Some Reflections on Agent-Based Models, Bounded Rationality and Heterogeneous Expectations

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Questions

- a. Is **simulation** an effective tool to go from microfoundations with heterogeneous agents to macro conclusions?
- b. How to handle the problem of individual expectations?
- c. How to handle **price determination** and **market clearing** (or not)?

Simulations alone are **not** enough, because

- "wilderness of ABMs", too many degrees of freedom;
- often simulation results are not (easily) reproducible;
- Imited insights: what exactly causes the (macro) outcome?

Einstein: A scientific theory should be **as simple as possible**, but **no simpler**.

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Strong need for the **simplest ABM** explaining macro phenomena through micro interactions

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How to model Heterogeneous Expectations of Boundedly Rational Individuals?

Brock and Hommes (1997): as an **evolutionary selection process** among different rules ranging from simple to sophisticated

- agents choose from a (small) list of simple forecasting heuristics and more complicated (costly) rules
- ► adaptive learning: some parameters of the heuristics are updated over time, e.g. anchor = time average
- performance based reinforcement learning: agents evaluate the performances of all heuristics, and tend to switch to more successful rules; impacts are evolving over time

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- small list of forecasting heuristics with fixed parameters or some adaptive learning
- evolutionary selection of better performing rules, with 2 or 3 parameters for speed of switching (details not important)
- same heterogeneous expectations model fits empirical data and laboratory experiments in different market environments

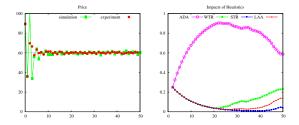
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Lab Experiments and Heuristics Switching Model: (Hommes, 2010)

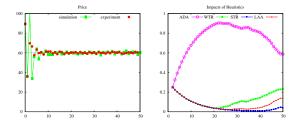


for Muth's cobweb model with **negative expectations feedback** quick convergence to RE

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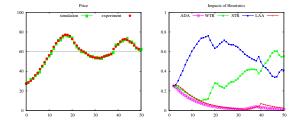
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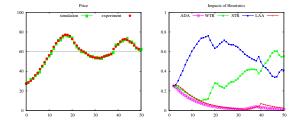
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General finding: in **positive expectations feedback** systems, **trendfollowing behavior** performs relatively well and causes persistent deviations from RE benchmark

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What about **price determination** and **market clearing** (or not)?

Two observations:

- price adjustment rule through excess demand/supply has speed of adjustment parameter which critically affects dynamics. How large is it??
- equilibrium model with heterogeneous expectations can explain important macro phenomena, such as persistence in inflation and output, price stickiness, misalignments, excess volatility, etc.

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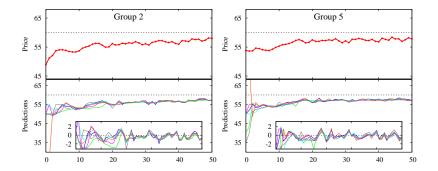
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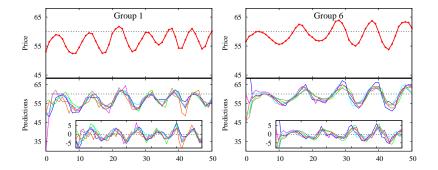
2 Groups with (Almost) Monotonic Convergence

prices, individual predictions and individual errors



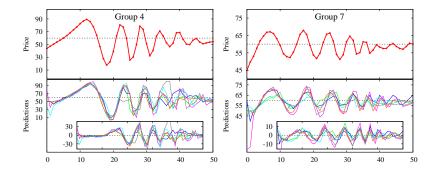
2 Groups with Perpetual Oscillations

prices, individual predictions and individual errors



2 Groups with Damping Oscillations

prices, individual predictions and individual errors



Four forecasting heuristics

adaptive rule

ADA $p_{1,t+1}^e = 0.65 p_{t-1} + 0.35 p_{1,t}^e$

weak trend-following rule

WTR
$$p_{2,t+1}^e = p_{t-1} + 0.4 (p_{t-1} - p_{t-2})$$

strong trend-following rule

STR
$$p_{3,t+1}^e = p_{t-1} + 1.3 (p_{t-1} - p_{t-2})$$

anchoring and adjustment heuristics with learnable anchor

LAA
$$p_{4,t+1}^e = 0.5 p_{t-1}^{av} + 0.5 p_{t-1} + (p_{t-1} - p_{t-2})$$

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Evolutionary Switching

Brock and Hommes (1997), Anufriev and Hommes (2009)

performance measure of heuristic *i* is

$$U_{i,t-1} = -(p_{t-1} - p_{i,t-1}^e)^2 + \eta U_{i,t-2}$$

parameter $\eta \in [0,1]$ – the strength of the agents' memory

discrete choice model with asynchronous updating

$$n_{i,t} = \delta n_{i,t-1} + (1-\delta) \frac{\exp(\beta U_{i,t-1})}{\sum_{i=1}^{4} \exp(\beta U_{i,t-1})}$$

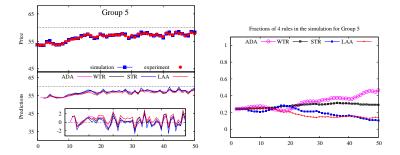
parameter $\delta \in [0, 1]$ – the inertia of the traders parameter $\beta \ge 0$ – the intensity of choice

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Group 5 (Convergence)

experimental prices simulated prices, predictions and errors

Parameters: $\beta = 0.4, \eta = 0.7, \delta = 0.9$



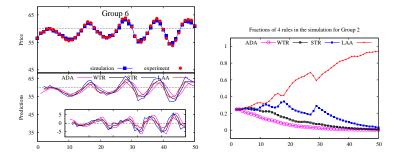
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Group 6 (Constant Oscillations)

experimental prices simulated prices, predictions and errors

Parameters: $\beta = 0.4, \eta = 0.7, \delta = 0.9$



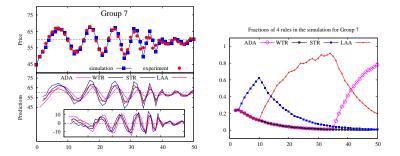
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Group 7 (Damping Oscillations)

experimental prices simulated prices, predictions and errors

Parameters: $\beta = 0.4, \eta = 0.7, \delta = 0.9$



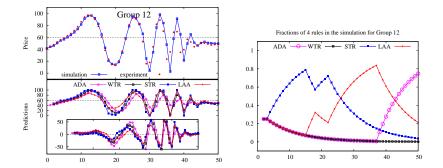
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Asset Pricing Experiments without Fundamental Trader

experimental prices simulated prices, predictions and errors

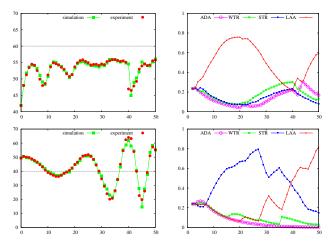
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More Asset Pricing Experiments Group 3 (Typing Error) and Fundamental $p^* = 40$



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Positive versus Negative Feedback Experiments

Heemeijer et al. (JEDC 2009); Te Bao, MPhil thesis, 2009

negative feedback (strategic substitute environment)

$$p_t = 60 - \frac{20}{21} \left[\sum_{h=1}^{6} \frac{1}{6} p_{ht}^e \right] - 60 \right] + \epsilon_t$$

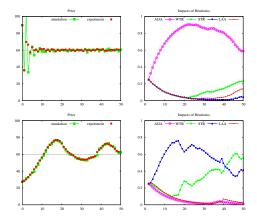
positive feedback (strategic complementarity environment)

$$p_t = 60 + \frac{20}{21} \left[\sum_{h=1}^{6} \frac{1}{6} p_{ht}^e - 60 \right] + \epsilon_t$$

- different types of shocks ϵ_t : small resp. large permanent shocks
- common feature: same RE equilibrium
- only difference: sign in the slope of linear map +0.95 vs -0.95

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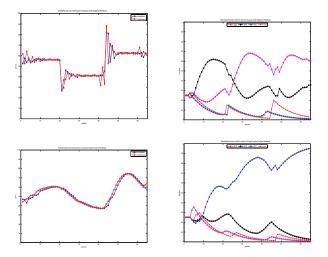
Positive vs Negative Feedback; Small Shocks Heuristics Switching Model Simulations Parameters: $\beta = 0.4, \eta = 0.7, \delta = 0.9$



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Positive/Negative Feedback; Large Shocks

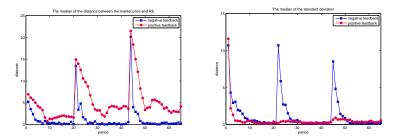


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Positive/Negative Feedback; Large Shocks Coordination & Price Discovery median absolute distance to RE fundamental price;

median standard deviation of individual predictions



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New Keynesian Macro Model; Expectations on Inflation & Output Gap

Assenza et al. (2010)

$$y_t = y_{t+1}^e - \varphi(i_t - \pi_{t+1}^e) + g_t, \quad \text{output}$$
(1)

$$\pi_t = \lambda y_t + \beta \pi_{t+1}^e + u_t, \quad \text{inflation}$$
(2)

$$i_t = \phi_\pi(\pi_t - \overline{\pi}) + \overline{\pi}, \quad \text{interest rate rule}$$
(3)

Two treatments:

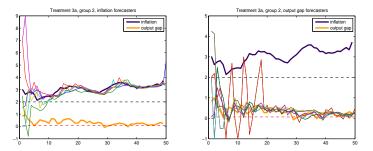
- **passive** monetary policy ($\phi_{\pi} = 1$)
- aggressive monetary policy ($\phi_{\pi} = 1.5$)

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New Keynesian Macro Model; Expectations on Inflation & Output Gap

Assenza et al. (2010)

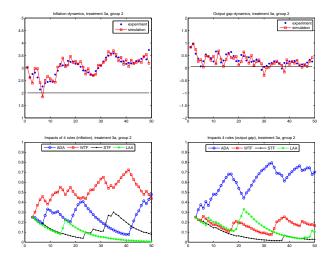
passive monetary policy (i.e. $\phi_{\pi} = 1$)



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New Keynesian Macro Model: Simulations (Domenico Massaro)



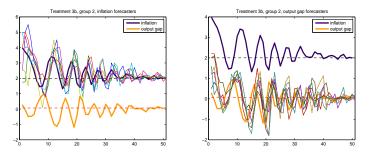
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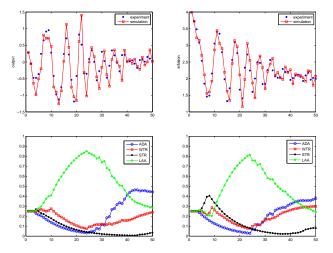
aggressive monetary policy (i.e. $\phi_{\pi} = 1.5$)



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New Keynesian Macro Model: Simulations (Domenico Massaro)



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Concluding Remarks on Heterogeneous Expectations

- **no homogeneous** expectations model fits **all** experiments
- only in stable cobweb/negative feedback quick convergence to REE
- heterogeneity in expectations is crucial, because one model explains observed
 - path dependence in same market environment
 - different aggregate outcomes in different markets
 - different forecasting behavior for different variables in one macro economy
- challenge: universal theory of heterogeneous expectations