

# Planar Beauty Contests

Mikhail Anufriev<sup>a</sup>   John Duffy<sup>b</sup>  
Valentyn Panchenko<sup>c</sup>

<sup>a</sup> University of Technology Sydney

<sup>b</sup> University of California Irvine

<sup>c</sup> University of New South Wales, Sydney

CEF, Milan

21 June 2018

# Outline

- 1 Introduction
- 2 Experiment
- 3 Results
  - First Period Choices
  - Dynamics
- 4 Learning Models
- 5 Conclusion

# Planar Beauty Contests

- We present an **experiment** that **extends** the Keynesian beauty contest game to the second dimension
  - Beauty Contest is a “forecasting game”, convenient to study theories of **level- $k$  thinking** and of **adaptive learning**
  - coupled dynamical systems arise naturally as a reduced-form of many macroeconomic models
- We confront behavior of participants in **four treatments** that differ in feedback coefficients and lead to various stability predictions
- We find that not only the location of eigenvalues w.r.t. unit circle but also the **sign of unstable eigenvalue** is important for convergence

# Plan

- 1 Introduction
- 2 Experiment
- 3 Results
  - First Period Choices
  - Dynamics
- 4 Learning Models
- 5 Conclusion

## Experiment: The guessing game

Game between  $N$  players.

Every participant writes down a number from the set

$$S = [0, 100]$$

Let  $x_i$  be the number written down by participant  $i$  and let

$$\bar{x} = \frac{1}{N} \sum_{i=1}^N x_i,$$

be the average of these numbers. The participant whose guess  $x_i$  is closest to the **target** number

$$x^* = \frac{2}{3}\bar{x} + 30$$

wins the game.

# The guessing game: Nash Equilibrium

In the game with target

$$x^* = \frac{2}{3}\bar{x} + 30$$

the **Nash Equilibrium** is “everyone plays 90”.

- it can be computed as the fixed point of the equation
- alternatively, it can be approached via an **iterative elimination of the dominated strategies** (from both sides)
- Beauty contest game is subject to **strategic uncertainty**: playing equilibrium strategy hinges on assumption that all other players play equilibrium strategies as well.
- **k-level thinking theories** assume that players play best response on less sophisticated strategies of other players

# Literature Overview

- **Guessing game**
  - Nagel (AER, 1995); Stahl and Wilson (GEB, 1995); Duffy and Nagel (EJ, 1997); Ho et al (AER, 1998); Bosch-Domenech et al (AER, 2002); Guth et al (EL, 2002); Costa-Gomes and Crawford (AER, 2006); Grosskopf and Nagel (GEB, 2008); Sutan and Willinger (JEDC, 2009); Agranov et al (JESI, 2015)
- **Learning-to-Forecast experiments**
  - Hommes et al (RFS, 2005); Heemeijer et al (JEDC, 2009); Sonnemans and Tuinstra (JEPs, 2010); Bao et al (JEDC, 2012)
- **k-Level thinking theories**
  - Camerer et al (QJE, 2004); de Giorgi, Reimann (GEB, 2008)
- **Adaptive vs. Eductive learning**
  - Camerer and Ho (ETCA, 1999); Bao and Duffy (EER, 2016)

## Why planar Guessing Game?

- It is interesting to see how  $k$ -level reasoning generalizes on the second dimension
- In many macroeconomic models the reduced forms are given by the two-dimensional systems.
- The properties of discovering REE depend on the eigenvalues.
- Learning-to-Forecast experiments in two dimensions focus on adaptive learning, whereas we present and explain the laws of motion to the subjects
  - Assenza et al (WP, 1999)



# Plan

- 1 Introduction
- 2 Experiment**
- 3 Results
  - First Period Choices
  - Dynamics
- 4 Learning Models
- 5 Conclusion

## Basic Design

- $N = 10$  students play game for  $T = 15$  periods
- At each time they have to submit a pair of numbers called in the experiment “A-number” and “B-number”, each number within an interval  $[0, 100]$ .
- With the pairs  $(x_i, y_i)$  the target numbers are given by

$$\begin{pmatrix} x^* \\ y^* \end{pmatrix} = \mathbf{M} \begin{pmatrix} \bar{x} \\ \bar{y} \end{pmatrix} + \mathbf{b} = \begin{pmatrix} m_{11} & 0 \\ m_{21} & m_{22} \end{pmatrix} \begin{pmatrix} \bar{x} \\ \bar{y} \end{pmatrix} + \begin{pmatrix} b_1 \\ b_2 \end{pmatrix},$$

where  $\bar{x}$  and  $\bar{y}$  are the averages of  $x_i$  and  $y_i$

- Matrix  $\mathbf{M}$  and vector  $\mathbf{b}$  depend on the treatment and are known to all participants.
- Reward of participant  $i$  is

$$\frac{500}{5 + |x_i - x^*| + |y_i - y^*|} \quad \text{points}$$

## Equilibrium Prediction

- The game has a **unique Nash Equilibrium**

$$\begin{pmatrix} x_i \\ y_i \end{pmatrix} = (\mathbf{I} - \mathbf{M})^{-1} \mathbf{b} \quad \text{for every } i,$$

- Alternatively, participants may play best responses on assumed level of sophistication of others (**level- $k$  theories**)
- The game is played repeatedly, and players may respond to the previous **history**; each participant can see:
  - past own pairs of numbers for each period
  - past averages of both numbers for each period
  - past target numbers for each period
  - past points earned

# Treatments

Sink

$$\mathbf{M} = \begin{pmatrix} 2/3 & 0 \\ -1/2 & -1/2 \end{pmatrix}, \mathbf{b} = \begin{pmatrix} 30 \\ 75 \end{pmatrix} \rightsquigarrow \begin{pmatrix} x^{NE} \\ y^{NE} \end{pmatrix} = \begin{pmatrix} 90 \\ 20 \end{pmatrix}$$

SaddleNeg

$$\mathbf{M} = \begin{pmatrix} 2/3 & 0 \\ -1/2 & -3/2 \end{pmatrix}, \mathbf{b} = \begin{pmatrix} 30 \\ 95 \end{pmatrix} \rightsquigarrow \begin{pmatrix} x^{NE} \\ y^{NE} \end{pmatrix} = \begin{pmatrix} 90 \\ 20 \end{pmatrix}$$

SaddlePos

$$\mathbf{M} = \begin{pmatrix} 2/3 & 0 \\ -1/2 & 3/2 \end{pmatrix}, \mathbf{b} = \begin{pmatrix} 30 \\ 35 \end{pmatrix} \rightsquigarrow \begin{pmatrix} x^{NE} \\ y^{NE} \end{pmatrix} = \begin{pmatrix} 90 \\ 20 \end{pmatrix}$$

Source

$$\mathbf{M} = \begin{pmatrix} 3/2 & 0 \\ -1/2 & -3/2 \end{pmatrix}, \mathbf{b} = \begin{pmatrix} -45 \\ 95 \end{pmatrix} \rightsquigarrow \begin{pmatrix} x^{NE} \\ y^{NE} \end{pmatrix} = \begin{pmatrix} 90 \\ 20 \end{pmatrix}$$

# Plan

- 1 Introduction
- 2 Experiment
- 3 Results**
  - First Period Choices
  - Dynamics
- 4 Learning Models
- 5 Conclusion

# Plan

- 1 Introduction
- 2 Experiment
- 3 Results
  - First Period Choices
  - Dynamics
- 4 Learning Models
- 5 Conclusion

## First Period A-Choices

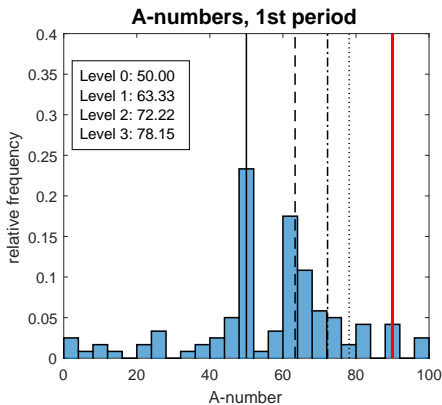
- The first (“A”) number is motivated by guessing

$$x^* = p\bar{x} + b_1,$$

- it is identical to the standard (1D) beauty contest game with  $p = 2/3$  and  $b_1 = 30$  ([Sink](#), [SaddleNeg](#), [SaddlePos](#)) and  $p = 3/2$  and  $b_1 = -45$  ([Source](#))
- levels of rationality (for  $p = 2/3$  and  $b_1 = 30$ ):
  - 0:  $x_0 = 50$  (average of the interval)
  - 1:  $x_1 = px_0 + b_0 = 63.33$
  - 2:  $x_2 = px_1 + b_0 = 72.22$
- levels of rationality (for  $p = 3/2$  and  $b_1 = -45$ ):
  - 0:  $x_0 = 50$  (average of the interval)
  - 1:  $x_1 = px_0 + b_0 = 30$
  - 2:  $x_2 = px_1 + b_0 = 0$

# First Period A-Choices

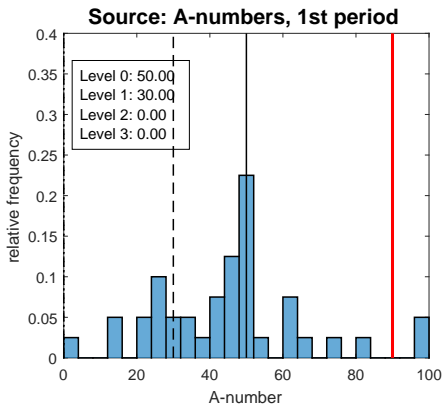
$$x^* = \frac{2}{3}\bar{x} + 30$$





# First Period A-Choices: Source Treatment

$$x^* = \frac{3}{2}\bar{x} - 45$$



## First Period B-Choices

- Target of the second (“B”) number depends on both “A” and “B” numbers

$$y^* = m_{21}\bar{x} + m_{22}\bar{y} + b_2,$$

- For example in **Sink** we have

$$y^* = -\frac{1}{2}\bar{x} - \frac{1}{2}\bar{y} + 75,$$

- levels of rationality:

0:  $y_0 = 50$  (average of the interval)

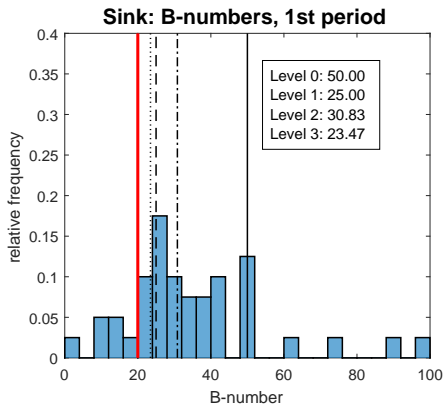
1:  $y_1 = m_{21}x_0 + m_{22}y_0 + b_2 = 25$

2:  $y_2 = m_{21}x_1 + m_{22}y_1 + b_2 = 30.83$

3:  $y_3 = m_{21}x_2 + m_{22}y_2 + b_2 = 23.47$

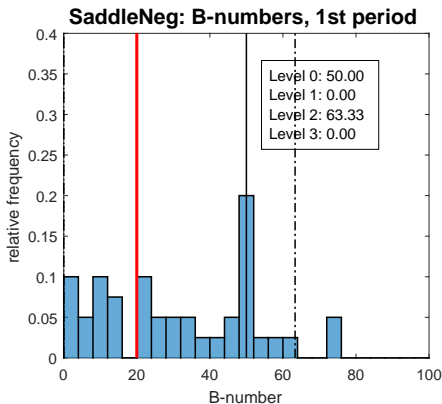
# First Period B-Choices: Sink

$$y^* = -\frac{1}{2}\bar{x} - \frac{1}{2}\bar{y} + 75$$



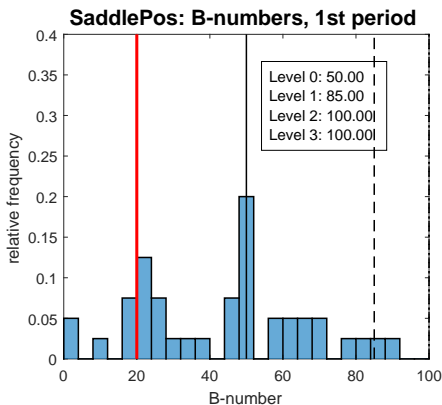
# First Period B-Choices: SaddleNeg Treatment

$$y^* = -\frac{1}{2}\bar{x} - \frac{3}{2}\bar{y} + 95$$



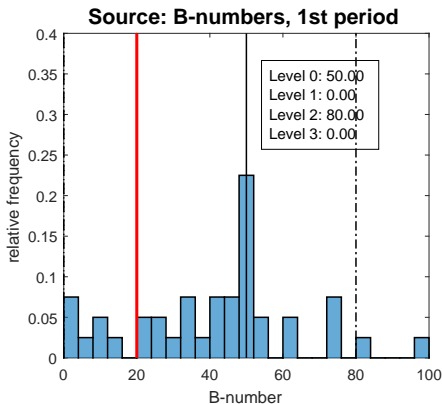
# First Period B-Choices: SaddlePos Treatment

$$y^* = -\frac{1}{2}\bar{x} + \frac{3}{2}\bar{y} + 35$$



# First Period B-Choices: Source Treatment

$$y^* = -\frac{1}{2}\bar{x} - \frac{3}{2}\bar{y} + 95$$



## First Period Choices

- First period choices are heterogeneous. There is a spike around the middle of the interval (0-level of rationality)
- For the **A-number** there is another spike near the 1st level of rationality and tendency of subjects to move towards the 2nd level of rationality
- For the **B-number** there is a tendency to move towards the 1st level of rationality

Treatment	x (A-number)			y (B-number)		
	Mean $x_1$	Median $x_1$	Std.Dev $x_1$	Mean $y_1$	Median $y_1$	Std.Dev $y_1$
<b>Sink</b>	57.39	60	17.67	34.98	30	20.01
<b>SaddleNeg</b>	55.01	56.01	22.28	31.06	29	21.46
<b>SaddlePos</b>	56.25	58.80	20.50	42.97	46.75	23.18
<b>Source</b>	44.78	45.75	20.97	40.92	45	23.21

## First period choices

- level  $k$  theories assume that subjects are **boundedly rational** stopping at some level of thinking
- in the Beauty Contest game, it is typical to find  $k = 1$  and  $k = 2$ , but rarely higher than 3.

### Take Away Messages:

- 1 for **A-number**: the Beauty Contest game with the NE located inside the interval seems to be more difficult to subjects and leads to a decrease in the level of rationality
- 2 for **B-number**: the presence of a coupled variable in the Beauty Contest game leads to even further decrease in the level of rationality



# Plan

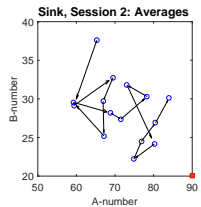
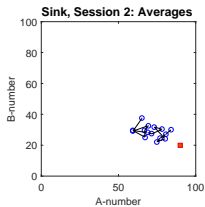
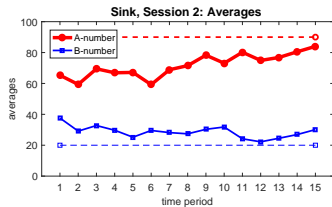
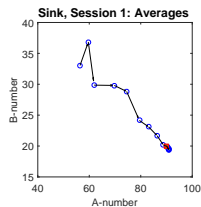
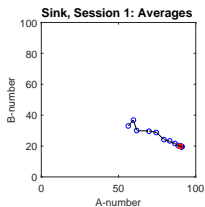
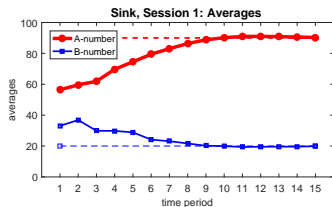
- 1 Introduction
- 2 Experiment
- 3 Results
  - First Period Choices
  - Dynamics
- 4 Learning Models
- 5 Conclusion

## Dynamics: from the first period to the last

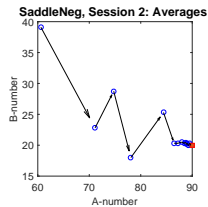
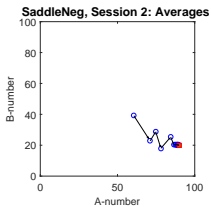
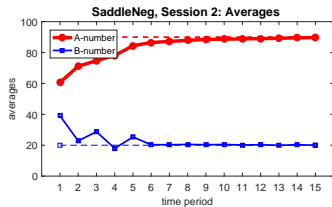
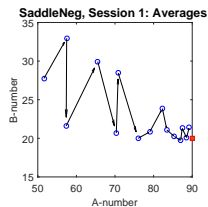
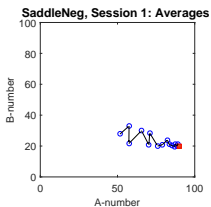
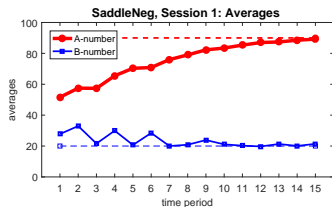
Treatment	x (A-number)			y (B-number)		
	Mean $x_1$	Median $x_1$	Std.Dev $x_1$	Mean $y_1$	Median $y_1$	Std.Dev $y_1$
<b>Sink</b>	57.39	60	17.67	34.98	30	20.01
<b>SaddleNeg</b>	55.01	56.01	22.28	31.06	29	21.46
<b>SaddlePos</b>	56.25	58.80	20.50	42.97	46.75	23.18
<b>Source</b>	44.78	45.75	20.97	40.92	45	23.21

Treatment	x (A-number)			y (B-number)		
	Mean $x_{15}$	Median $x_{15}$	Std.Dev $x_{15}$	Mean $y_{15}$	Median $y_{15}$	Std.Dev $y_{15}$
<b>Sink</b>	84.21	86.86	14.40	25.40	21.48	17.44
<b>SaddleNeg</b>	89.81	89.93	0.55	20.31	20	1.49
<b>SaddlePos</b>	87.33	88	3.06	92.31	100	24.95
<b>Source</b>	7.46	0	13.49	35.56	35	6.03

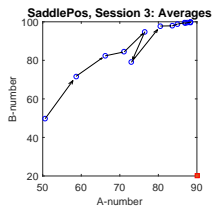
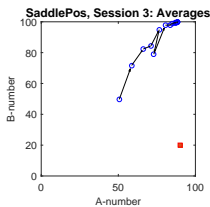
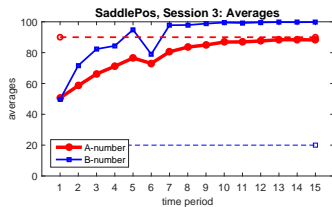
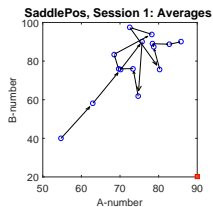
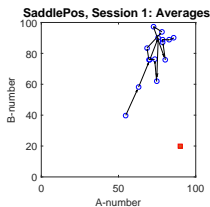
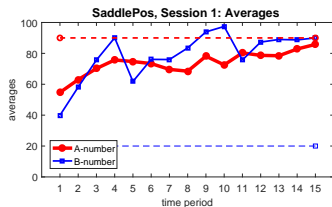
## Sink T (Averages, sessions 1 and 2)



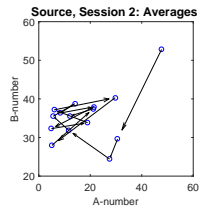
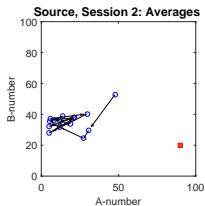
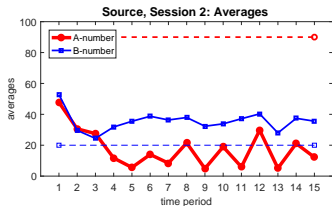
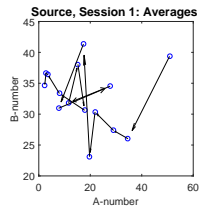
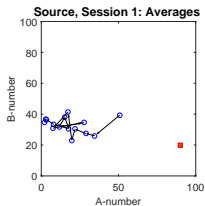
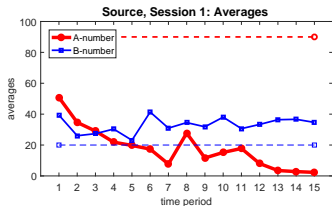
# SaddleNeg T (Averages, sessions 1 and 2)



# SaddlePos T (Averages, sessions 1 and 3)



# Source T (Averages, sessions 1 and 2)



# Summary of Results on Dynamics

**Dynamics converge only in Sink and SaddleNeg**

**Dynamics of  $x$  ...** (identical to the beauty contest game)

- ... converges almost monotonically in Sink, SaddleNeg and SaddlePos.
- ... diverges in the Source treatment to the lower bound 0.

**Dynamics of  $y$  ...**

- ... converges almost monotonically in Sink and converges (through oscillations) in SaddleNeg.
- ... does not converge to the equilibrium value in SaddlePos and Source.
- ... for the Source the dynamics is stable, near a constant level inside the range.

# Plan

- 1 Introduction
- 2 Experiment
- 3 Results
  - First Period Choices
  - Dynamics
- 4 Learning Models**
- 5 Conclusion



## Model Comparison

We compared the following models

- 1 NE
- 2 Naive
- 3 Adaptive models with estimated  $\lambda_x, \lambda_y$
- 4 Average with 2, 3, 4, 5 and 14 lags
- 5 Heuristic Switching Model (Anufriev and Hommes, 2012)

using the **AIC**, i.e., the sum of the MSEs for  $x$  and  $y$  adjusted for the number of parameters.

For every session, at each step  $t = 3, \dots, 15$  we compute **model predictions**  $x^{e, \text{Mod}}$  and  $y^{e, \text{Mod}}$  to find  $\text{MSE}_x + \text{MSE}_y$

$$\text{MSE}_x = \frac{1}{13} (\bar{x}_t - x^{e, \text{Mod}})^2 \quad \text{and} \quad \text{MSE}_y$$

# Winners of the Model Comparison

MSE's by treatments and in total:

Treatment	NE	Average		Adaptive		HSM	
		Naive	MAve(4)	$\lambda_x \neq \lambda_y$ , loc	$\lambda_x = \lambda_y$ , glob	local	glob
<b>Sink</b>	953	161	128	118	126	<b>104</b>	124
<b>SaddleNeg</b>	420	60	79	<b>22</b>	42	35	30
<b>SaddlePos</b>	19642	848	723	<b>321</b>	338	330	355
<b>Source</b>	24490	1239	958	<b>432</b>	476	<b>432</b>	499
<b>Total MSE</b>	45506	2309	1888	<b>893</b>	983	901	1007
<b>AIC</b>	2231	1611	1569	1477	<b>1435</b>	1479	1440

Using AIC to penalize different number of parameters, we find a support to **Adaptive model** with common  $\lambda = 0.44$ .

# Plan

- 1 Introduction
- 2 Experiment
- 3 Results
  - First Period Choices
  - Dynamics
- 4 Learning Models
- 5 Conclusion

## Conclusion

- the first experiment of planar Beauty Contest game
- the simplest possible structure (variable  $x$  is decoupled)
- four treatments with various possibilities for eigenvalues
- more complicated 1D equation and presence of extra variable for another equation results in lower level of rationality
- convergence is achieved in the **Sink** treatment and in **Saddle** treatment but only with **negative feedback**
- in **Saddle with positive feedback** and **Source** dynamics diverge
- adaptive model describes the data best