

# Beliefs, Opinion Dynamics and Extremism

André C. R. Martins

NISC – EACH – Universidade de São Paulo – Brasil

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# Opinion Dynamics Models

- Different models are used for different problems.
  - Discrete Opinions: few options, can represent decisions well.
  - Continuous Opinions: more useful to represent choices of numerical values.
  
- And decisions that have a strength of opinion attached?

# Continuous Opinions and Discrete Actions

- Definition: Opinion means the subjective probability that a certain statement is right.
- Opinions are not observable, but discrete actions or choices are.
- The observable actions are a function of the internal opinion.
- Agents update their opinions based on the observation of other agents choices by using Bayesian rules.

# CODA model

- Local consensus emerges, but not global!
- Opinions get stronger and islands of similar behavior appear.
- Barriers to changing choices become increasingly stronger and freeze islands eventually.
- Extremists appear everywhere.

# Structure of the Presentation:

1 Theoretical Framework

2 Models for extremism

# A Theoretical Framework for Update Rules.

- Define an opinion as the subjective probability on the debated issue.
- Bayes Theorem can provide rules for changing the opinion.

# Details

- Assign a variable  $x$  to the issue (continuous or discrete? what range? one issue or a cultural problem, with several dimensions?).
- Each agent makes inferences about  $x$ .
- Each agent  $i$  needs to have a subjective opinion about  $x$ , represented by a probability distribution  $f_i(x)$ .
- The function indicates agent  $i$  belief on how likely each possible value of  $x$  is.
- $f_i(x)$  can be continuous or discrete, depending on the characteristics of  $x$ .

# Details

## Communicating

- Communication depends on the agent opinion  $f_i(x)$ : a functional  $A[f_i]$ .
- Important: Communication does not need to be intentional, it could be a an observed behavior of  $j$ .
- The agents must have in their minds a relationship between the true value of  $x$ ,  $x^*$ , and the stated value  $A_j$ , given by a likelihood distribution  $p(A_j|x^*)$ .
- That is, given that a possible value  $x^*$  were the correct value, how likely it would be that the neighbor  $j$  would communicate  $A_j$ .



# Details

## Updating

- The probability distribution  $p(A_j|x)$  plays the role of a likelihood of the observation  $A_j$  and thus defines a Bayesian update rule.
- Agent  $i$  already had a prior opinion  $f_i(x)$ , obtaining its posterior opinion  $f_i(x|A_j)$  is a simple task of applying Bayes Theorem.

# Structure of the Presentation:

1 Theoretical Framework

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# Bounded Confidence or CODA

- Bounded Confidence: continuous opinion over range 0 to 1. Tendency to moderate opinions.
- CODA: Internal probability (0 to 1), observed choice (A or B). Tendency to extremism.

# What drives extremism?

- One obvious difference: communication - discrete versus continuous
- One subtle difference: world model - choosing sides versus mixing choices

# What drives extremism?

	<b>Certainty wishers</b>	<b>Mixers</b>
<b>Discrete observation</b>	CODA	New model 2
<b>Continuous observation</b>	New model 1	Bounded Confidence (BC)

# Wishers with Continuous observations (New model 1)

- Estimate  $p$  - probability that  $A$  is best choice.
- Agent observes opinion  $p_j$  of agent  $j$ .
- Likelihood:  $Be(p_j|\alpha, \beta, A) = \frac{1}{B(\alpha, \beta)} p_j^{\alpha-1} (1 - p_j)^{\beta-1}$ .

# Huge increase in extremism

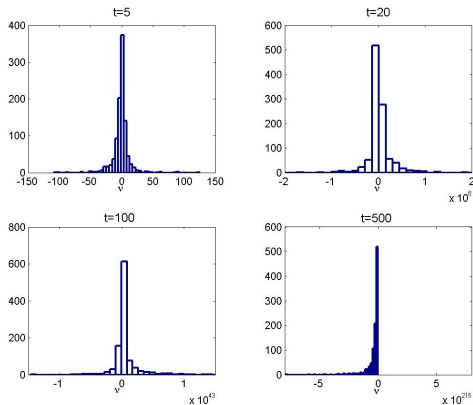


Figura : Log-odd opinions  $\nu_i$  after  $t$  interactions per agent.

## Mixers with Discrete Observations (New Model 2)

- Estimate  $f$  - optimal proportion for  $A$ .
- Requires a continuous probability distribution over  $0 \leq f \leq 1$ .
- Agent observes only if  $j$  thinks there should be more  $A$  or more  $B$ .
- Likelihood: Binomial distribution.



# Comparing

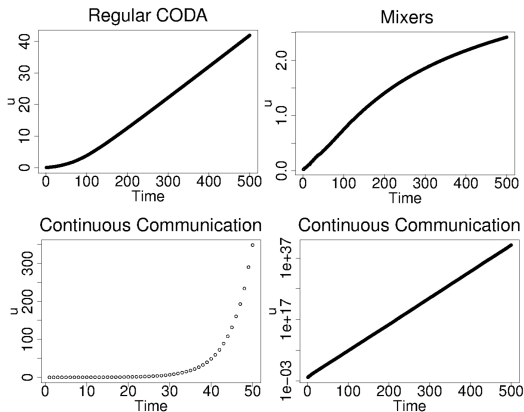


Figura : Strength of the opinion ( $\nu = \ln\left(\frac{p}{1-p}\right)$ ) as a function of time.



# Thank you!