

Bubble-and-bust dynamics under walrasian asset pricing and heterogeneous traders

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Pisa, July 11th, 2014

- 1 Empirical stylised facts of financial time-series
- 2 Methodological perspective
- 3 The model
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- 5 Concluding remarks and conceivable extension

Empirical stylised facts

Empirical analysis of actual financial data often yields puzzling statistical properties. Notably, asset returns time-series use to exhibit:

- Volatility clustering;
- Excess volatility;
- Excess covariance;
- Heavy tails.

To some extent at odds with accepted theory and practice.

Financial bubbles: a largely neglected stylised fact



Financial bubbles (cont'd)

“if the reason that the price is high today is only because investors believe that the selling price will be high tomorrow – when ‘fundamental’ factors do not seem to justify such a price – then a bubble exists” (Stiglitz)

“a sharp rise in the price of an asset or a range of assets in a continuous process, with the initial rise generating expectations of further rises and attracting new buyers – generally speculators interested in profits from trading in the asset rather than its use or earnings capacity” (Kindleberger)

Our proposal

We set up a model able to yield:

- *endogenous* bubble-and-bust dynamics
- as a result of the sole interaction among *heterogeneous adaptive traders*
- highlighting booms and crashes as intrinsic features of financial markets.

Methodological perspective

Heterogeneous Agents Models

- analytical investigations of the dynamical systems representing the laws of motion of the economy;
- analytical tractability often leads to restrictive simplifying assumptions;
- focus on asymptotic properties.

Agent-Based Models

- computational (numerical) study of economies modelled as evolving systems of interacting agents;
- complex behaviour specifications;
- keep track of the whole dynamics.

The model

Following Anufriev et al. (2012), consider a pure-exchange economy:

- N heterogeneous traders (index $\mathcal{N} = \{1, \dots, n, \dots, N\}$);
- L long-lived risky securities (index $\mathcal{L} = \{1, \dots, \ell, \dots, L\}$);
- a riskless bond;
- time is discrete;
- risky securities, present in fixed amount, have ex-dividend price p_t^ℓ and pay a random dividend d_t^ℓ at the end of each period;
- the bond, inelastically supplied, have price normalized to 1 (*numéraire*) and yields $r_f > 0$ in every t ;
- trader wealth equals the market value of the portfolio he holds:

$$W_{n,t} = \sum_{\ell=1}^L A_{n,t}^\ell \cdot p_t^\ell + B_{n,t}$$

Trader behaviour

At the beginning of each time step, trader n invests a share $x_{n,t}^\ell$ of his wealth in security ℓ ; the decision is made according to the information set

$$\mathcal{I}_t = \{p_\tau^1, \dots, p_\tau^L; d_\tau^1, \dots, d_\tau^L \mid \tau < t\}$$

that is common knowledge, and to trader-specific investment function

$$f_n : \mathbb{R}^{\tau \times L} \longrightarrow \mathbb{R}^L \quad \text{such that} \quad \mathbf{x}_{n,t} = f_{n,t}(\mathcal{I}_t)$$

that is independent on wealth, as we assume overall CRRA attitude. The amount of wealth invested in the bond is residually determined:

$$x_{n,t}^0 = 1 - \sum_{\ell=1}^L x_{n,t}^\ell$$

The evolution of individual wealth, in terms of wealth fractions, holds:

$$W_{n,t} = W_{n,t-1} \cdot \left[x_{n,t-1}^0 \cdot (1 + r_f) + \sum_{\ell=1}^L x_{n,t-1}^\ell \cdot \left(\frac{p_t^\ell}{p_{t-1}^\ell} + e_t^\ell \right) \right]$$

Useful additional definitions

Dividend yield: the ratio of dividend over past realized price (proxy for fundamentals)

$$e_t^l = \frac{d_t^l}{p_{t-1}^l} \quad \forall l \in \mathcal{L}$$

Aggregate wealth: the sum of all individual wealth levels

$$W_t = \sum_{n=1}^N W_{n,t}$$

Individual wealth shares: the ratios of each individual wealth out of aggregate wealth

$$\varphi_{n,t} = \frac{W_{n,t}}{W_t} \quad \forall n \in \mathcal{N}$$

Market portfolio: the wealth-weighted sum of individual portfolios

$$\mathbf{x}_t = \sum_{n=1}^N \mathbf{x}_{n,t} \cdot \varphi_{n,t}$$

Trader optimisation problem

At every time step, each trader faces an optimisation problem of the form:

$$\max_{\mathbf{x}_t} \mathbb{E}[U(W_t)]$$

s.t.

$$W_t = W_{t-1} \cdot \left[x_{t-1}^0 \cdot (1 + r_f) + \sum_{\ell=1}^L x_{t-1}^\ell \cdot \left(\frac{p_t^\ell}{p_{t-1}^\ell} + e_t^\ell \right) \right]$$

For notational convenience the n subscript is dropped here. Coherently with CRRA attitude, the utility function reads:

$$U(W_t) = \frac{W_t^{1-\gamma} - 1}{1-\gamma}$$

where $\gamma > 0$ denotes the risk-aversion coefficient.

Trader expectations

We assume the trader forms expectations about future price returns and their (co)variances as smooth functions of the EWMA estimators over the information set previously defined:

$$\hat{\rho}_t^\ell = \lambda \cdot \sum_{\tau=0}^{\infty} (1 - \lambda)^\tau \cdot \rho_{t-\tau-1}^\ell$$

$$\hat{\sigma}_{\rho,t}^{\ell,h} = \lambda \cdot \sum_{\tau=0}^{\infty} (1 - \lambda)^\tau \cdot \left[\rho_{t-\tau-1}^\ell - \hat{\rho}_{t-\tau-1}^\ell \right] \cdot \left[\rho_{t-\tau-1}^h - \hat{\rho}_{t-\tau-1}^h \right]$$

where $\rho_t^\ell = \frac{p_t^\ell}{p_{t-1}^\ell} - 1$ is the price return of security ℓ between $t - 1$ and t .

The decay factor $\lambda \in [0, 1]$ captures the way relative weights are distributed across more recent and older observations.

Trader investment function

We adopt the same mean-variance approximation of the optimal investment function proposed in Chiarella and He (2001):

$$x_t = f(\mathcal{I}_t) = \frac{1}{\gamma} \cdot \hat{\mathbb{C}}_t^{-1} \cdot \left[\mathbb{E}_t - r_f \cdot \mathbf{1} \right]$$

where \mathbb{E}_t and $\hat{\mathbb{C}}_t^{-1}$ are, respectively, the vector of expected total returns and the inverse of the expected variance-covariance matrix, whose elements read:

$$\begin{aligned}\mathbb{E}_t^\ell &= \bar{e}^\ell + d \cdot \hat{\rho}_t^\ell \\ \hat{\mathbb{C}}_t^{\ell,h} &= \hat{\sigma}_{\rho,t}^{\ell,h} + \sigma_e^{\ell,h}\end{aligned}$$

where d is a behavioural parameter:

- $d = 0$ trader is a fundamentalist;
- $d > 0$ trader is a trend-chaser;
- $d < 0$ trader is a trend-contrarian.

Demand and equilibrium condition

Individual demand for a risky asset reads:

$$z_{n,t}^l = \frac{x_{n,t}^l \cdot W_{n,t}}{p_t^l} \quad \forall l \in \mathcal{L}$$

Normalising the supply of each risky asset to 1, the equilibrium condition reads:

$$\sum_{n=1}^N z_{n,t} = \mathbf{1}$$

Solving for single security prices yields:

$$p_t^l = \sum_{n=1}^N x_{n,t}^l \cdot W_{n,t}(\mathbf{p}_t) \quad \forall l \in \mathcal{L}$$

where prices appear both in the LHS and RHS of the equation, as argument of wealth evolution.

Pricing of risky assets

Proposition

If short positions are not allowed, i.e.

$$x_{n,t}^{\ell} \in (0, 1) \quad \forall n \in \mathcal{N}, \forall \ell \in \mathcal{L}, \forall t$$

then prevailing prices exist, are unique and strictly positive. It holds:

$$W_t = W_{t-1} \cdot \frac{x_{t-1}^0 \cdot (1 + r_f) + \sum_{\ell=1}^L x_{t-1}^{\ell} \cdot e_t^{\ell}}{x_t^0}$$

$$p_t^{\ell} = p_{t-1}^{\ell} \cdot \frac{x_t^{\ell}}{x_{t-1}^{\ell}} \cdot \frac{x_{t-1}^0 \cdot (1 + r_f) + \sum_{\ell=1}^L x_{t-1}^{\ell} \cdot e_t^{\ell}}{x_t^0}$$

Proof.

See **Appendix A** in the paper. □

Simulation results: survival patterns

Our first simulation studies the way market selection occurs among different trading strategies by means of individual wealth shares $\varphi_{n,t}$

Definition

A trader n is said to “*survive*” the economy if his long-run wealth-share is significantly different from 0, i.e. if $\lim_{t \rightarrow \infty} \varphi_{n,t} > 0$.

A trader n is said to “*dominate*” the economy if his long-run wealth-share is significantly close to 1, i.e. if $\lim_{t \rightarrow \infty} \varphi_{n,t} = 1$.

Following Anufriev et al. (2006), two types of equilibria are possible:

- 1 Single-survivor equilibria (most ‘*aggressive*’ trader);
- 2 Multiple-survivor equilibria (though non-generic).

Survival patterns (cont'd)

Description	Value
Initial population size	$N = 200$
Number of risky assets	$L = 1$
Static population	true
Riskless rate of return	$r_f = 0.02$
γ distribution	$\gamma_n \sim \mathcal{U}(1.0, 1000.0)$
λ distribution	$\lambda_n = 0.1, \forall n \in \mathcal{N}$
d distribution	$d_n = 1.0, \forall n \in \mathcal{N}$
Initial wealth endowment	$W_{n,0} = 50.0, \forall n \in \mathcal{N}$
Yield mean	$\bar{e} = 0.04$
Yield variance	$\sigma_e^2 = 1.0\text{e-}4$
Yield realisation distribution	$e_t \sim \mathcal{N}(\bar{e}, \sigma_e^2)$
Initial risky asset price level	$p_0 = 0.1$
x_n admissible interval	$x_{n,t} \in [0.01, 0.99], \forall n \in \mathcal{N}, \forall t$

Table: Parameters and initial conditions (1)

Survival patterns (cont'd)

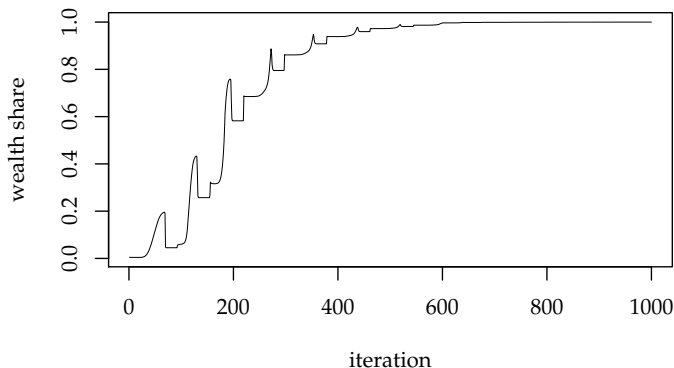


Figure: Evolution of wealth-share for the least-risk-averse trader. Single-survivor.

Survival patterns (cont'd)

Description	Value
Initial population size	$N = 200$
Number of risky assets	$L = 1$
Static population	true
Riskless rate of return	$r_f = 0.02$
γ distribution	$\gamma_n \sim \mathcal{U}(100.0, 1000.0)$
λ distribution	$\lambda_n = 0.1, \forall n \in \mathcal{N}$
d distribution	$d_n = 1.0, \forall n \in \mathcal{N}$
Initial wealth endowment	$W_{n,0} = 50.0, \forall n \in \mathcal{N}$
Yield mean	$\bar{e} = 0.04$
Yield variance	$\sigma_e^2 = 1.0e-4$
Yield realisation distribution	$e_t \sim \mathcal{N}(\bar{e}, \sigma_e^2)$
Initial risky asset price level	$p_0 = 0.1$
x_n admissible interval	$x_{n,t} \in [0.01, 0.99], \forall n \in \mathcal{N}, \forall t$

Table: Parameters and initial conditions (2)

Survival patterns (cont'd)

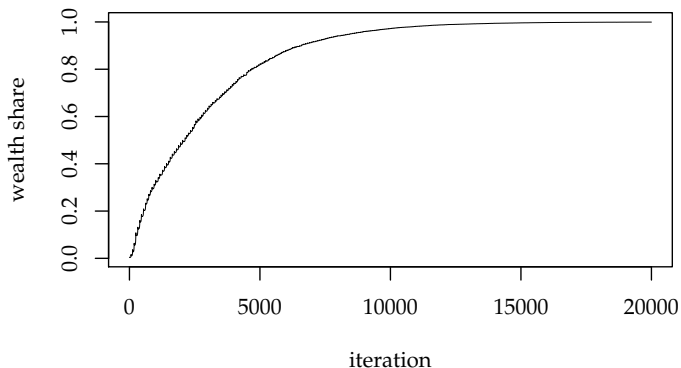


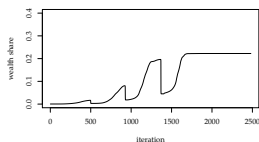
Figure: Evolution of wealth-share for the least-risk-averse trader. Single-survivor.

Survival patterns (cont'd)

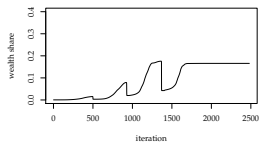
Description	Value
Initial population size	$N = 200$
Number of risky assets	$L = 1$
Static population	true
Riskless rate of return	$r_f = 0.02$
γ distribution	$\gamma_n \sim \mathcal{U}(1.0, 1000.0)$
λ distribution	$\lambda_n = 0.01, \forall n \in \mathcal{N}$
d distribution	$d_n = 1.0, \forall n \in \mathcal{N}$
Initial wealth endowment	$W_{n,0} = 50.0, \forall n \in \mathcal{N}$
Yield mean	$\bar{e} = 0.04$
Yield variance	$\sigma_e^2 = 1.0e-4$
Yield realisation distribution	$e_t \sim \mathcal{N}(\bar{e}, \sigma_e^2)$
Initial risky asset price level	$p_0 = 0.1$
x_n admissible interval	$x_{n,t} \in [0.01, 0.99], \forall n \in \mathcal{N}, \forall t$

Table: Parameters and initial conditions (3)

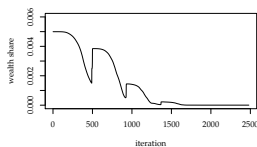
Survival patterns (cont'd)



(a) lowest γ_n



(b) second-lowest γ_n



(c) highest γ_n

Figure: Multiple-survivor equilibrium. Evolution of wealth-shares.

Survival patterns (cont'd)

Heterogeneity in the risk-aversion coefficient, within the stability domain of the system, triggers a wealth-driven selection mechanism.

- For a short memory-span (large λ), the least risk-averse trader survives and dominates the economy;
- For a long memory-span (small λ), multiple traders, still low-risk averse, survive and display identical investment decisions.

Out of the stability domain of the system (i.e. for large enough λ , following Anufriev et al. 2006) selection does not occur: individual wealth-shares keep fluctuating indefinitely with no clear-cut outcome.

Simulation results: transitional price dynamics

Description	Value
Initial population size	$N = 1,000$
Number of risky assets	$L = 1$
Static population	true
Riskless rate of return	$r_f = 0.02$
γ distribution	$\gamma_n \sim \mathcal{U}(1.0, 500.0)$
λ distribution	$\lambda_n = 0.0036, \forall n \in \mathcal{N}$
d distribution	$d_n = 1.0, \forall n \in \mathcal{N}$
Initial wealth endowment	$W_{n,0} = 50.0, \forall n \in \mathcal{N}$
Yield mean	$\bar{e} = 0.04$
Yield variance	$\sigma_e^2 = 1.0e-4$
Yield realisation distribution	$e_t \sim \mathcal{N}(\bar{e}, \sigma_e^2)$
Initial risky asset price level	$p_0 = 0.1$
x_n admissible interval	$x_{n,t} \in [0.01, 0.99], \forall n \in \mathcal{N}, \forall t$

Table: Parameters and initial conditions (4)

Transitional price dynamics (cont'd)

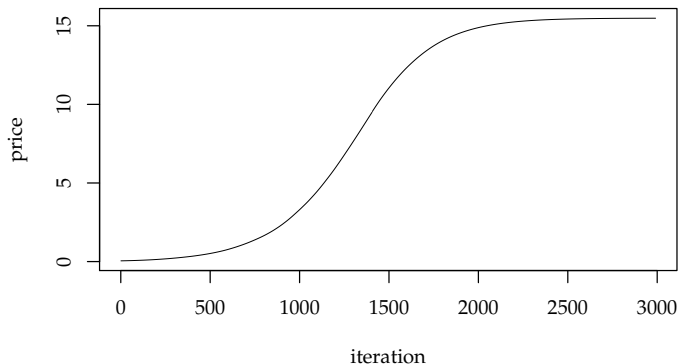


Figure: Price dynamics. Smooth and monotone convergence to equilibrium.

Transitional price dynamics (cont'd)

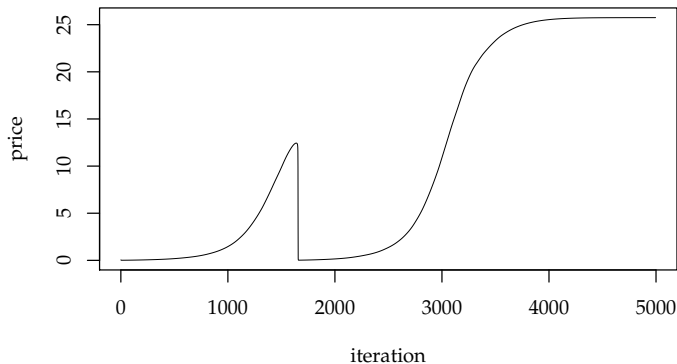


Figure: Price dynamics. Emergence of a bubble-and-bust cycle.

$$\lambda_n = 0.00365, \forall n \in \mathcal{N}$$

Transitional price dynamics (cont'd)

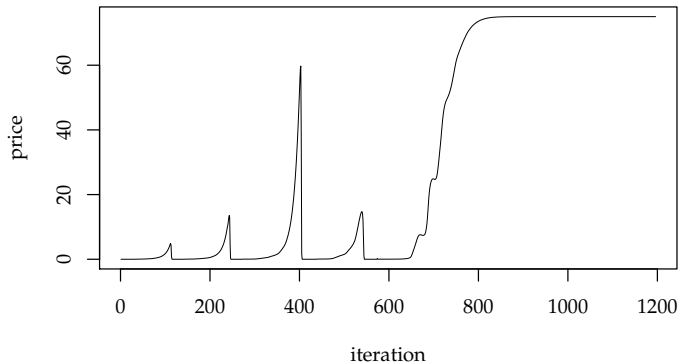


Figure: Price dynamics. Emergence of multiple bubble-and-bust cycles.

$$\lambda_n = 0.155, \forall n \in \mathcal{N}$$

Transitional price dynamics (cont'd)

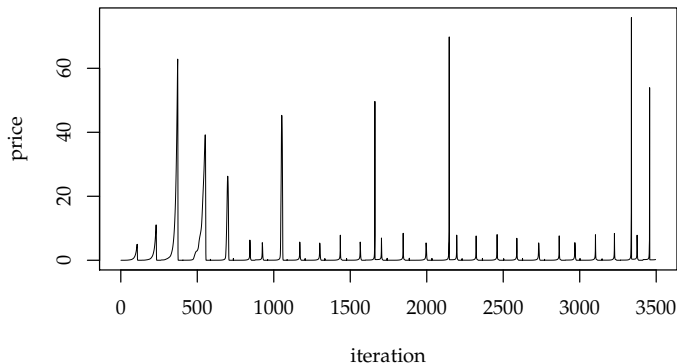


Figure: Price dynamics. No convergence to an equilibrium value.

$$\lambda_n = 0.16, \forall n \in \mathcal{N}$$

Transitional price dynamics (cont'd)

The emergent properties observed in a trend-chasers-only setting maintain robustness with respect to the introduction of fundamentalist and trend-contrarian traders in the economy.

- Fundamentalists are expected to stabilise the price as they act against chartists whenever current price deviates from its fundamental value;
- Contrarians shall counteract the attempt made by trend-chasers to exacerbate the price trend by acting in a symmetrical fashion.

We differentiate the d parameter in order to model a population largely composed of quasi-fundamentalists and well balanced crowds of trend-chasers and contrarians.

Transitional price dynamics (cont'd)

We now shift the analysis to the transitional price dynamics:

Description	Value
Initial population size	$N = 1,000$
Number of risky assets	$L = 1$
Static population	true
Riskless rate of return	$r_f = 0.02$
γ distribution	$\gamma_n \sim \mathcal{U}(1.0, 500.0)$
λ distribution	$\lambda_n = 0.1, \forall n \in \mathcal{N}$
d distribution	$d_n \sim \mathcal{N}(0, 1)$
Initial wealth endowment	$W_{n,0} = 50.0, \forall n \in \mathcal{N}$
Yield mean	$\bar{e} = 0.04$
Yield variance	$\sigma_e^2 = 1.0e-4$
Yield realisation distribution	$e_t \sim \mathcal{N}(\bar{e}, \sigma_e^2)$
Initial risky asset price level	$p_0 = 0.1$
x_n admissible interval	$x_{n,t} \in [0.01, 0.99], \forall n \in \mathcal{N}, \forall t$

Table: Parameters and initial conditions (5)

Transitional price dynamics (cont'd)

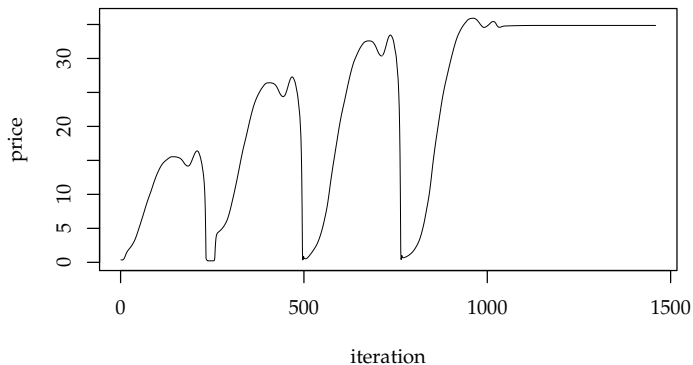


Figure: Price dynamics. Fundamentalists vs. chartists.

Transitional price dynamics (cont'd)

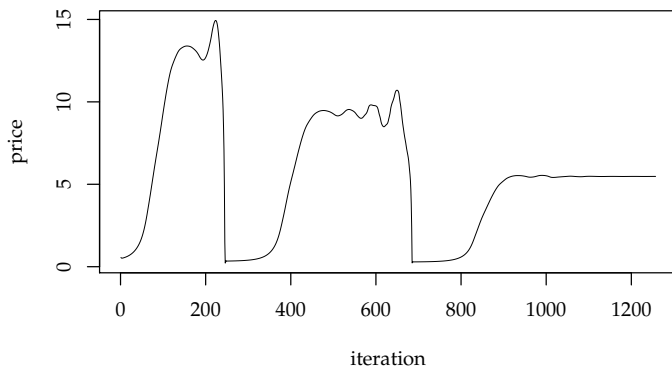


Figure: Price dynamics. Fundamentalists vs. chartists. $\gamma_n \sim \mathcal{U}(1.0, 1000.0)$

Concluding remarks

1 Analysis of individual wealth-shares

- strong market selection mechanism
- single- and multiple-survivor equilibria
- riskier investment functions globally dominate

2 Analysis of transitional price dynamics

- emergence of bubble-and-bust cycles
- robust to the introduction of fundamentalist and trend-contrarian

In a nutshell

risk-aversion heterogeneity



decoupling of price dynamics from the fundamental yield process

Conceivable improvements

Our framework can be extended in a number of directions:

- multiple risky assets;
- dynamic population;
- more realistic traders' behaviour (prospect theory, herding) and learning (genetic algorithms, classifier systems).

Our conjecture is that a sharper departure from rationality assumption is needed in order to obtain dynamics that are closer to reality.

Selected references

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