Experiments and Models of Non-Rational Behaviour

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Main Ingredients of the Talk

- financial markets and the economy as complex evolving systems of interacting agents
- behavioral theory of heterogeneous expectations of boundedly rational individuals
- empirical validation of individual (micro) and aggregate (macro) behaviour through laboratory experiments

Economy as Expectations Feedback System

beliefs/expectations



realizations

mapping from all heterogeneous beliefs to price realizations

$$p_t = F(p_{1,t+1}^e, p_{2,t+1}^e, \cdots, p_{H,t+1}^e)$$

simple mapping from average beliefs into price realizations

$$p_t = f(\frac{1}{H}\sum_{h=1}^{H} p_{h,t+1}^e) = f(\overline{p_{t+1}^e})$$

rational solution: beliefs on average equal to realizations

 $p^{\ast}=f(p^{\ast}):$ perfectly self-fulfilling expectations

Learning to Forecasts Laboratory Experiments

- individuals only have to forecast price, ceteris paribus,
 e.g. with all other behavior assumed to be rational,
 demand/supply derived from profit/utility maximization
- computerized trading yields market equilibrium price, consistent with benchmark model, e.g.
 - cobweb model
 - asset pricing model

Experiments

- New Keynesian macro model
- advantage: clean data on expectations
- Challenge: universal theory of heterogeneous expectations

Learning to Forecast Experiments (Ctd)

Subjects' task and incentive

- forecasting a price for 50 periods
- better forecasts yield higher earnings

Subjects know

- only qualitative information about the market
- \triangleright price p_t derived from equilibrium between **demand** and **supply**
- type of expectations feedback: positive or negative
- **past information**: at time t participant h can see past prices (up to p_{t-1}), own past forecasts (up to $p_{t,h}$) and own earnings (up to $e_{t-1,h}$)

Subjects do not know

- exact equilibrium equation, e.g. $p_t = f(\bar{p}_{t+1}^e)$ or $p_t = f(\bar{p}_t^e)$
- exact **demand schedule** of themselves and others
- number and forecasts of other participants

Example Computer Screen Experiment



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Three Different Experimental Settings

- asset pricing experiment (with/without robot trader)
 - two-period ahead
 - positive feedback

$$p_t = \frac{1}{1+r} \Big((1-n_t) \frac{p_{t+1,1}^e + \dots + p_{t+1,6}^e}{6} + n_t \, p^f + \bar{y} + \varepsilon_t \Big)$$

• **positive** versus **negative** feedback; one-period ahead $p_t = f(p_t^e)$:

- ▶ **positive** feedback: linear, slope +0.95;
- **negative** feedback: linear, slope -0.95.
- New Keynesian Macromodel: aggregate inflation and output depend on individual forecasts of both inflation and output (and monetary policy rule):

$$(\pi_t, y_t) = F(\pi_{t+1}^e, y_{t+1}^e)$$

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Conclusions

Asset Pricing Experiment Simulation Benchmarks



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Asset Pricing Experiment (with Robot Trader)



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Asset Pricing Experiment

Strong coordination of individual forecasts and errors



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Conclusions

Groups with (Almost) Monotonic Convergence prices, individual predictions and individual errors



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2 Groups with Perpetual Oscillations

prices, individual predictions and individual errors



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2 Groups with Damping Oscillations

prices, individual predictions and individual errors Group 4 Group 7 Price Price Predictions Predictions -10 -30

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Summary Results Asset Pricing Experiment

Results are inconsistent with rational, fundamental forecasting

One would like to explain:

- three qualitatively different patters
 - (almost) monotonic convergence
 - constant oscillations
 - damping oscillations
- coordination of agents in their predictions
- no homogeneous expectations model fits these experiments need heterogeneous expectations model

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Estimation of Individual Predictions

... for the last 40 periods

in converging groups agents use adaptive expectations

 $p_{t+1}^e = w \, p_{t-1} + (1 - w) \, p_t^e$

often agents used simple linear rules anchor and adjustment rule

$$\begin{array}{lll} p_{t+1}^e = & \alpha + \beta_1 \, p_{t-1} + \beta_2 \, p_{t-2} \\ \text{e.g.} & (60 + p_{t-1})/2 + (p_{t-1} - p_{t-2}) \\ \text{or LAA} & (p_{t-1}^{av} + p_{t-1})/2 + (p_{t-1} - p_{t-2}) \end{array}$$

in particular trend-extrapolating rules

$$p_{t+1}^{e} = p_{t-1} + \gamma \left(p_{t-1} - p_{t-2} \right) \qquad 0.4 \le \gamma \le 1.3$$

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Examples of Individual Predictions and Switching







Group 7, participant 3



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Heterogeneous Expectations Heuristics Switching Model Anufriev and Hommes, AEJ:Micro 2012

- agents choose from a number of simple forecasting heuristics
- ► adaptive learning: some parameters of the heuristics are updated over time, e.g. anchor = average
- performance based reinforcement learning: (extension of Brock and Hommes, *Econometrica* 1997) agents evaluate the performances of all heuristics, and tend to switch to more successful rules; impacts are evolving over time

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

strong trend-following rule

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

anchoring and adjustment heuristics with learnable anchor

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

Evolutionary Switching with Asynchronous Updating

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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Stochastic Simulations (one step ahead forecast)

Anufriev and Hommes (2012)

- uses past experimental data
- **same information** as participants in experiments

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

- initial fractions equal, i.e. $n_{ht} = 0.25$
- initial prices as in experiments

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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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Muth (1961) on Deviations from Rationality [emphasis added]

Allowing for **cross-sectional differences** in expectations is a simple matter, because their **aggregate affect is negligible** as long as the deviation from the rational forecast for an individual firm is **not strongly correlated with those of the others**. Modifications are necessary only if the **correlation of the errors is large** and depends systematically on other explanatory variables.

key issues:

- are individual expectations coordinated?
- if so, do individuals coordinate on a rational or a boundedly rational aggregate outcome?

This can be tested in Learning to Forecast Experiments

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Positive versus Negative Feedback Experiments Heemeijer et al. (JEDC 2009); Bao et al. (JEDC 2012

negative feedback (strategic substitute environment)

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

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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Negative vs. Positive Feedback Experiments

Prices, Individual Predictions and Errors



Positive Feedback: coordination on "wrong" non-RE price; coordination on **almost self-fulfilling equilibria**

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Negative Feedback Experiment: Session 1

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Price in Experiments with Negative Feedback (6 groups) (Heemeijer et al., JEDC 2009)



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Positive Feedback Experiment: Session 1

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Prices in Experiments with Positive Feedback (7 groups)

(Heemeijer et al., JEDC 2009)



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Positive vs Negative Feedback; Small Shocks Heuristics Switching Model Simulations



positive feedback: trend-followers amplify fluctuations

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Conclusion: Empirical and Exper. Data consistent with Complexity View

- simple heterogeneous expectations heuristics switching model fits experimental micro and macro data quite nicely
- heterogeneity and heuristics switching explains
 - path dependence
 - different behaviour in different feedback systems
 - different behaviour in aggregate variables of same economy
- agents are behaviorally rational at the individual level: they use simple heuristics such as adaptive expectations, trend following rules and anchor and adjustment rules
- positive feedback markets are "irrational" due to coordination on "wrong" price and survival of (almost) self-fulfilling trend following strategies