# Network Preference Dynamics using Lattice Theory

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### Introduction

### Key contributions:

- Preference lattice
  - Algebraic model for preference: meet (∧) and join (∨) operations
  - Median aggregation: majority rule for lattices
- Message passing preference dynamics
  - Decentralized networked consensus-like algorithm on preference lattices
  - Prove existence/computability of equilibria in a few cases
  - Software implementation of our algorithm



### Problem Formulation

### Assumptions.

- Agents have preference ordering over a number of alternatives (i.e. options)
- Agents' preference relations are incomplete (missing comparisons)
  - limited information available
  - all pairs of alternatives
- Agents are organized in a network
  - modeled by an undirected graph
  - can share all preference relations with neighboring agents

**Problem 1.** Design a decentralized mechanisms for nodes to share, receive and aggregate preferences in order to augment their independent decision-making capabilities.

Problem 2. Analyze and compute equilibria of the above dynamical systems.



### Related Work

Preference lattice on complete preference (Curello & Sinander 2024)

- Based on single-crossing dominance
- Applications in economics (monotone comparative statics, risk-aversion,

Laplacian consensus with lattice theory priors (Ghrist & Riess, 2022)

- Tarski Laplacian (Ghrist & Riess, 2022)
- Synchronization (Riess, Munger, Zavlanos; 2023)

#### Consensus on lattices

Aggregation functions on lattices (Karacal & Mesiar, 2017)

#### Distributed consensus

- Min/max consensus (Tahbaz-Salehi & Jadbabaie, 2006)
- ▶ Biased min consensus (Zhang & Li, 2017)
- Minmax consensus (Shi, Xia, Johansson, 2015)

Opinion dynamics (DeGroot 1974; Hegselman & Krause, 2022; etc.)



# Lattice Theory

**Definition.** A *poset* is a set with a binary relation  $(\mathcal{P}, \leq)$  satisfying the following properties:

- ▶  $a \leq b$  and  $b \leq c$  implies  $a \leq c$  (transitivity)
- ►  $a \leq a$  (reflexivity)
- ▶  $a \leq b$  and  $b \leq a$  implies a = b (anti-symmetry)

**Definition.** A *lattice* is a poset  $(\mathcal{L}, \leq)$  such that

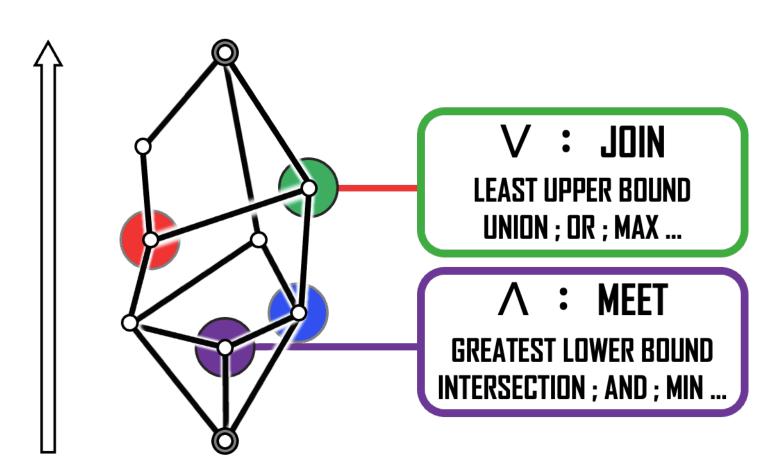
- $a \lor b = \min\{c : c \succeq a, c \succeq b\}$  (called *join*)
- →  $a \land b = \max\{c : c \leq a, c \leq b\}$  (called *meet*) exist for all  $a, b, \in \mathcal{L}$ .

Write, 
$$a_1 \lor a_2 \lor \cdots \lor a_n = \bigvee a_i$$

$$a_1 \land a_2 \land \cdots \land a_n = \bigwedge a_i$$

### Examples.

- $\mathbb{R}^n$  or  $\mathbb{Z}^n$ ,  $\wedge$ : min,  $\vee$ : max
- Boolean algebras: ∧:&,∨:|
- Powersets,  $\wedge : \cap, \vee : \cup$
- Subspaces, subgroups, ideals, etc...
- Convex subsets of a convex set
- Equivalence relations
- Preference relations
- Formal Concept Analysis (FCA)





### Preference Relations

#### Setup:

- $\mathcal{N} = \{1, 2, ..., N\}$ , set of agents
- $\mathcal{A} = \{1, 2, ..., M\}$ , set of alternatives (i.e. options)

**Definition.** A *preference relation*  $\gtrsim$  is a binary relation (subset or predicate) satisfying the following:

- ▶  $\forall a, a', b \in \mathcal{A}$ ,  $(a \gtrsim a') \& (a' \gtrsim b) \rightarrow (a \gtrsim b)$  implies  $i \gtrsim_u k$  (transitivity);
- $\forall a \in \mathcal{A}, \quad a \geq a \text{ (reflexivity)};$

We write  $a \gtrsim_i b$  if Agent  $i \in \mathcal{N}$  weakly prefers Option  $a \in \mathcal{A}$  over Option  $b \in \mathcal{A}$ .

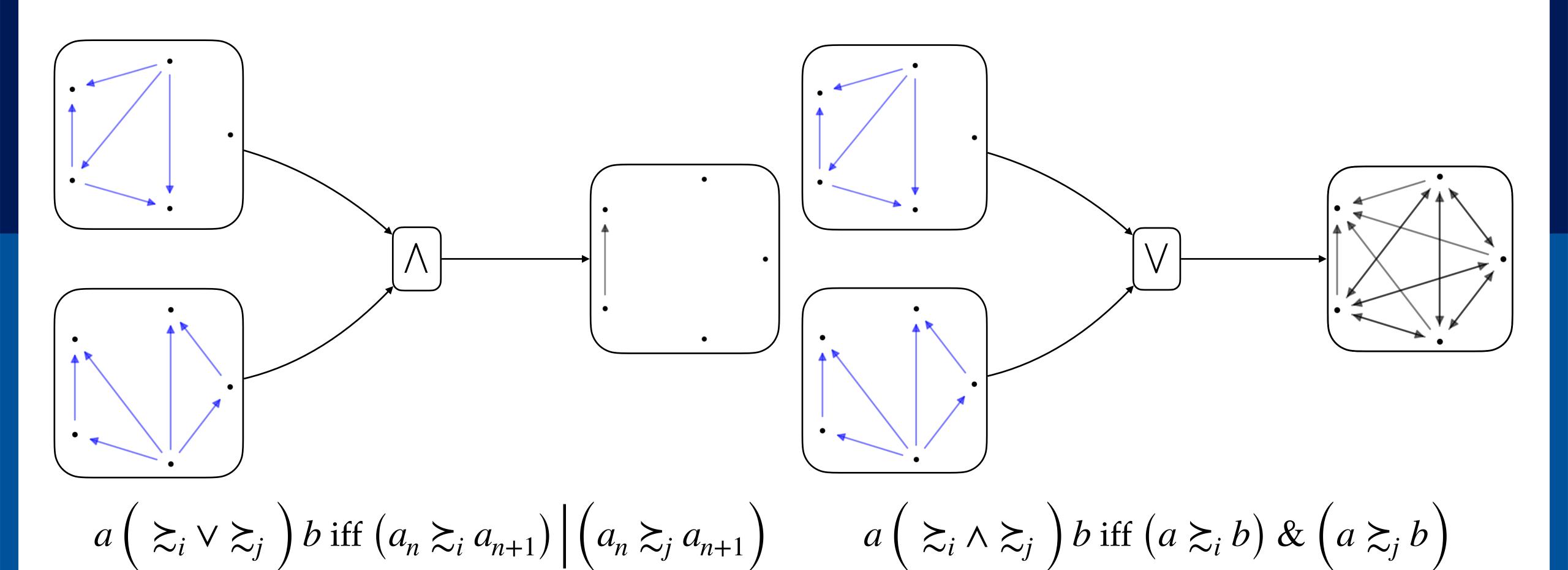
A preference profile is a tuple  $\Pi = (\gtrsim_i)_{i \in \mathcal{N}}$ 

**Proposition.** Pre( $\mathscr{A}$ ), the set of all preference relations forms a lattice with the following operation:

- $a\left(\gtrsim_i \land \gtrsim_j\right)b$  if and only if  $\left(a\gtrsim_i b\right) \& \left(a\gtrsim_j b\right)$
- $a \left( \gtrsim_i \lor \gtrsim_j \right) b \text{ iff } \exists \{a_n\}_{n=0}^{\ell} \in \mathscr{A} \text{ for } a_0 = a, a_{\ell} = b \text{ such that } \left( a_n \gtrsim_i a_{n+1} \right) \middle| \left( a_n \gtrsim_j a_{n+1} \right) \forall n \in \{0, \dots, \ell-1\}$

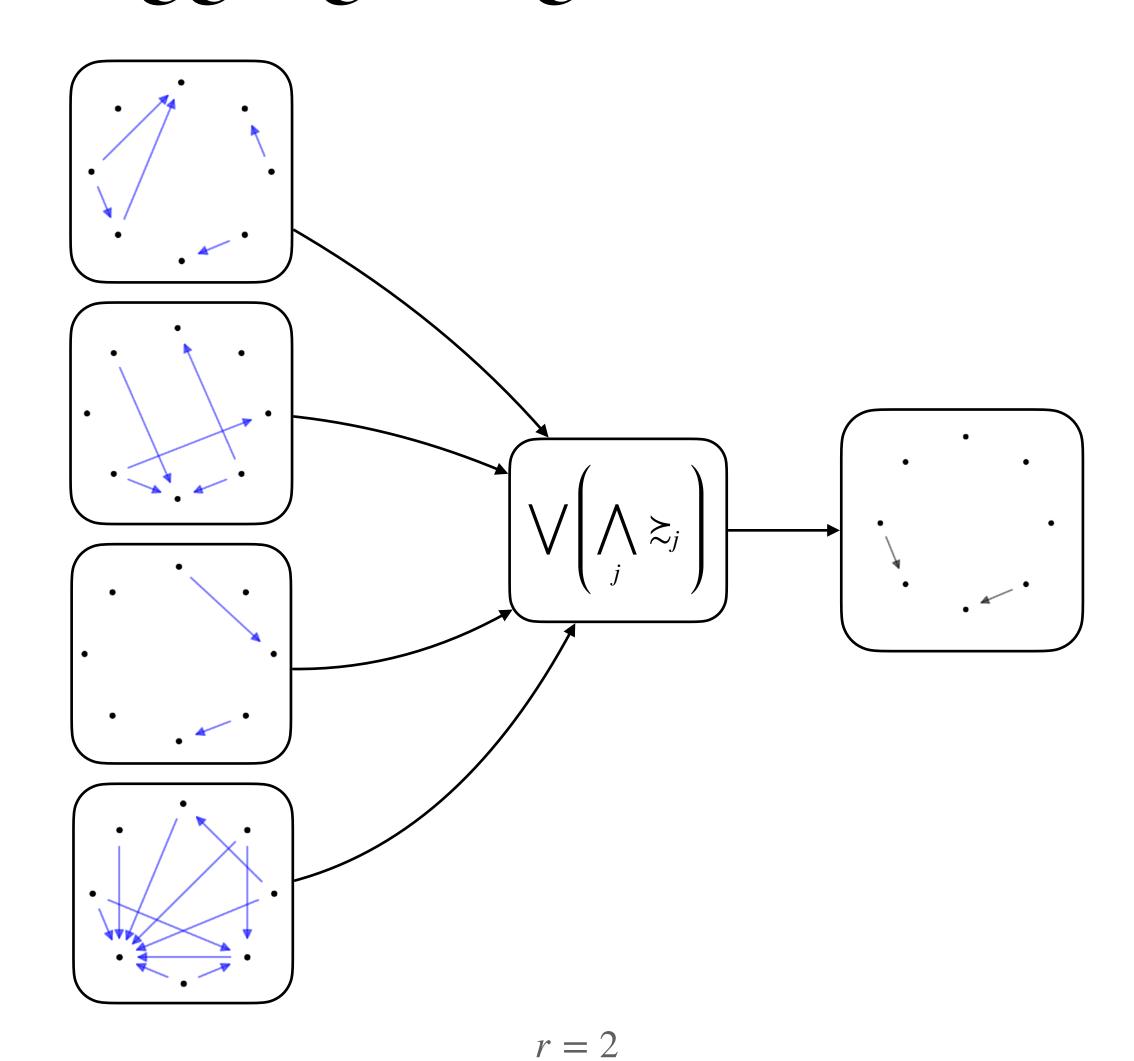
 $Pre(\mathcal{A})$  is called the *preference lattice*.

### Preference Lattice



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# Aggregating Preferences: Centralized



Definition. (Lattice median, Leclerc 1990, 1993):

• Suppose  $\pi = (\geq_i)_{i \in \mathcal{N}}$  is a preference profile

$$\operatorname{Median}_{r}(\pi) = \bigvee_{\substack{J \subseteq \{1,2,\ldots,n\}\\|J| \ge r}} \left( \bigwedge_{j \in J} \gtrsim_{j} \right)$$

 $r \in \{1,2,...,N\}$ , majority threshold

- ▶ Join consensus, Median<sub>1</sub>( $\pi$ ) =  $\forall_{j \in \mathcal{N}} \gtrsim_j$
- Meet consensus, Median<sub>N</sub> $(\pi) = \land_{j \in \mathcal{N}} \gtrsim_j$

#### Properties:

- anonymity
- unanimity
- ► r-middle

Can we extend the lattice median to the decentralized setting?



# Aggregating Preferences: Dentralized

#### Message-passing preference dynamics:

$$\gtrsim_i' = \varphi \Big( \gtrsim_i , \text{Median}_r \Big[ \psi(\gtrsim_i , \gtrsim_j ) \Big]_{j \in \mathcal{N}_i} \Big)$$

- $\rightarrow \psi$ , message function
  - $\psi(\lesssim_i, \lesssim_j) = \lesssim_j \text{(truthful)}$
  - $\psi(\lesssim_i, \lesssim_j) = \lesssim_i \text{(mimicking)}$
- $\rightarrow \varphi$ , updates function
  - $\varphi(\gtrsim_i, \gtrsim) = \gtrsim_i \land \gtrsim \text{(confirming)}$
  - $\varphi(\gtrsim_i, \gtrsim) = \gtrsim_i \lor \gtrsim \text{(learning)}$
  - $\varphi(\gtrsim_i, \gtrsim) = \gtrsim_i$  (single-minded)
  - $\varphi(\gtrsim_i,\gtrsim)=\gtrsim$  (open-minded)

Local updates lead to operator on preference profiles:

$$F: \operatorname{Pre}(\mathscr{A})^N \to \operatorname{Pre}(\mathscr{A})^N$$

#### Algorithm.

*Inputs*: Initial preference relations

- 1. Agent j sends "message"  $\psi(\lesssim_i, \lesssim_j)$  (preference relation) to Agent i
- 2. Agent *i* collects messages received by neighbors
- 3. Agent *i* aggregates messages with median
- 4. Agent *i* updates prior preference  $\succeq_i$  with  $\varphi(\succeq_i, -)$
- 5. Go to back to 1.



# Operator Theory on Lattices

lacksquare Suppose  $\mathcal L$  is a lattice

**Definition.** A map  $f: \mathcal{L} \to \mathcal{L}'$  is *monotone* if  $x_1 \leq x_2$  implies  $f(x_1) \leq f(x_2)$ .

**Definition.** A map  $f: \mathcal{L} \to \mathcal{L}$  is inflationary if  $f(x) \succeq x$ , deflationary if  $f(x) \leq x$ 

**Lemma** (Tarski Fixed Point Theorem). Suppose  $f: \mathcal{L} \to \mathcal{L}$  is monotone and  $\mathcal{L}$  a complete\* lattice. Then, Fix(f) is a complete lattice.

We apply this result to our preference dynamics model

**Theorem.** Suppose the message ( $\psi$ ) and update ( $\varphi$ ) functions are monotone. Then, the equilibrium points (preference profile) of the global message-passing preference dynamical system form a (non-empty) lattice.

**Theorem.** Suppose the update function ( $\varphi$ ) is inflationary or deflationary. Then, for every initial preference profile, the dynamics given locally by

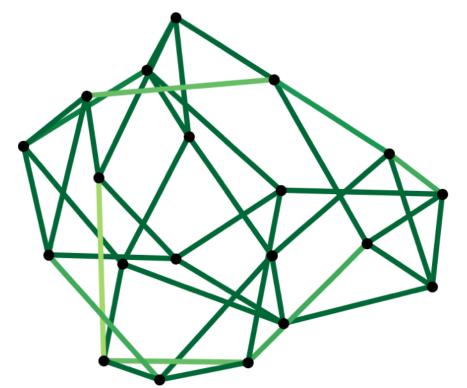
$$\succeq_i' = \varphi(\succeq_i, \operatorname{Median}_r[\psi(\succeq_i, \succeq_j)]_{j \in \mathcal{N}_i})$$
 converge to an equilibrium preference profile.

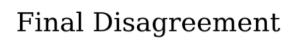


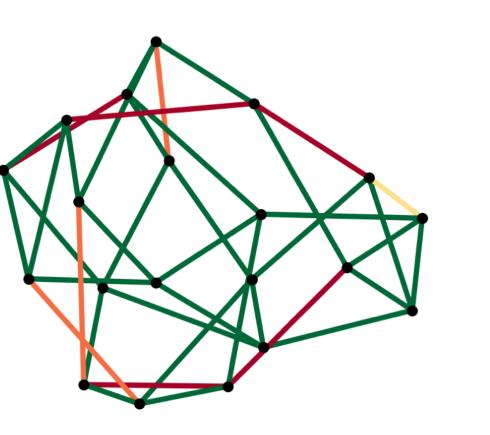
# Evaluating Preference Consensus Algorithms

Comparison of Initial and Final Disagreement

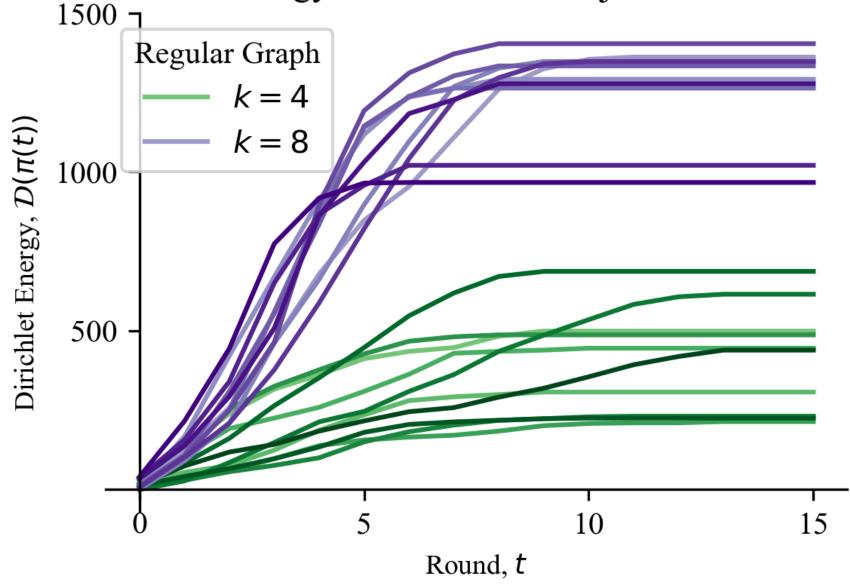
Initial Disagreement







Energy of Preference Trajectories



Kendall tau-distance:  $d(\gtrsim_i,\gtrsim_j)=\#\{(a,b):(a\succsim_i b)\ \&\ (b\succsim_j a)\}$ 

- N = 20 agents; M = 5 alternatives
- graph is k-regular; r chosen uniformly at random from  $\{1,2,\ldots,k\}$
- $\varphi$ , truthful;  $\psi$ , learning
- Initial preference relations:
  - sample pairs (a, b) with probability p (parameter)
  - discard preference relations violating transitivity



## Open Questions & Future Work

- Dynamics
  - How does the distribution of *r* effect the structure of the equilibria lattice? Structure of the graph?
  - when do limit cycles emerge?
- Adversarial setting (control inputs)
  - understand strategic manipulation of preferences in network setting
  - what nodes or edges are especially vulnerable?
  - design control inputs on a subset of infected nodes
- Alignment (optimal control)
  - enforce constraints on allowable preference relations
  - barrier functions



# Thanks! Questions

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#### "Network Preference Dynamics using Lattice Theory"

#### Network Preference Dynamics using Lattice Theory

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feasible options are ranked based on independent information not enough information or time to adequately compare all feasible options, a decision dilemma ensues [1]. In this ton, as well as addition and subtraction of alternatives [9]. Our time and the subtraction of alternatives [9]. Our time and time and time are time are time and time are time and time are time are time are time and time are time are time and time are time paper, we design a decentralized mechanism for multi-agent systems to locally share preferences in order to augment each agent's independent decision-making capabilities and, agent's independent decision-making capabilities and agent's independent decision-making capabilities and agent's independent decision-making capabilities and agent's independent decision-making ca to strategic manipulation (for instance, see [2]).

In their traditional use in economics and social science, they are computable (Section V). preferences are commonly associated with subjective taste. Here, we take the liberty to extend the notion of preference to arbitrary decision-making contexts, ranging from ence to arbitrary decision-making contexts, ranging from reinforcement learning (RL) [3] to games [4] to language models [5]. Formally, a *preference relation* is a relative attice structure they introduce is different from ours. Several tattice structure they introduce is different from ours. Several structure they introduce is different from ours. Several structure they introduce is different from ours.

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Abstract—Preferences, fundamental in all forms of strategic behavior and collective decision-making, in their raw form, are an abstract ordering on a set of alternatives. Agents, we assume, revise their preferences as they gain more information about other agents. Exploiting the ordered algebraic structure exploitable analytical structure. To analyze raw preference their preferences based on aggregations of the preferences of their neighbors in a graph. We demonstrate the existence of equilibrium points of the resulting global dynamical system of local preference updates and provide a sufficient condition for trajectories to converge to equilibria: stable preferences. Finally, we present numerical simulations demonstrating our primary motivation for avoiding utility theory is that time-variant utility functions cannot capture preference formation for trajectories to converge to equilibria: stable preferences. Finally, we present numerical simulations demonstrating our primary motivation for avoiding utility theory is that time-variant utility functions cannot capture preference formation cardinal utility functions cannot capture preference formation and utility functions cannot and indecisiveness, i.e., the agent has not made up its mind or faced a choice [7], is captured by incomplete preference relations and not utility functions, although extended preference are another approach here [8].

We propose an entirely new methodology that relies are responsible for unilateral decision making, a number of on imposing an information order on preference relations feasible options are ranked based on independent information sources as well as the biases of agents. When there is also, is capable of modeling preference change (Section IV). when possible, avoid operational paralysis. As our main contribution, we introduce and offer a qualified solution to the preference dynamics problem which asks how preferences the preference dynamics problem which asks how preferences the preference dynamics model can be viewed as a of agents evolve as they gather information and interact with other agents. This challenge is distinguished from the social other agents. This challenge is distinguished the choice problem, to design (centralized) preference-aggregation mechanisms that satisfy criteria such as fairness or resistance mechanisms that satisfy criteria such as fairness or resistance points in a preference dynamical system. We characterize the stable preferences and discuss sufficient conditions for when

options). While preference relations represented by utility

[14], [15], and there is a corpus of literature addressing the design of preference-aggregation mechanisms, originating with [16], but, to our knowledge, none of these works have approached the problem of preference dynamics from either a decentralized or algebraic point-of-view, both of which we do here. The several efforts to formalize classical consensu

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Code: <a href="https://github.com/hans-riess/prefdynamo/">https://github.com/hans-riess/prefdynamo/</a>

