

How **assimilative** and **idiosyncratic**
attitude **change**
generate
oscillations of
public policy mood

Jan Lorenz


Measuring, Modeling and Mitigating Opinion
Polarization and Political Cleavage (MMM)
2023-09-13 @ ETH Zurich, Chair of Systems Design




Jan Lorenz

Assistant Professor
Social Data Science

janlo.de

 datasci.social/@janlo

 twitter.com/jalorenz

[constructor.
university](https://constructor.university)

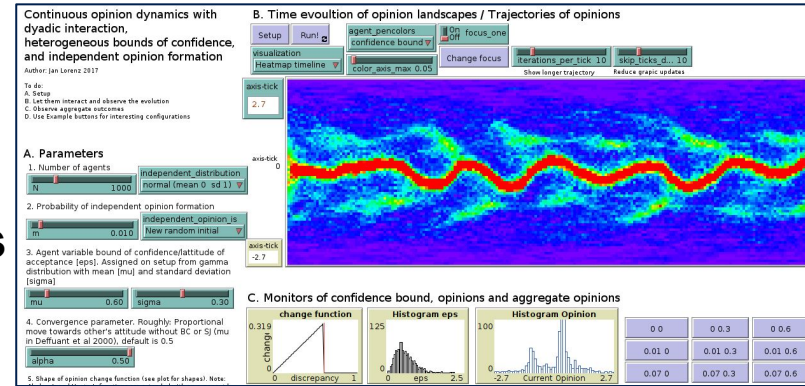
Outline

1. ~~Measuring~~ Defining polarization (for the sake of this talk)
2. Agent-based model of opinion dynamics
 - a. under bounded confidence
 - b. with idiosyncrasy/turn-over/noise
 - c. with heterogeneous bounds of confidence
3. Map it to real data

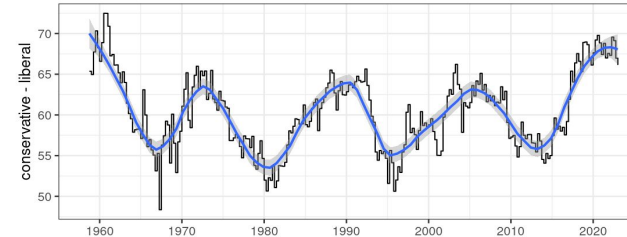
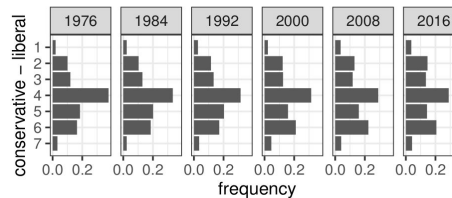
Terminology for this talk:

opinion = attitude

bound of confidence = latitude of acceptance



Liberal-Conservative Self-Identification (ANES)



Blue line from ggplot2::stat_smooth with loess method and span=0.2
Data from Slimson (2019)

This Talk within the Scope of this Workshop

General Definition: **Polarization is the Accentuation of Differences**

This talk is about

- **unidimensional**
- polarization of **attitudes**
- ranging **continuously** between extremes over moderate and neutral
- of the **general population**

Ideological alignment over multiple issues.

Issue partisanship (difference between groups / party supporters)

Polarization in income, employment, or ethnicity.

Yes-No issues

Affective polarization (emotion towards other political groups)

Polarization in **elites, party, or media**

not about

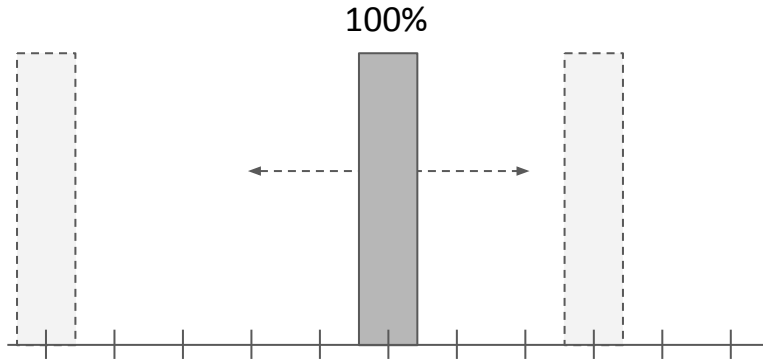
not about

Measuring one-dimensional polarization

Polarization is a measure of a distribution. On a **bounded scale** there is agreement what is minimally and maximally polarized.

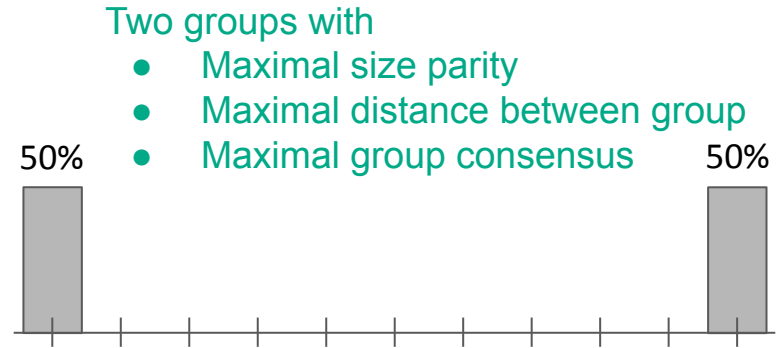
Minimum (Polarization = 0)

- All have the same opinion
- Does not matter where



Maximum (Polarization = 1)

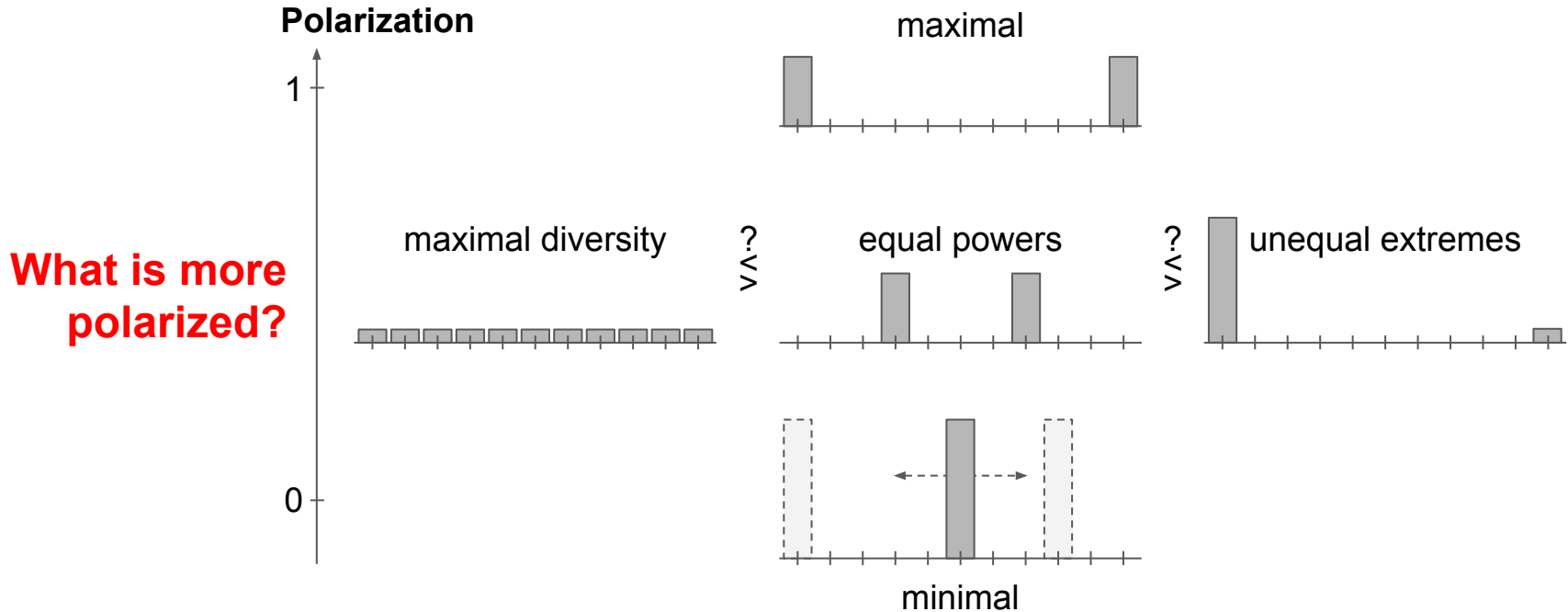
- One half maximally positive
- Other half maximally negative



Bramson, A., Grim, P., Singer, D. J., Fisher, S., Berger, W., Sack, G., & Flocken, C. (2016). Disambiguation of social polarization concepts and measures. *The Journal of Mathematical Sociology*, 40(2), 80–111. <https://doi.org/10.1080/0022250x.2016.1147443>

But what is moderate polarization?

The conceptual measurement problem:



The core measure: Average Pair Discrepancy / Antagonism

For n opinion $0, 1, 2, \dots, n$ and populations $p_0, p_1, p_2, \dots, p_n$ (summing up to one).

$$\text{Pol}_0(p) = \frac{2}{n} \sum_{i,j=0}^n p_i p_j |i - j|$$

Normalization such that polarization is bounded by one

Probability of a pair with opinion i and j

Discrepancy / Antagonism between i and j

Note: Without normalization called *mean absolute difference*.

Gestefeld, M., Lorenz, J., Henschel, N. T., & Boehnke, K. (2022). Decomposing attitude distributions to characterize attitude polarization in Europe. *SN Social Sciences*, 2(7), 110. <https://doi.org/10.1007/s43545-022-00342-7>

Extension: Identification-weighted Polarization

$$\text{Pol}_1(p) = \frac{4}{n} \sum_{i,j=0}^n p_i^2 p_j |i - j|$$

We use a definition implying this is not much polarized.

antagonism weight more when a person has more peers

least polarized
maximal diversity



$\text{Pol}_0 = 0.727$
 $\text{Pol}_1 = 0.132$

Most polarized
equal powers



$\text{Pol}_0 = \text{Pol}_1 = 0.4$

2nd most
unequal extremes



$\text{Pol}_0 = \text{Pol}_1 = 0.36$

With Pol_0 and Pol_1 we can differentiate peaked distribution from uniform

Esteban, J. M., & Ray, D. (1994). On the measurement of polarization. *Econometrica: Journal of the Econometric Society*, 819-851.

Micro Opinion Dynamics I: Assimilation under Bounded Confidence

Bounded Confidence Model *Deffuant et al. style*

The Micro-Interaction:

Agent i hears the opinion of (randomly selected) agent j and updates opinion as

$$x_i(t+1) = \begin{cases} x_i(t) + 0.5(x_j(t) - x_i(t)) & \text{if } |x_i(t) - x_j(t)| < \lambda \\ x_i(t) & \text{otherwise} \end{cases}$$

Parameter λ :
Bound of Confidence
(or latitude of acceptance)

Agent i assimilates in the direction of j 's opinion when close enough.

Deffuant, G., Neau, D., Amblard, Frédéric., & Weisbuch, G. (2000). Mixing Beliefs among Interacting Agents. *Advances in Complex Systems*, 3, 87–98.
<https://doi.org/10.1142/S0219525900000078>

Micro Opinion Dynamics II: Idiosyncratic Opinion Change

Bounded Confidence Model *Deffuant et al. style* with *Pineda et al. style “noise”*

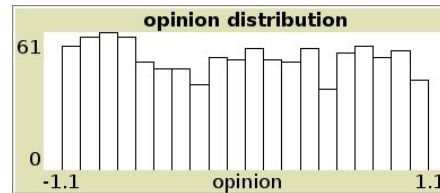
The Micro-Action:

Sometimes (with **idiosyncrasy probability θ**) agent i selects a **random opinion idiosyncratically** from the distribution of initial opinions.

Parameter θ



Initial and idiosyncratic opinions can come from uniform or normal distribution



Pineda, M., Toral, R., & Hernandez-Garcia, E. (2009). Noisy continuous-opinion dynamics. *Journal of Statistical Mechanics: Theory and Experiment*, 2009(08), P08001 (18pp). <http://stacks.iop.org/1742-5468/2009/P08001>

Micro Opinion Dynamics III: Heterogeneous Bounds Confidence

Bounded Confidence Model *Deffuant et al. style*

The Micro-Interaction:

Agent i hears the opinion of (randomly selected) agent j and updates opinion as

$$x_i(t+1) = \begin{cases} x_i(t) + 0.5(x_j(t) - x_i(t)) & \text{if } |x_i(t) - x_j(t)| < \lambda_i \\ x_i(t) & \text{otherwise} \end{cases}$$

Parameter λ become static agent variable λ_i
Drawn from **Gamma distribution** with mean μ_λ and standard deviation σ_λ (two new distribution parameters)

Agent i assimilates in the direction of j 's opinion when close enough.

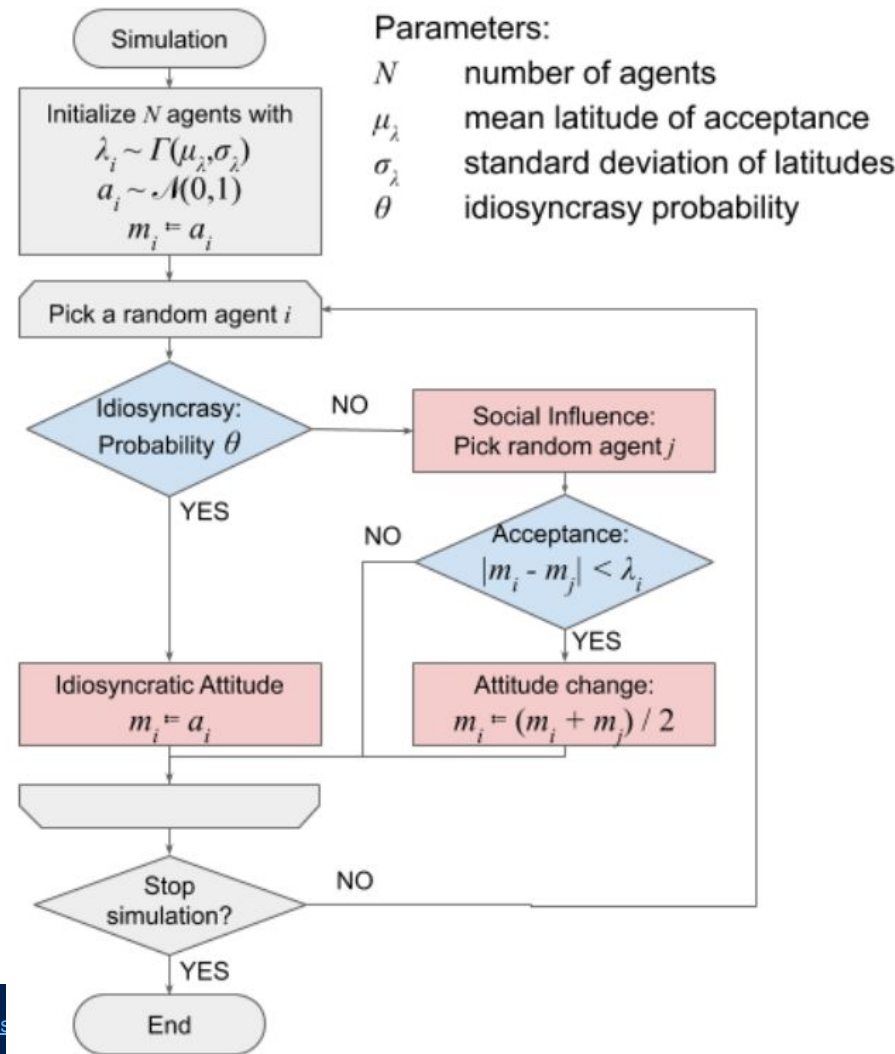
Deffuant, G., Neau, D., Amblard, Frédéric., & Weisbuch, G. (2000). Mixing Beliefs among Interacting Agents. *Advances in Complex Systems*, 3, 87–98.
<https://doi.org/10.1142/S0219525900000078>

From Micro to Macro Opinion Dynamics: Society model with Computer Simulation

- Initialize some 1,000 agents with idiosyncratic opinions
- Repeatedly, sweep over all agents and let them update their opinion through
 - listening to others (randomly) or
 - (occasionally) changing idiosyncratically

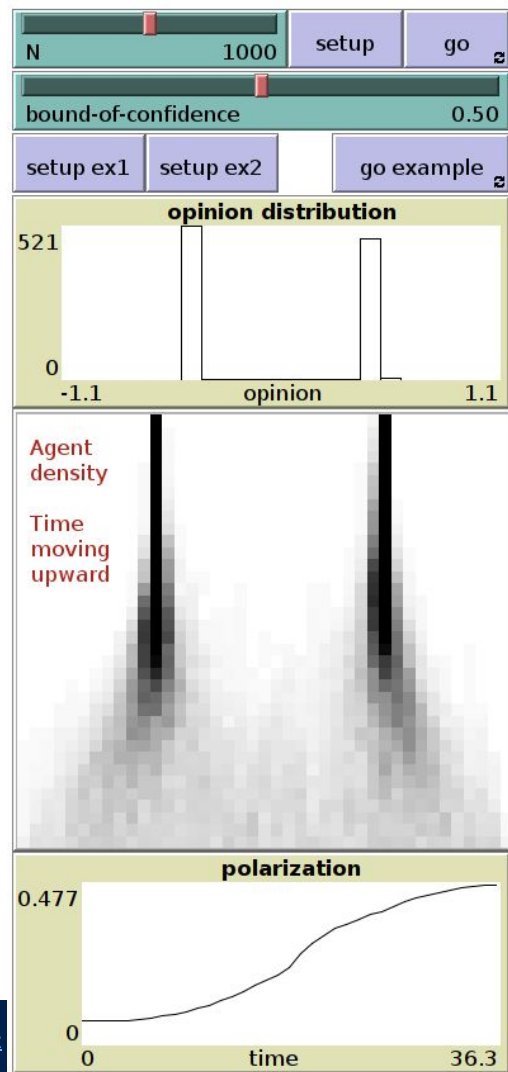
Specification of a society model motivated from social psychology:

Lorenz, J., Neumann, M., & Schröder, T. (2021). Individual attitude change and societal dynamics: Computational experiments with psychological theories. *Psychological Review*, 128(4), 623–642. <https://doi.org/10.1037/rev0000291>

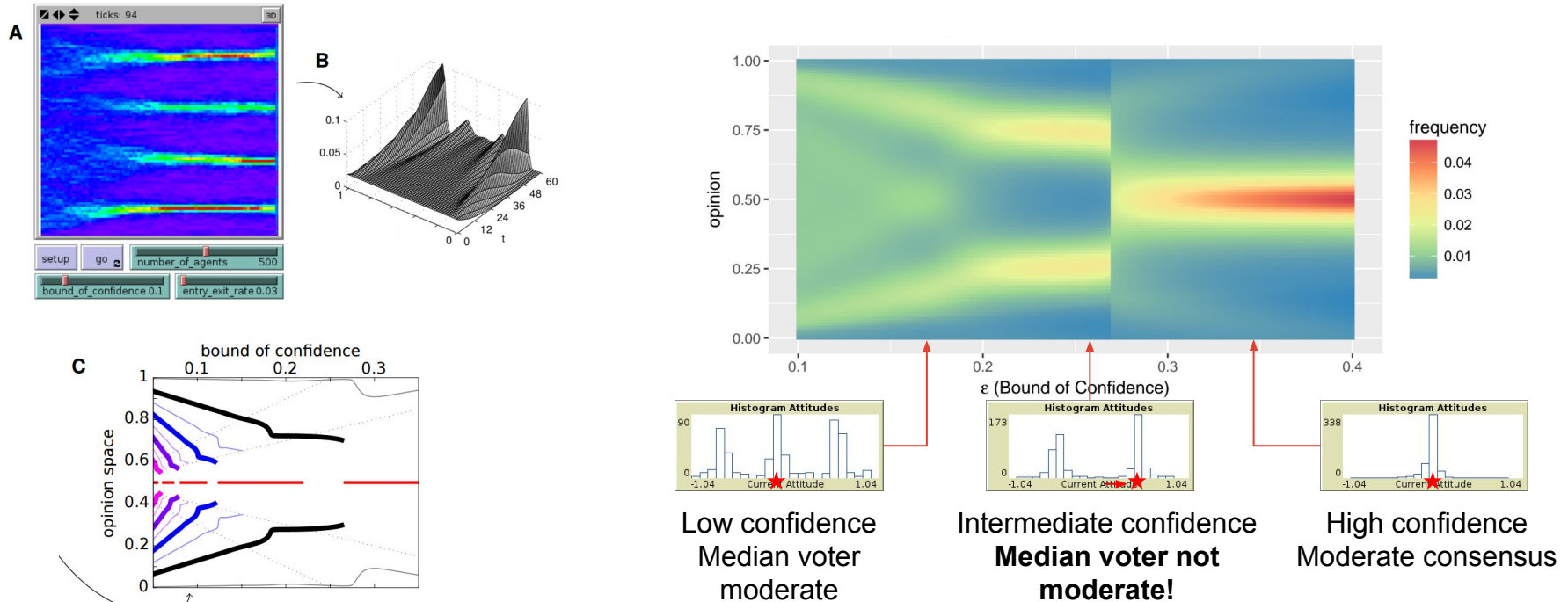


Macro Opinion Dynamics I: Assimilation under Bounded Confidence can trigger Polarization

Polarization in some extreme (and unrealistic) form of two peaks.

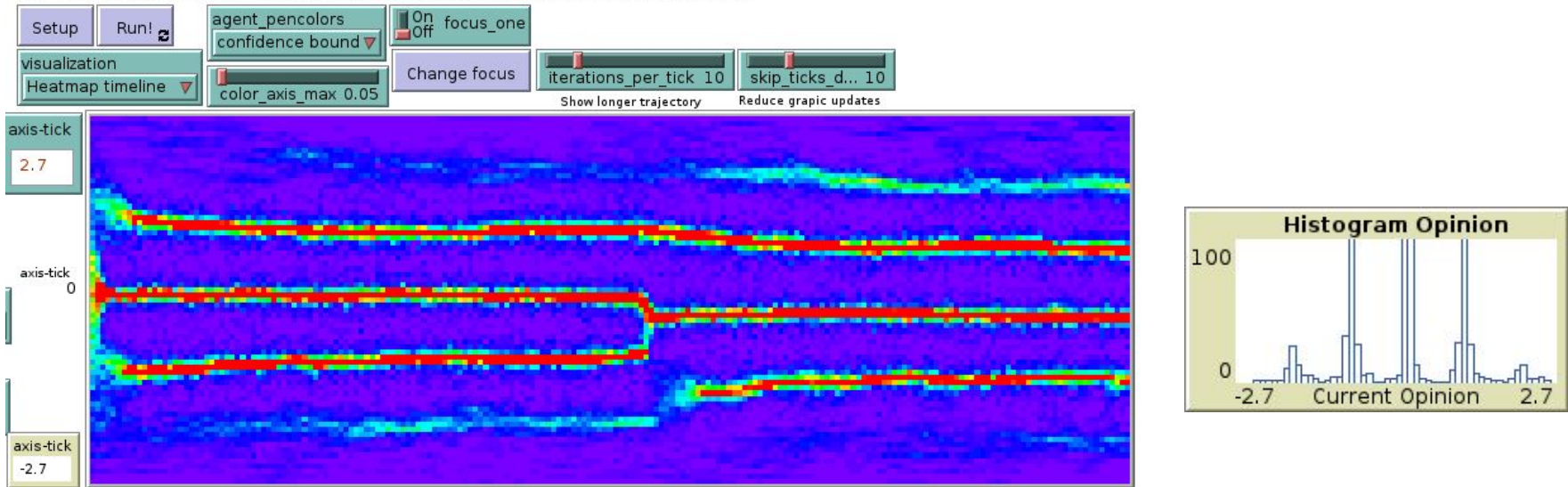


Macro Opinion Dynamics II.a: With idiosyncrasy stochastic stable distributions evolve



Macro Opinion Dynamics II.b: Low fraction (1%) of idiosyncratic opinion formation with normal distribution of opinions

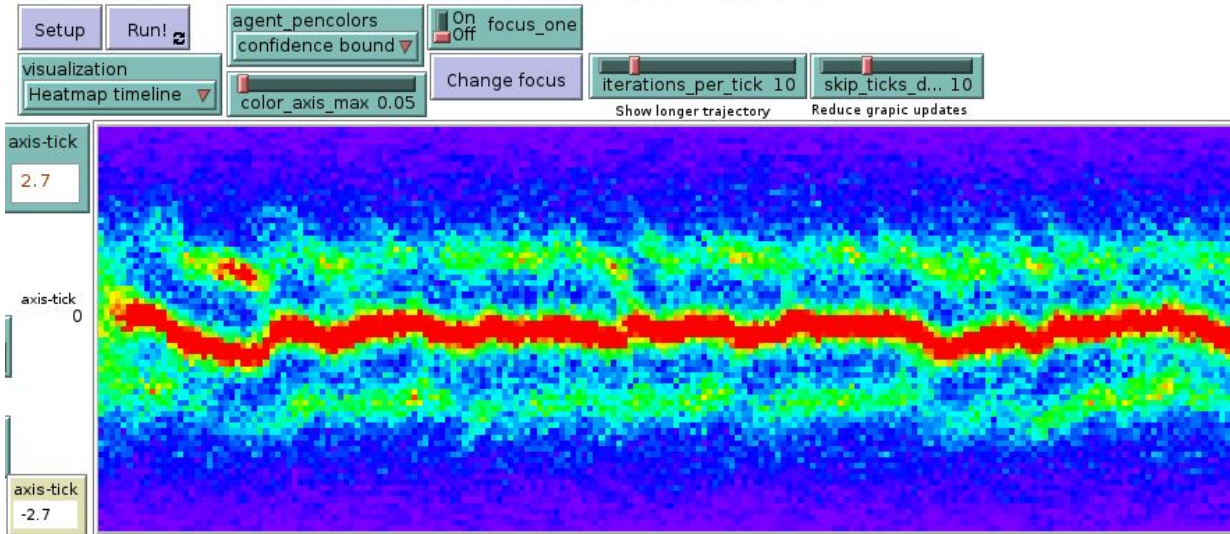
B. Time evolution of opinion landscapes / Trajectories of opinions



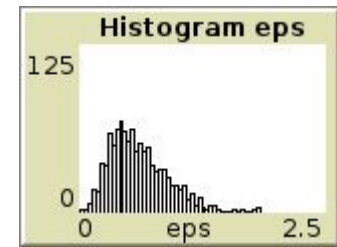
Macro Opinion Dynamics II.b:

Higher fraction (7%) of idiosyncratic opinion formation with normal distribution of opinions

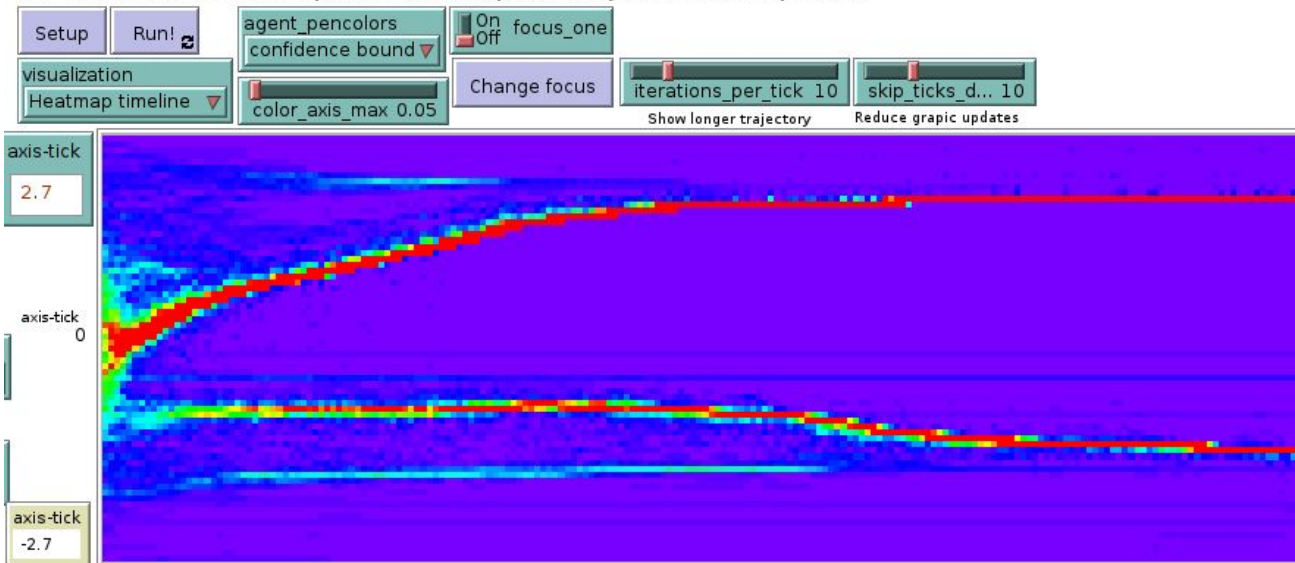
B. Time evolution of opinion landscapes / Trajectories of opinions



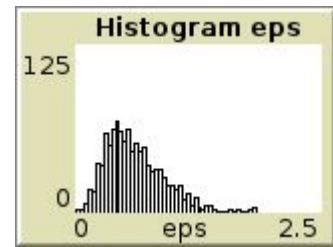
Macro Opinion Dynamics III.a: Heterogeneous bounds of confidence → Drifts of a majority cluster



B. Time evolution of opinion landscapes / Trajectories of opinions



Macro Opinion Dynamics III: Heterogeneous bounds of confidence → Drifts of a majority cluster



B. Time evolution of opinion landscapes / Trajectories of opinions

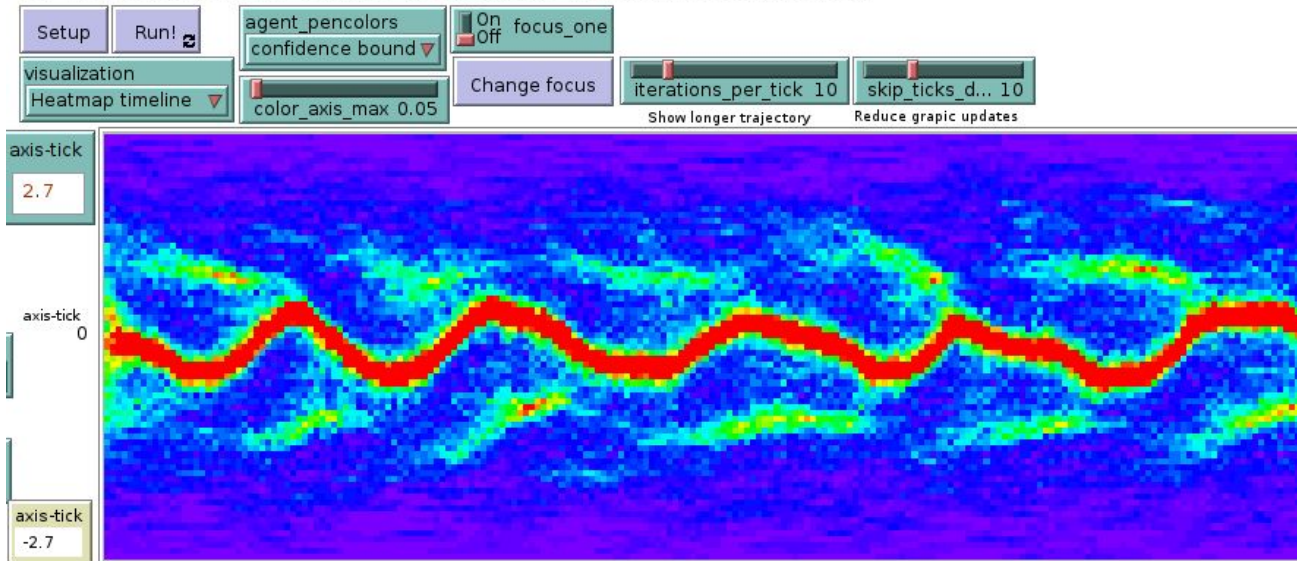


Same simulation but
different visualization with
individuals trajectories

Macro Opinion Dynamics IV: Idiosyncrasy and heterogeneous bounds → Regular oscillation

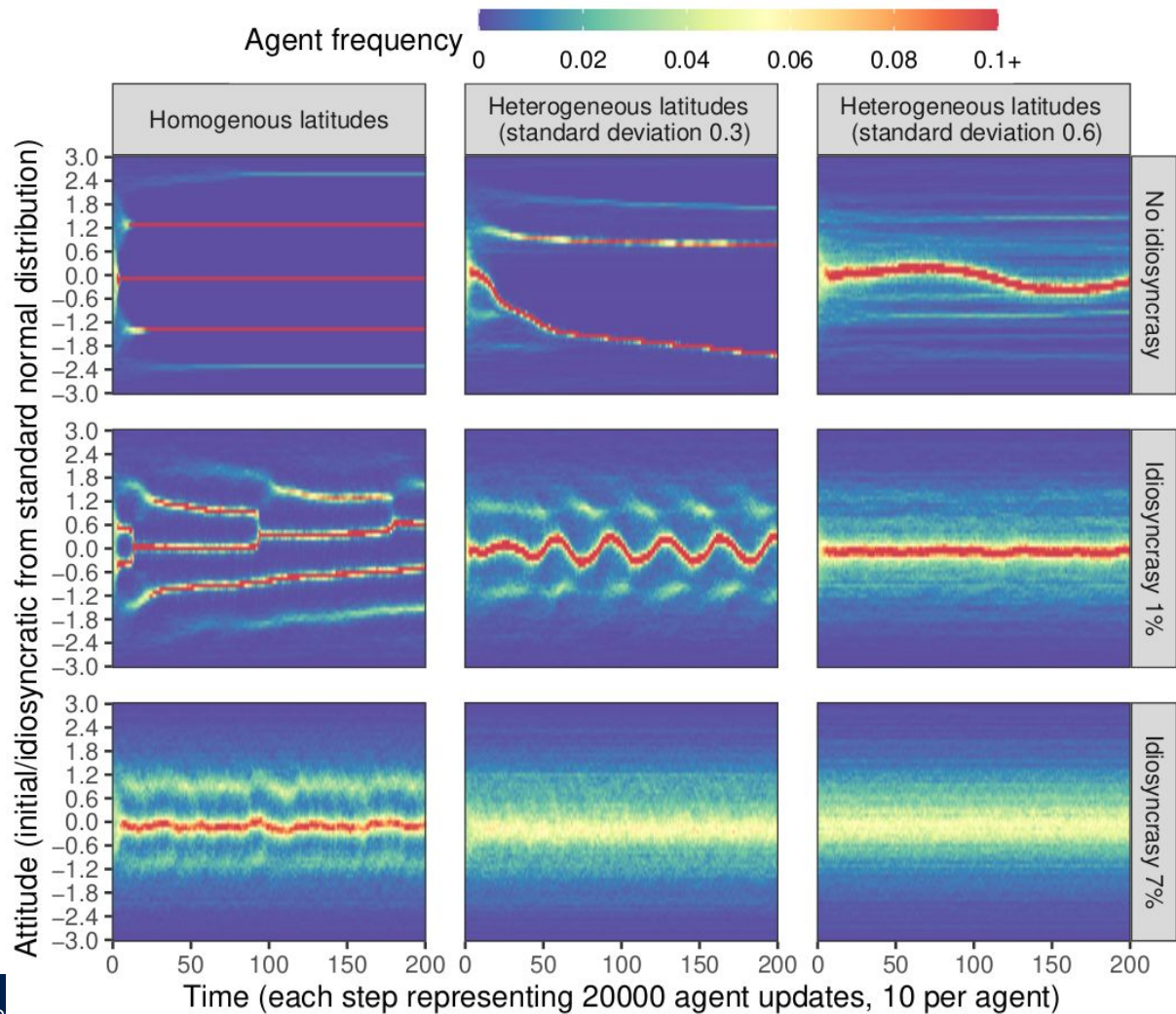
Macro Mechanism

B. Time evolution of opinion landscapes / Trajectories of opinions



- Open-minded form majority cluster in center
- Majority cluster drifts to one side where there are a bit more closed-minded
- Emptier space on the other side lets a small cluster evolve through come-in from turn-over
- At some point enough open-minded emerge in between to trigger a new drift

Oscillations happen with intermediate heterogeneity and intermediate idiosyncrasy



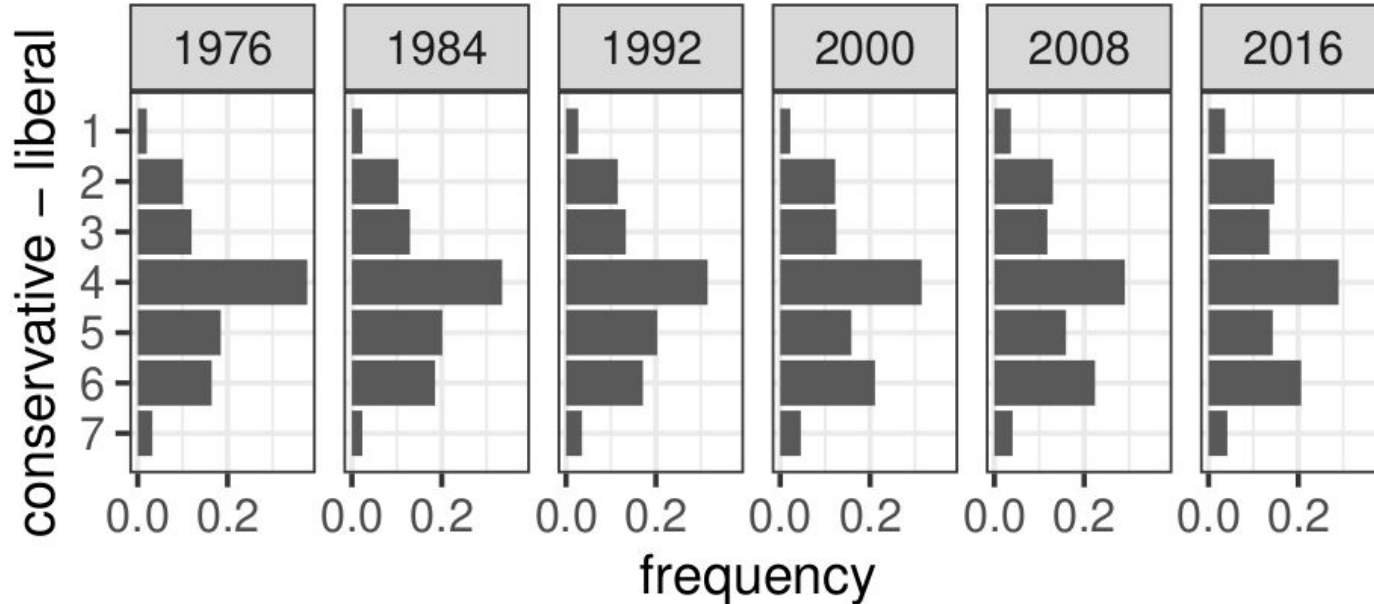
Do Polarization and Oscillations map to real political opinions?

Look at USA

- **Ideology**
(ANES 7-point Likert scale for very conservative to very liberal)
- **Public Policy Mood**
(Stimson's data)

Political Ideology

Liberal–Conservative Self–Identification (ANES)



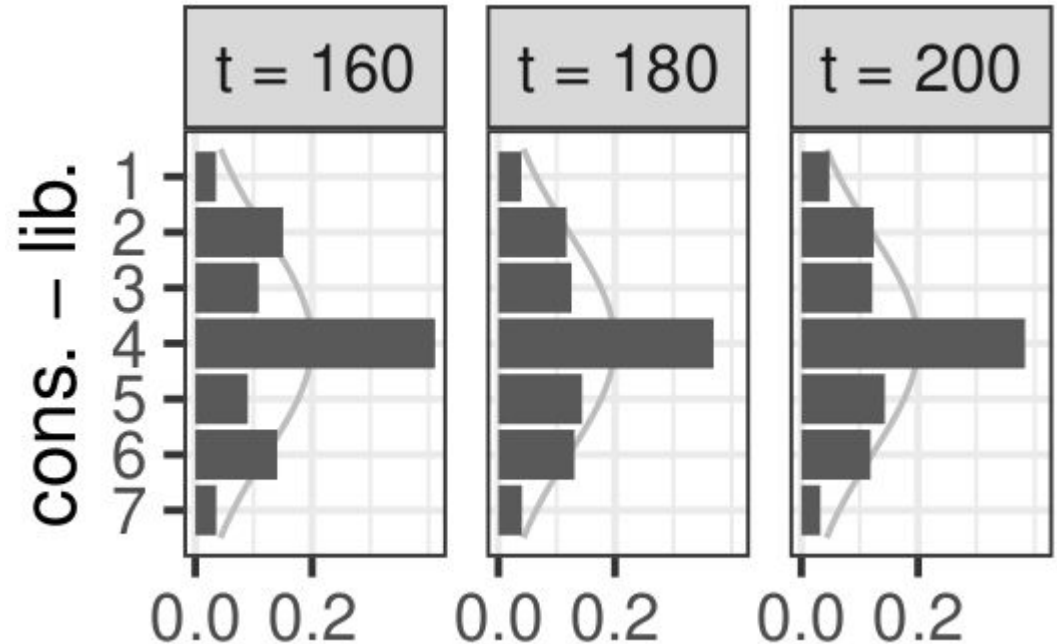
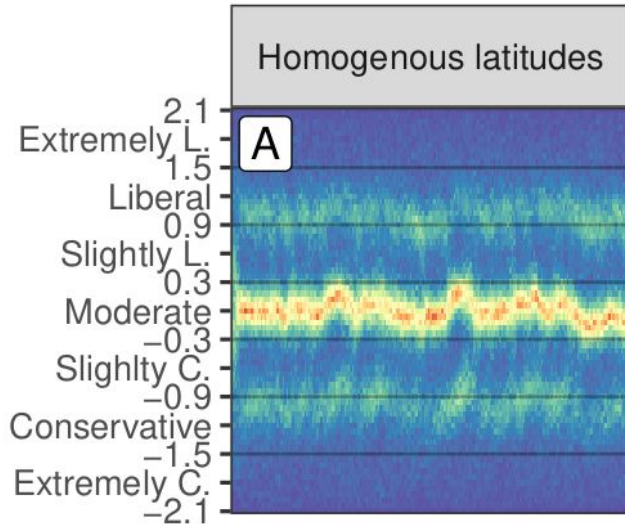
ANES = American National Elections Studies

Ideology: We can come close

How: No heterogeneity, high idiosyncrasy

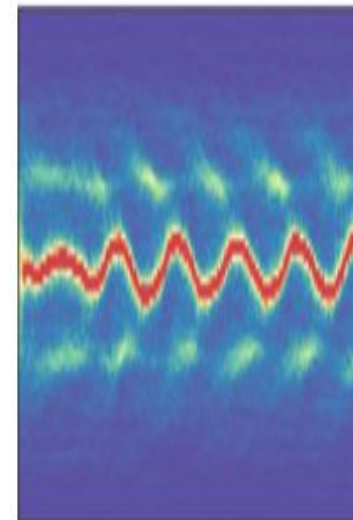
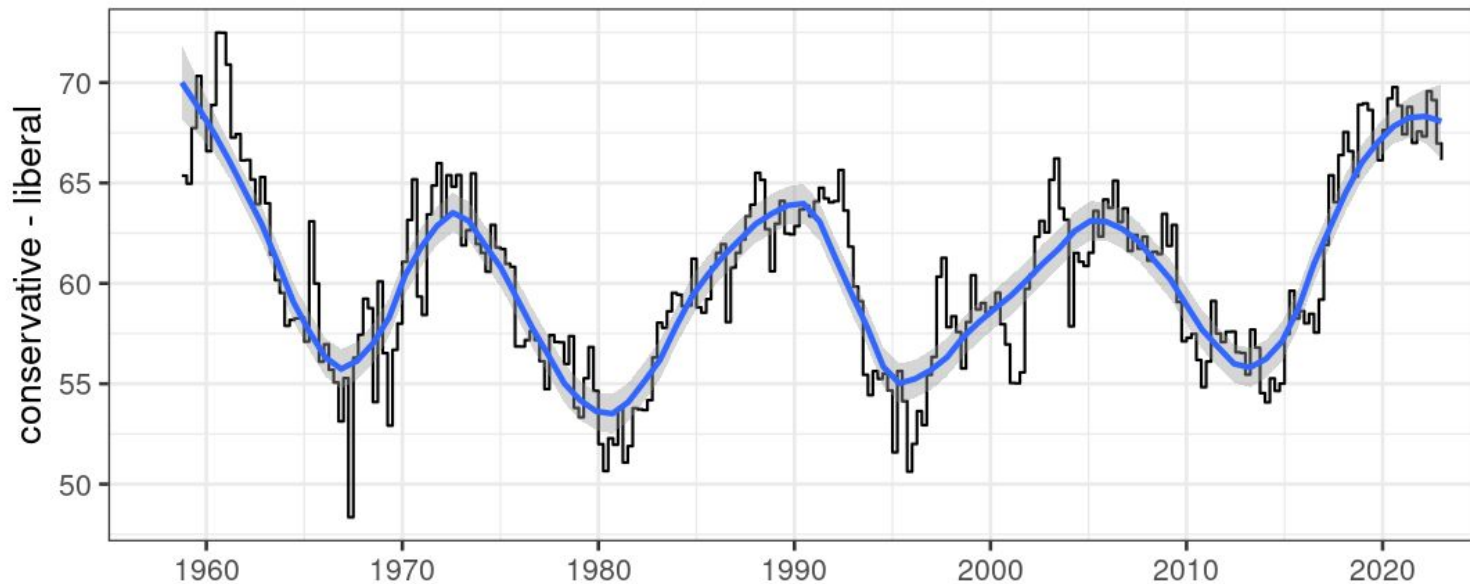
Mean latitude 0.6, standard deviation of latitudes 0

Idiosyncrasy probability 7%



Public Policy Mood: Regular oscillations

How: Intermediate Heterogeneity



Mean latitude 0.6
Standard deviation of latitudes 0.3
Idiosyncrasy probability 1%

Approximate wavelength:
14.5 years \leftrightarrow 22.4 agent updates a year

Blue line from `ggplot2::stat_smooth` with `loess` method and `span=0.2`
Data from Stimson (2019)

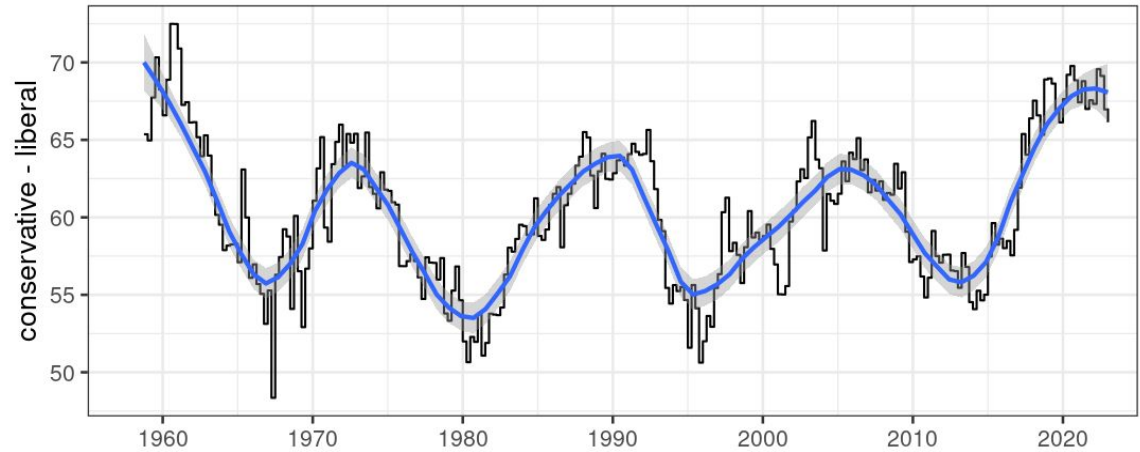
Note: No mood data for individuals available.

Stimson, James. <https://stimson.web.unc.edu/data/> Public Policy Mood

Takeaways

- **Mechanism 1:**
Assimilation under Bounded Confidence as a driving force of polarization!
 - Intrinsic radicalization or repulsive forces are not a necessary mechanism
- **Mechanism 2:**
Idiosyncrasy in opinion formation creates stochastically stable opinion distribution closer to empirical distributions of ideology
- **Mechanism 3:**
Heterogeneous Bounds creates drifts of the a majority cluster
- **Mechanism 4:**
Heterogeneous bounds and idiosyncrasy creates regular oscillations similar to public policy mood

Backup: Mood swings more and differently than Ideology



Mean Ideology

