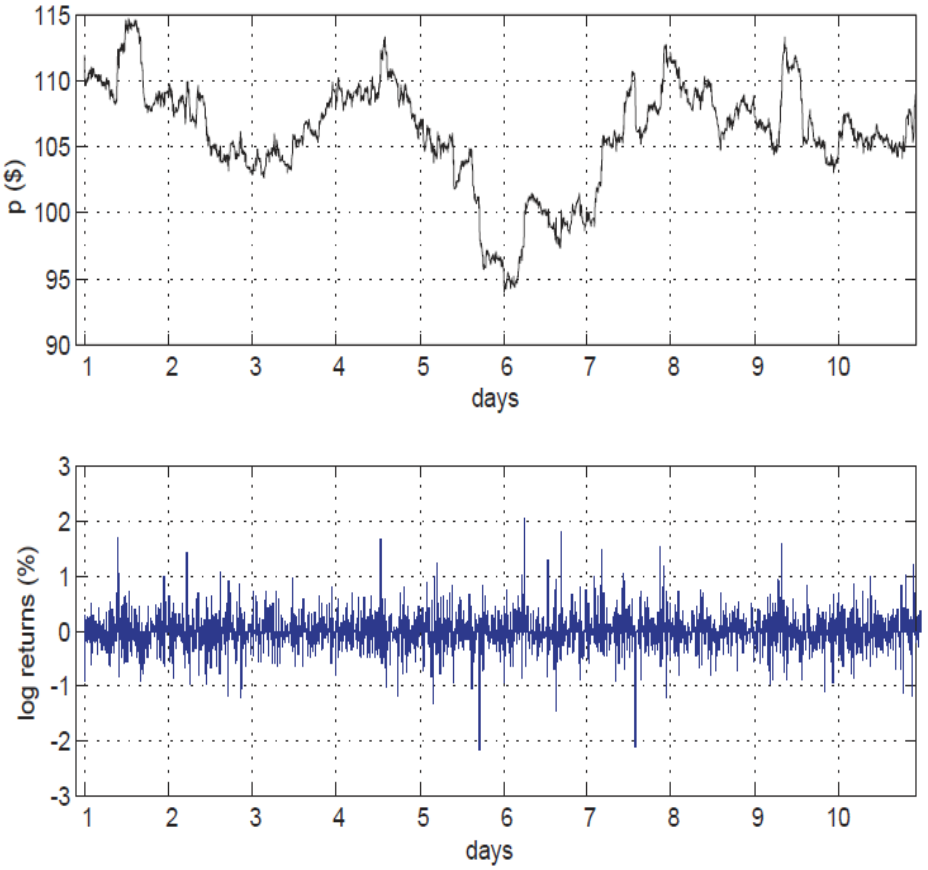


Results – Without Volatility Feedback

Our Results :

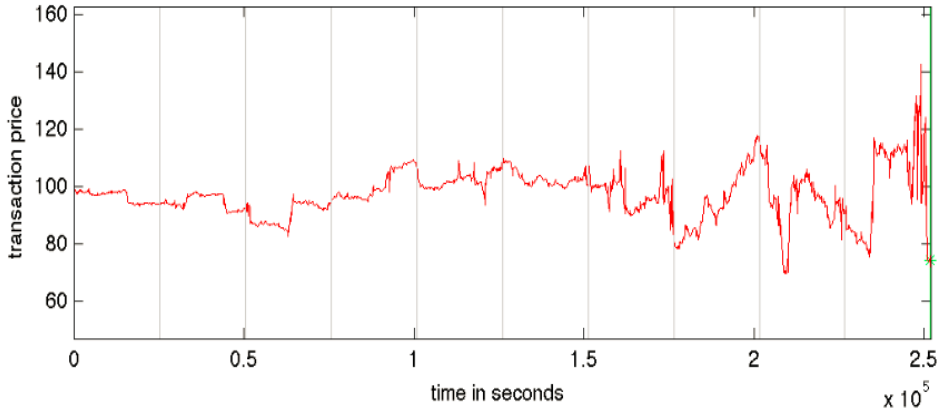


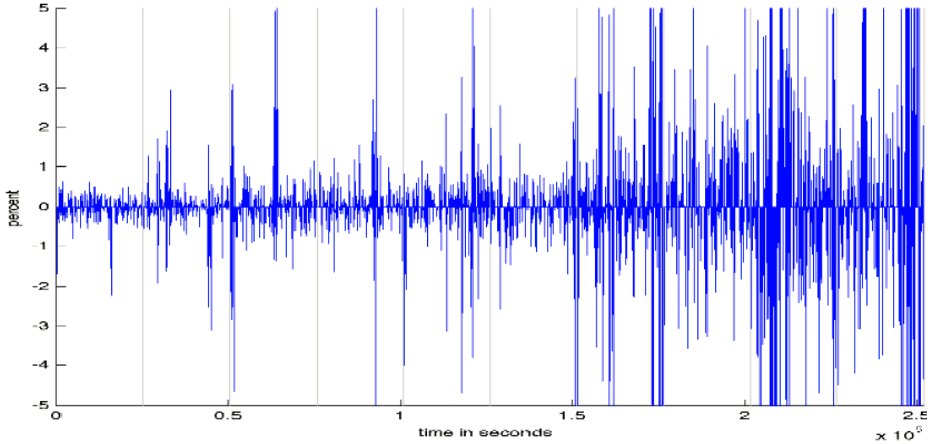
Raberto, 2005 :



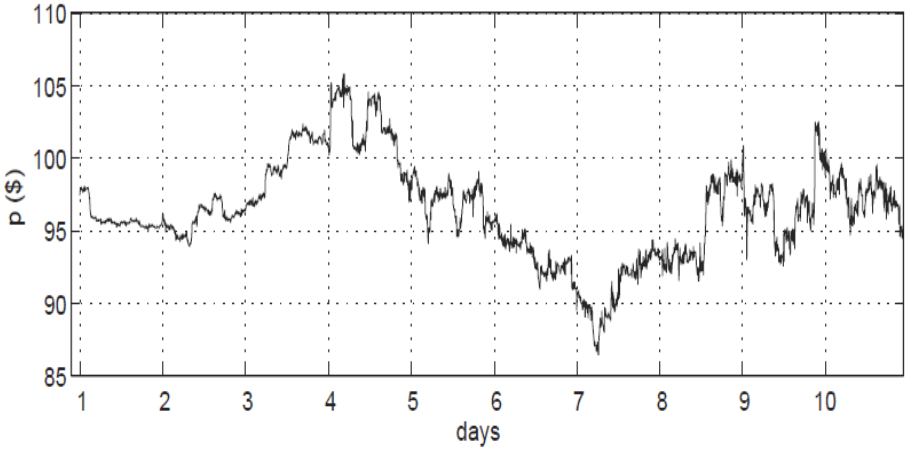
Results – With Volatility Feedback

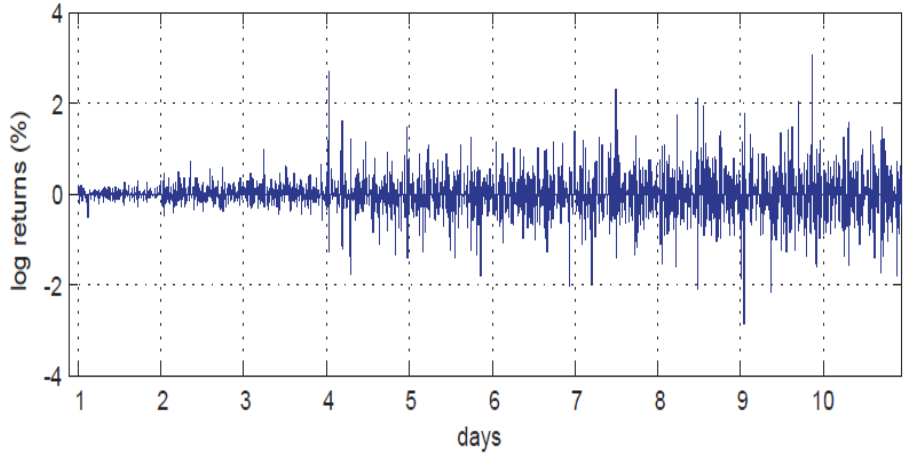
Our Results :





Raberto, 2005 :





Using Price Regulation to stabilize the Market

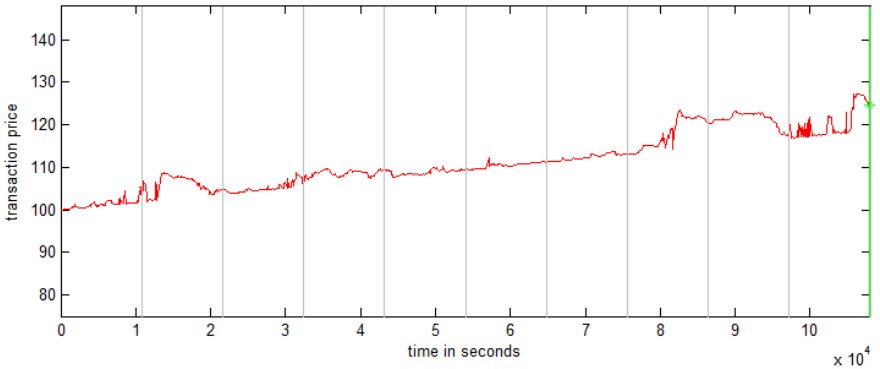
- Combination of price floor/ceiling
- In the model : prices which would be out of bounds are set to the nearest legal limit
- Enable growth (% per day)
- Result : regulation works. Market stays within bounds → stability
- However, no longer a free market → less economic welfare
- Tradeoff between stability and free market

$$\begin{cases} p_c(t) = p_{c0} + \dfrac{g \cdot p_{c0}}{24 \cdot 60 \cdot 60} \cdot t \\ p_f(t) = p_{f0} + \dfrac{g \cdot p_{f0}}{24 \cdot 60 \cdot 60} \cdot t \end{cases}$$

With g = growth (% per day)

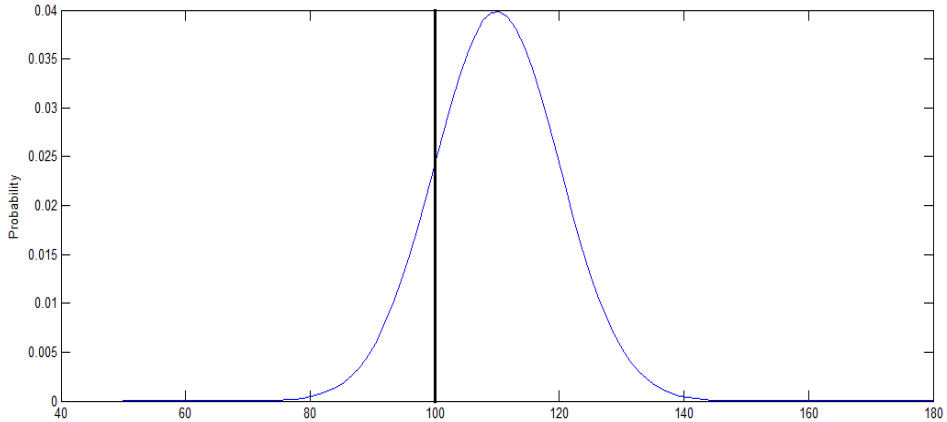
Example :

floor = 100
ceiling = 110
growth = 15% per day

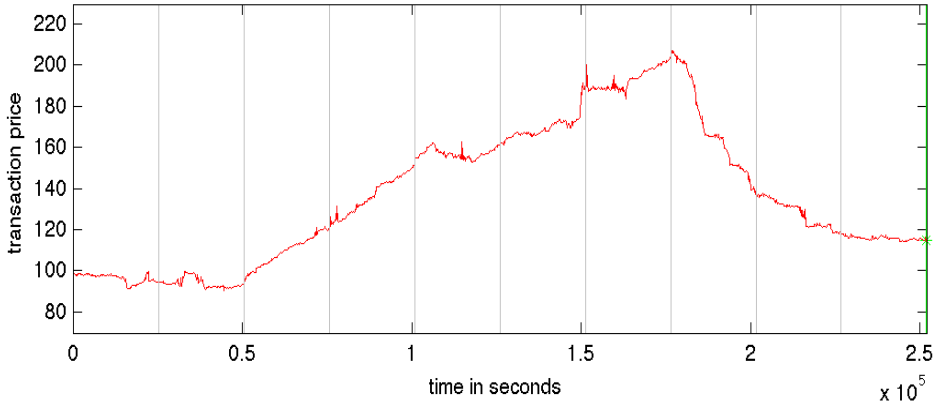


Modelling Financial Bubbles

Shifting the mean :



Result :



- Stock is consistently overvalued → prices explode
- Awareness of overvaluation → prices fall (bubble burst)
- In the model : at some point switch to a non-symmetric probability distribution for bid prices ⇒ prices are much more likely to increase. Later reverse the trend ⇒ prices are extremely likely to fall and keep falling
- Challenge : what causes the switch? When do we do it?
- Too many parameters

Implementation (1)

- Model implemented in 3 main sections:

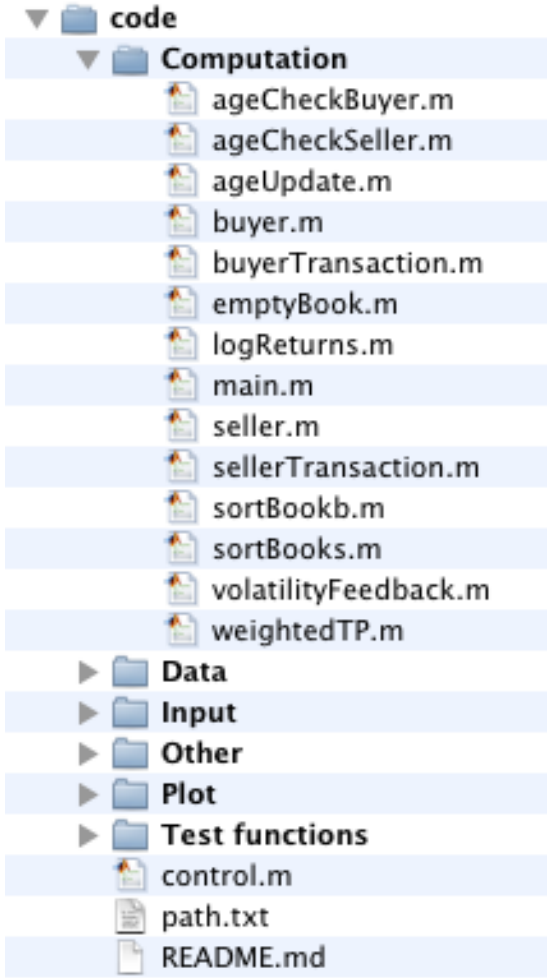
  → input generation

  → computation/simulation

  → result analysis (plot)
- Sections may be called seperately (independent)
- Input generation:

  → creates different sets of input parameters according to parameter sweep

  → saves to input files



Implementation (2)

- Computation/simulation:

  → runs model on specific set of parameters and saves the results

  → *main.m* function contains for-loop which iterates through each simulation second and calls different sub-functions (auction age update, new auction, book emptying, etc...)

- Result analysis:

  → generates plot of the results

  → live plot also available (simulation takes more time)

main.m (sample)

```
%% Simulation section
for i = 1:1:(SP.M)*(SP.T)


    %% Age Update
    SSM.bookbpaging = ageUpdate( SSM.bookbpaging, SSM.sbbp );
    SSM.bookspaging = ageUpdate( SSM.bookspaging, SSM.sbsp );

    %% Calculate Log Returns
    if i == lrt

        [ SSM ] = logReturns( SSM, SP );                    % calcu

        lrt = lrt + SP.dt;                                  % incre

    end


    %% New Book Entry Section
    if i == t

        Tau = 1+round(exprnd(SP.lambda));                   % step
                                                            % (rand
```

Parameter Sweep

- *param.txt* defines the sweep range of the involved parameters
- Every possible combination of parameters is then generated
- Input files are created

→ Example of *param.txt* :

```
SP.p0          100       1         110
SP.k           1.40      0.01      1.60
SP.volfeed     0         1         1
SP.entage      500       10        700
```

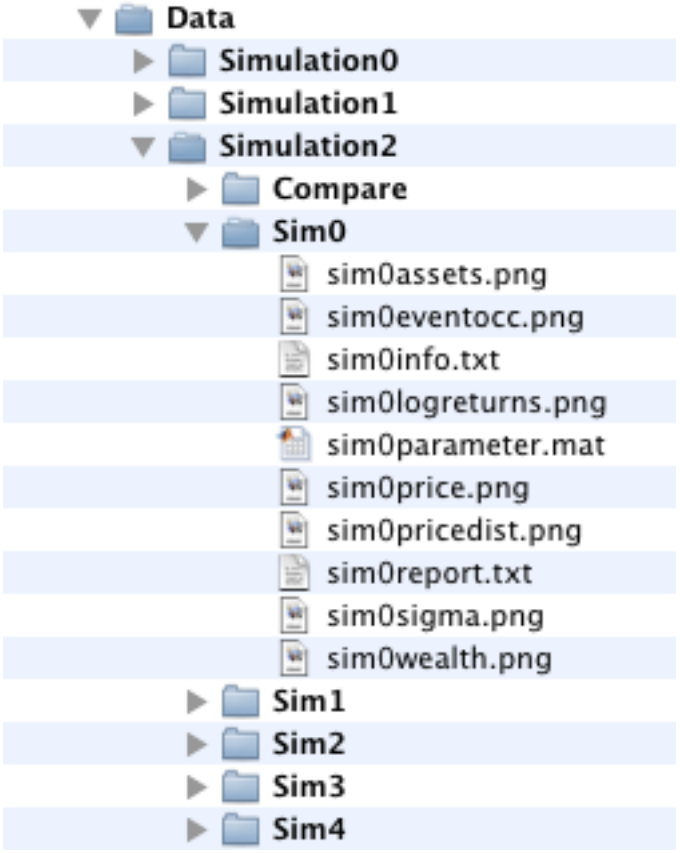
Automatic File Structure

- *control.m*:

  → generates file and folder structure using bash commands

  → executes parameter sweep according to chosen parameters (*param.txt*) and creates input files

  → runs simulation and calls plot functions
- *info-files* and *log-files* created for each simulation
- *"Compare folder"* with graphs to be compared



Information File (*sim0info.txt*)

```
                    **** SIMULATION 0 ****

                         Information file

                                  Date:  29-Oct-2012
                                  Time:  18:22:17

          Number of traders -        tnum:  100
          Amount of shares -     totShares:  100000
            Starting price -          p0:  100.00

    Mean of normal distribution -      mu:  1.0000
Initial std. deviation of normal dist. - sigma:  0.0050
 Parameter of exponential distribution - lambda:  20.0000

                    Total days -        M:  1
          Total time in seconds -        T:  10800

        Last tick iteration window -     dt:  60.00

             Empty book (On/Off) -    bkempty:  1
            Entry refresh (On/Off) -  entrefresh:  0
Entry erasure when aged (On/Off) -      entage:  1
          Multiple shares (On/Off) -  mulshares:  1
       Volatility feedback (On/Off) -   volfeed:  1
```

Our Experience - Why we didn't hand it in

- Time management
  → debugging can be VERY time consuming!
  → eg. ~12h : parameter is too big by a factor of 100

Problems we had to overcome

- Model badly described in paper
  → led to a lot of testing and bouncing ideas off eachother
- Very unclear concerning implementation of the model
  → eg. parameter choice
- Hard to control the model due to large number of random events
  → First simulated without random variables
- Things got complicated very quickly
  → Started with the bare minimum and slowly added to it

Things we did well

- MatLab Script/Function hierarchy/organization
  → keep track of the big picture
- Commenting almost every line of code
  → easier to find your way around a complex model
- Using existing MatLab functions
- Using data structures (eg. MatLab structs)
  → system state matrices as function inputs/outputs
  → grouping important output variables together
  → simplifies data analysis

Thanks for listening!