

Inflation expectation formation. Laboratory experiment in high and low inflation environment

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Motivation

Inflation expectations are one of the main determinants of inflation:

Theory: New Keynesian Phillips curve

Empirical studies: Galí, Gertler and López-Salido (2001), Taylor (2000)

Policy: efforts of central banks in measuring inflation expectations



But inflation expectations are unobservable!



Survey or experiment?

Methods

Analyze survey data	Conduct an experiment
+Panel data are available for many countries	+ Control the information set that subjects have
+Data for long periods of time	+Control the incentives
-No incentive to truthfully report the expectations	+ Control the shocks that hit the experimental economy
- Decision in real life situations may be based on different expectations than stated in the survey	-Small sample => internal validity issues
- Many shocks may hit the economy at the same time, so it is hard to disentangle the effects of various shocks	-External validity issues (artificial setting, etc.)
- Data on expectations from many developing countries is either not available or not reliable	

Main idea

Research questions:

- How are individual inflation expectations formed and adapted?
- How are aggregate expectations formed?

Hypotheses:

- The majority of people do not switch between forecasting rules. Instead, they use adaptive learning model ($\pi_t^e = \pi_{t-1}^e + \lambda(\pi_{t-1} - \pi_{t-1}^e)$) and change updating parameter in response to changes in the environment.
- Reaction to changes in macroeconomic parameters will be bigger in high inflation environment because underprediction inflation in unstable environment is more costly than overpredicting it.

Method: learning-to-forecast experiment with 2 treatments: high inflation environment (tested on students from Belarus) and low inflation environment (researched on UVA students)

Literature

Hypothesis 1:

Glensk and Wachtel (1981) adaptive expectations are the best fit to the survey data, heterogeneity in adaptive parameter;

Fajfar and Santoro (2010) – adaptive learning fits the data to the right of the median in the Michigan survey data

Petersen (2014) - adaptive expectations are the best fit to the data.

Hypothesis 2:

Galati et al. (2009) after the Great Recession sensitivity of expectations to macroeconomic information increases

Fajfar and Santoro (2010) agents update information more often when inflation is high and volatile;

Fajfar and Zakelj (2016) higher updating coefficients in treatments with higher inflation variability.

Contribution:

Effects of environment on individual and aggregate expectation formation

The new rule of change in the parameter of the adaptive learning model (asymmetric costs of underpredicting inflation => higher updating parameter in uncertain environment)

Low and high inflation environments

Belarus: high inflation throughout the whole history

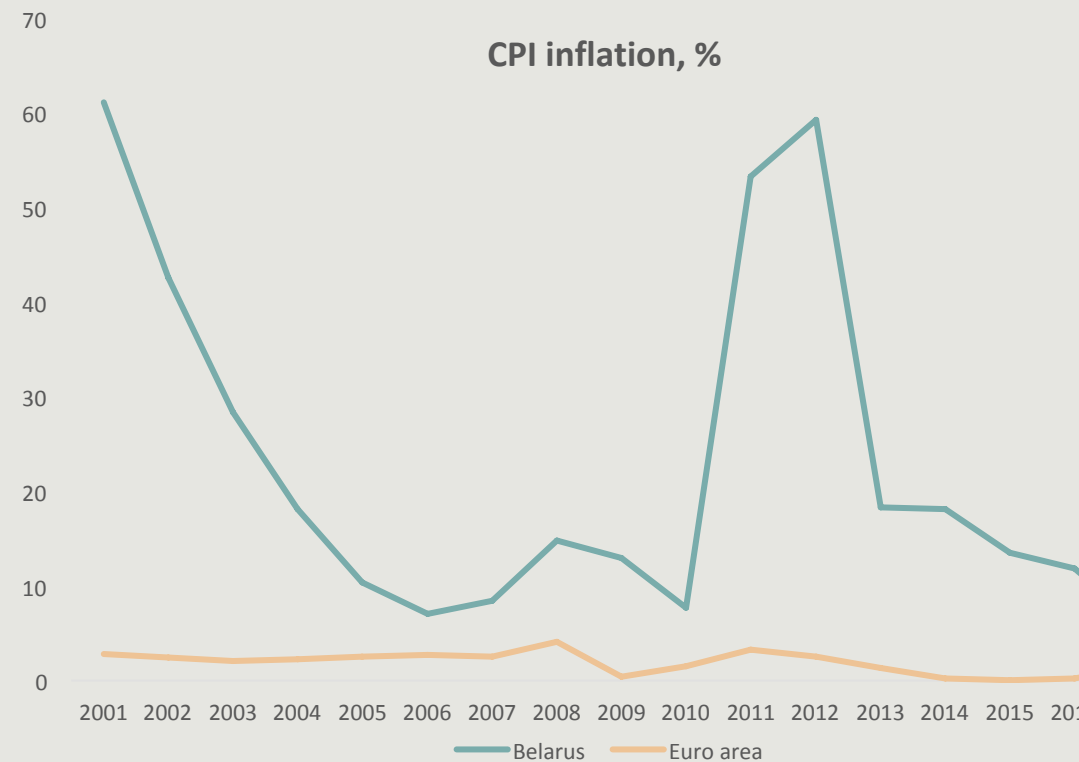


persistent expectations of high inflation currency
devaluation, high inflation

Euro area: low and stable inflation



low inflation expectations



Experimental design

- Groups of 5, each representing a separate experimental economy, constant group composition.
- Task: **forecast inflation** in the next period based on information up to previous period:

“During the next period, what do you expect inflation to be (negative value means decrease in prices, positive value refers to increase in prices and 0 denotes no change in prices)?”

- In period t - forecast inflation in period $t+1$ based on:
 - ✓ Subject's previous forecasts;
 - ✓ Data on inflation, GDP growth and interest rate up to period $t-1$; in period 1 subjects are given an interval of possible values of inflation forecast $[-5;15]$.
- Subjects' forecasts **have a feedback** on the experimental economy.

Experimental timeline

Period 1

Subjects forecast inflation in period 2 without any values of inflation, GDP growth and interest rate (possible values [-5;15])

Inflation, GDP growth and interest rate data for period 1 are generated using forecasts

Period 2

Subjects forecast inflation in period 3 based on the values of inflation, GDP growth, interest rate from period 1.

Inflation, GDP growth and interest rate data for period 2 are generated using forecasts

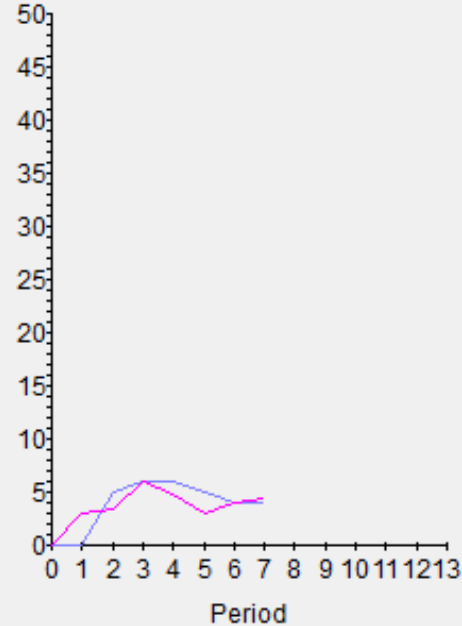
30 periods

Period

8 of 10

Remaining time [sec]: 60

Inflation



— Actual
— Your prediction

During the next period, what do you expect inflation to be (negative value means decrease in prices, positive value refers to increase in prices and 0 denotes no change in prices)?

OK

Period	GDP growth	Inflation	Interest rate	Inflation forecasts	Your profit	Accumulated profit
1	2.7	3.4	6.0	0.00	0.00	0.0
2	4.0	6.1	7.5	5.00	0.00	0.0
3	4.5	5.0	7.5	6.00	29.09	29.1
4	3.9	3.0	6.0	6.00	5.06	34.1
5	3.0	4.0	4.5	5.00	30.27	64.4
6	4.6	4.5	4.5	4.00	48.53	112.9
7	4.0	5.6	6.0	4.00	19.21	132.1

The model

The experimental economy is described by the following model:

$$y_t = E_t^* y_{t+1} - 0.164(i_t - E_t^* \pi_{t+1}) + \varepsilon_t$$

$$\pi_t = 0.3y_t + 0.7E_t^* \pi_{t+1} + u_t$$

$$i_t = 1.5(E_t^* \pi_{t+1} - \pi) + \pi$$

$$\varepsilon_t = 0.6\varepsilon_{t-1} + \varepsilon_t$$

$$u_t = 0.6u_{t-1} + u_t$$

y_t - GDP gap, π_t - inflation, i_t - interest rate, $E_t^* y_{t+1}$ - expected output gap (assumed to be equal to the previous value of the output gap $\Rightarrow E_t^* y_{t+1} = y_{t-1}$), $E_t^* \pi_{t+1}$ - expected inflation, π - inflation target (3%), ε_t - mean-zero demand shock, u_t - mean-zero supply shock. Shocks are assumed to be first-order autoregressive processes with coefficient equal to 0.6.

In period 15 each experimental economy is hit by a supply shock of magnitude 5.

Questionnaire

- general questions (sex, age, country, monthly expenditures, faculty)
- strategy used when forecasting inflation;
- whether they changed strategy during the experiment;
- was central bank successful in stabilizing inflation;
- questions to assess general optimism/pessimism about country's economic situation and personal economic situation.

Payoff

- Payment in cash based on the forecasting accuracy.
- Payoff in each period is found with the following formula (Adam, 2007):

$$points_t = \max\{100 / (1 + error_t) - 20, 0\}$$

- Forecasting accuracy: $error_t = |\pi_t - \pi_t^f|$

Payoff for selected values of forecast error

$error_t$	0	0.5	1	1.5	2	2.5	3	3.5	≥ 4
points	80	46.7	30	20	13.3	8.6	5	2.2	0

- Payoffs in each period were accumulated: $payoff = \sum_{t=3}^{30} points_t$.
- 2 trial periods, so only in the last 28 rounds subjects could earn money
- Exchange rate – 80 points/euro, 65 points/ruble plus 7 euro (5 rubles) participation fee.

Additional information

Sample: 2 groups of 10 people, first - students from the University of Amsterdam second - students from the Belarusian State University.

Sessions: 2 (in Amsterdam and in Belarus), on average 40 minutes.

Instructions: for students from UVA in English, for students from BSU in Russian.

Average payoff: 10 rubles, 14 euro.

Software: Z-TREE.

Data analysis

1. Test experimental data for matching the rational expectations hypothesis:

✓ unbiasedness: $\pi_{jt} = \alpha + \beta \pi_{jt}^e + \varepsilon_{jt}$, if $\alpha=0, \beta=1 \Rightarrow$ unbiased

✓ efficiency:

- strong form: $\pi_{jt} - \pi_{jt}^e = \alpha + \beta \pi_{jt-1} + \gamma y_{jt-1} + \delta i_{jt-1} + u_{jt}$,

- weak form: $\pi_{jt} - \pi_{jt}^e = \alpha + \beta \pi_{jt-1} + u_{jt}$?

2. Fit the data to

- linear prediction rule: $\pi_{j,t+1}^e = c + \sum_{i=0}^2 \alpha_i \pi_{j,t-i}^e + \sum_{i=1}^3 \beta_i \pi_{jt-i} + \sum_{i=1}^3 \gamma_i y_{jt-i} + \mu_{jt}$

- trend extrapolation rule: $\pi_{j,t+1}^e = \beta_0 + \beta_1 \pi_{jt-2} + \beta_2 (\pi_{jt-2} - \pi_{jt-3}) + \mu_{jt}$

- adaptive learning rule: $\pi_{jt}^e = \pi_{jt-1}^e + \lambda (\pi_{jt-1} - \pi_{jt-1}^e)$

3. Compare coefficients for different individuals and between groups, conduct statistical test to find out whether these coefficients are significantly different;

4. If $\alpha_i^{BSU}, \beta_i^{BSU}, \gamma_i^{BSU} > \alpha_i^{UVA}, \beta_i^{UVA}, \gamma_i^{UVA} \Rightarrow$ evidence supporting the 2nd hypothesis

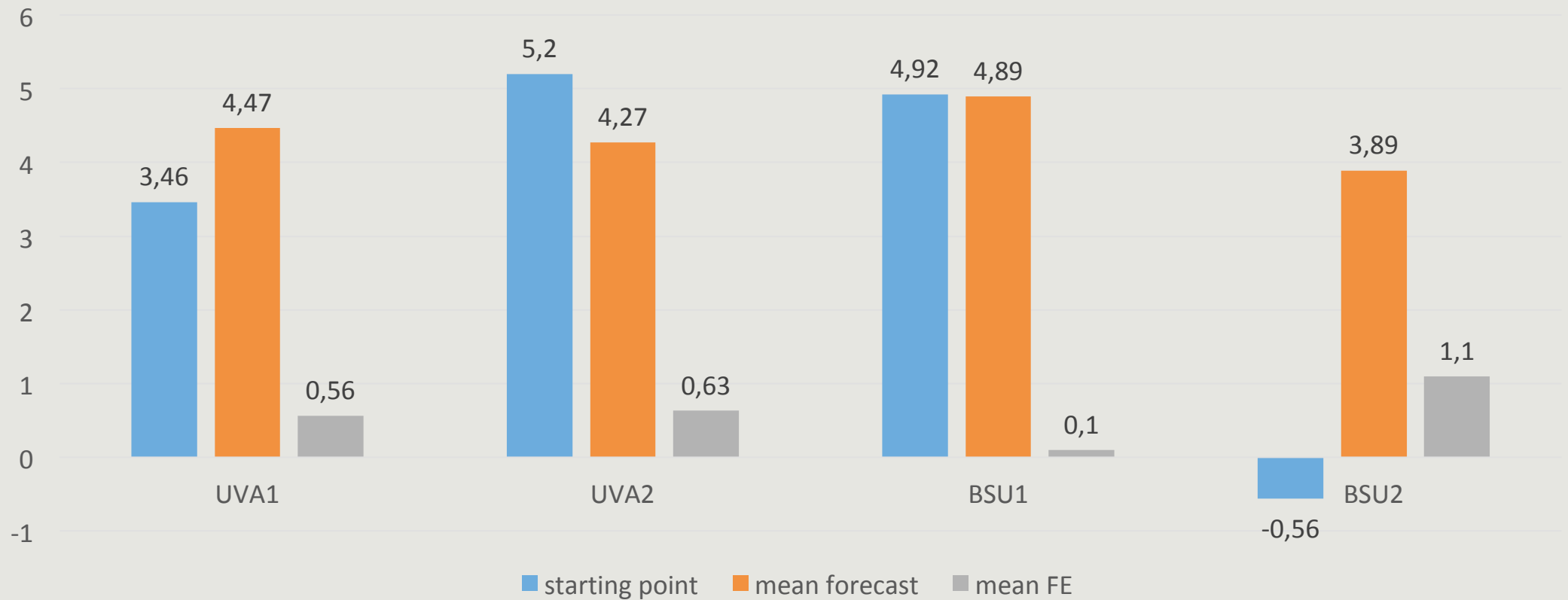
Expected findings

- The majority of the individual data from the experiment was expected to fit adaptive learning rule.
- Agents were expected to be heterogeneous not in terms of forecasting rules but in terms of updating parameter (λ).
- People from high-inflation environment were expected to adapt to changes in inflation faster than people who are used to low-inflation.

Descriptive statistics

Characteristics	BSU	UVA
age	19.8	21.5
Female/male	8/2	6/4
Education	Bachelors of Corporate Finance	Bachelors and Masters: 9 Economics and Business and 1 Political Science
Country	Belarus	2 Netherlands, Albania, USA, Ukraine, Germany, Latvia, Pakistan, Cuba, Sweden
Participation in experiments in the past	No	?

Group results



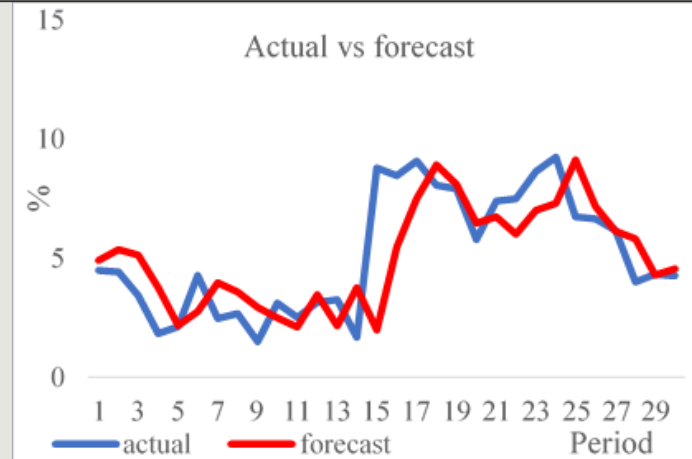
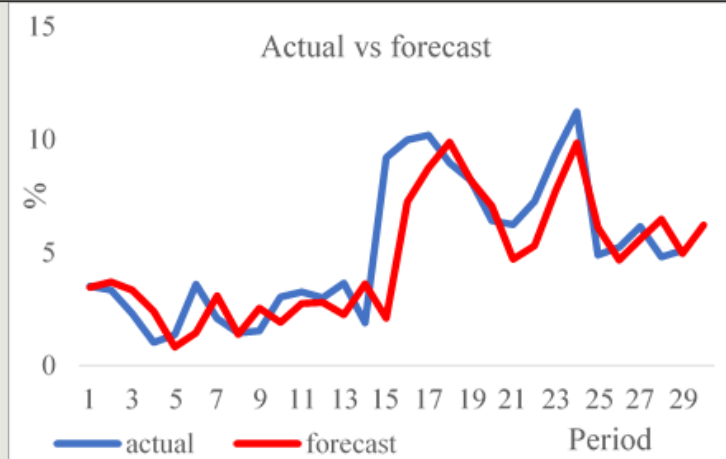
Group results. 2

Group

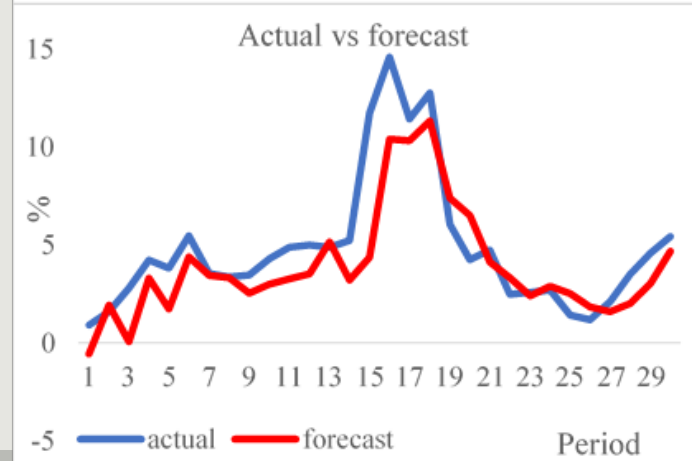
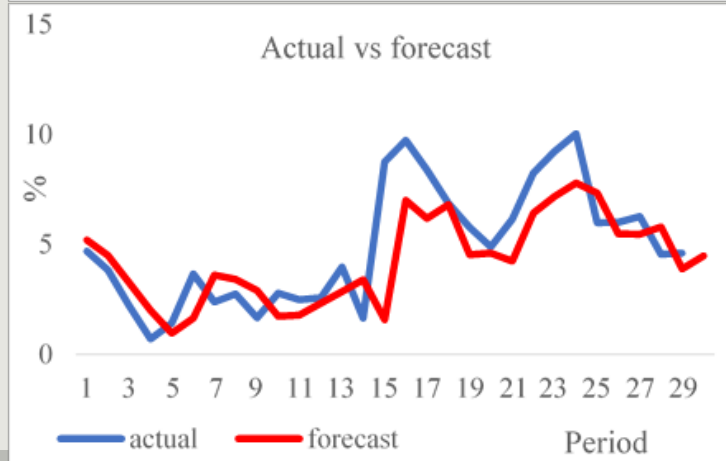
UVA

Belarus

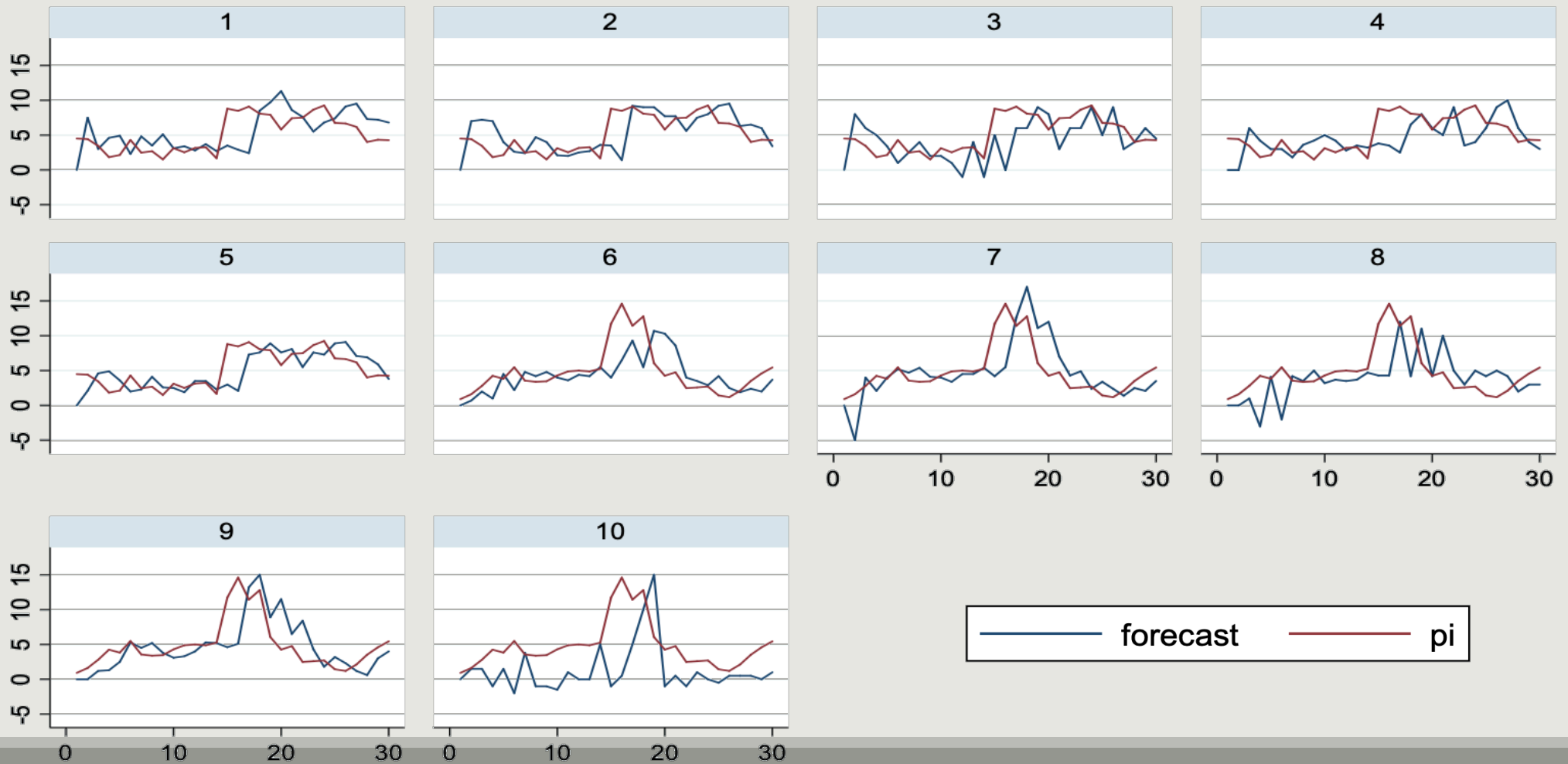
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2







Individual results. BSU



Individual results. UVA



Testing aggregate expectations

- Rational expectations hypothesis 
- Trend extrapolation rule: $\pi_{j,t+1}^e = \beta_0 + \beta_1 \pi_{t-2} + \beta_2 (\pi_{t-2} - \pi_{t-3}) + \mu_t$ 
- Adaptive expectations: $\pi_{t+1}^e = \beta_0 + \beta_1 \pi_{t-1}^e + \beta_2 f_{t-2} + \varepsilon_t$ 
- General linear rule: $\pi_{j,t+1}^e = \beta_0 + \beta_1 \pi_{t-1} + \beta_2 y_{t-1} + \beta_3 i_{t-1} + \mu_t$ 

Testing individual expectations

- Rational expectations hypothesis ✖
- Trend extrapolation rule: $\pi_{j,t+1}^e = \beta_0 + \beta_1 \pi_{j,t-2} + \beta_2 (\pi_{j,t-2} - \pi_{j,t-1}) + \dots$
- Adaptive expectations: $\pi_{j,t}^e = \beta_0 + \beta_1 \pi_{j,t-1}^e + \beta_2 \pi_{j,t-2} + \dots + \varepsilon_{j,t}$
- General linear rule: $\pi_{j,t+1}^e = \beta_0 + \beta_1 \pi_{j,t-1} + \beta_2 y_{j,t-1} + \beta_3 i_{j,t-1} + \mu_{j,t}$
- **After the shock updating parameter increased in BSU treatment but not at the UVA treatment.**



Hypothesis




Hypothesis 2

Conclusions

1. Hypothesis 1:

- ✓ Adaptive expectations fitted the experimental data better than trend extrapolation and general linear forecasting rule 
- ✓ When divided into 2 subsamples, adaptive expectations fitted the data only after the shock 

2. Hypothesis 2:

- ✓ The updating parameter increased after the shock for both Belarus groups but not for the UVA treatment 

Weak points

- small sample (10 people in each session);
- confounding effect of possible previous forecasting experiments of UVA participants;
- different cultural backgrounds of the UVA sample (half of the participants were from countries with relatively unstable inflation);
- limited understanding of such concepts as inflation, GDP growth and interest rate
- sample of only master students (=> younger, with lower income and the share of females higher than on average in the population of Belarus and the Netherlands)

Further research

- larger sample from a balanced demographically subject pool
- ask subjects to forecast changes in price level, not inflation
- research the differences inflation perception in high- and low-inflation environment
- research difference in expectation formation between people who have lived in a low-inflation country for a long time but were brought up in high-inflation environment and people who have always lived in a high-inflation country
- research how the choice of forecasting rules depends on the macroeconomic environment

Thank you for attention

