

Network Preference Dynamics using Lattice Theory

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2024 American Control Conference
July 10-12, 2024
Toronto, Canada

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Introduction

Key contributions:

- ▶ Preference lattice
 - Algebraic model for preference: meet (\wedge) and join (\vee) 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

- ▶ $a \vee b = \min\{c : c \geq a, c \geq b\}$ (called *join*)
- ▶ $a \wedge b = \max\{c : c \leq a, c \leq b\}$ (called *meet*)

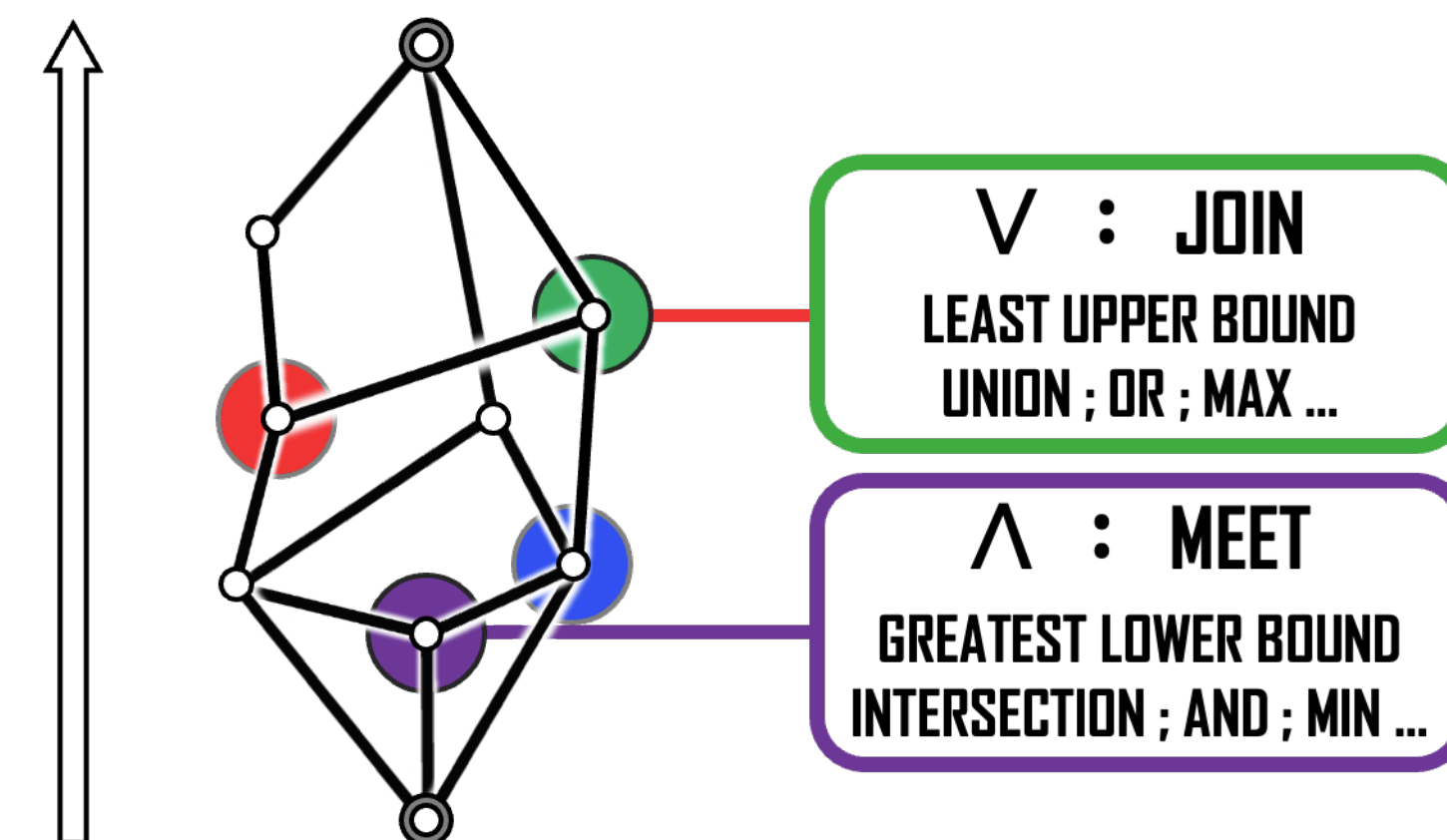
exist for all $a, b, \in \mathcal{L}$.

Write,

$$a_1 \vee a_2 \vee \dots \vee a_n = \bigvee a_i$$
$$a_1 \wedge a_2 \wedge \dots \wedge a_n = \bigwedge a_i$$

Examples.

- \mathbb{R}^n or \mathbb{Z}^n , $\wedge : \min$, $\vee : \max$
- Boolean algebras: $\wedge : \&$, $\vee : |$
- 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:

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

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

- $\forall a, a', b \in \mathcal{A}, (a \succsim a') \& (a' \succsim b) \rightarrow (a \succsim b)$ implies $i \succsim_u k$ (transitivity);
- $\forall a \in \mathcal{A}, a \succsim a$ (reflexivity);

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

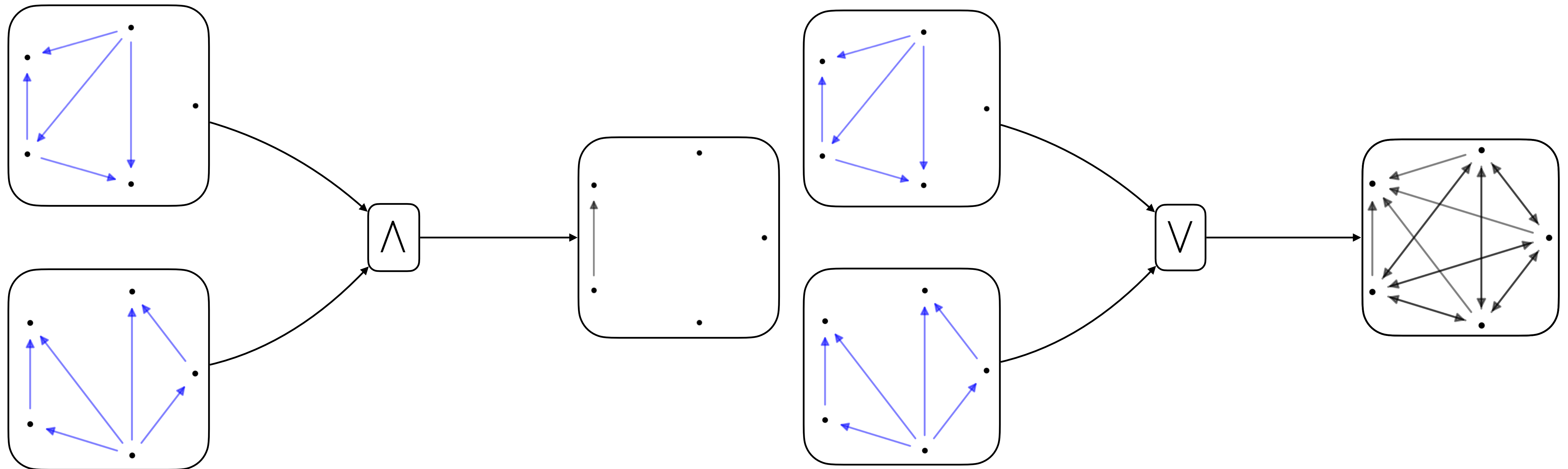
A *preference profile* is a tuple $\Pi = (\succsim_i)_{i \in \mathcal{N}}$

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

- $a \left(\succsim_i \wedge \succsim_j \right) b$ if and only if $(a \succsim_i b) \& (a \succsim_j b)$
- $a \left(\succsim_i \vee \succsim_j \right) b$ iff $\exists \{a_n\}_{n=0}^{\ell} \in \mathcal{A}$ for $a_0 = a, a_{\ell} = b$ such that $(a_n \succsim_i a_{n+1}) \mid (a_n \succsim_j a_{n+1}) \forall n \in \{0, \dots, \ell - 1\}$

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

Preference Lattice



$$a \left(\succeq_i \vee \succeq_j \right) b \text{ iff } (a_n \succeq_i a_{n+1}) \mid (a_n \succeq_j a_{n+1})$$

$$a \left(\succeq_i \wedge \succeq_j \right) b \text{ iff } (a \succeq_i b) \& (a \succeq_j b)$$

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Aggregating Preferences: Centralized

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

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

$$\text{Median}_r(\pi) = \bigvee_{\substack{J \subseteq \{1,2,\dots,n\} \\ |J| \geq r}} \left(\bigwedge_{j \in J} \succsim_j \right)$$

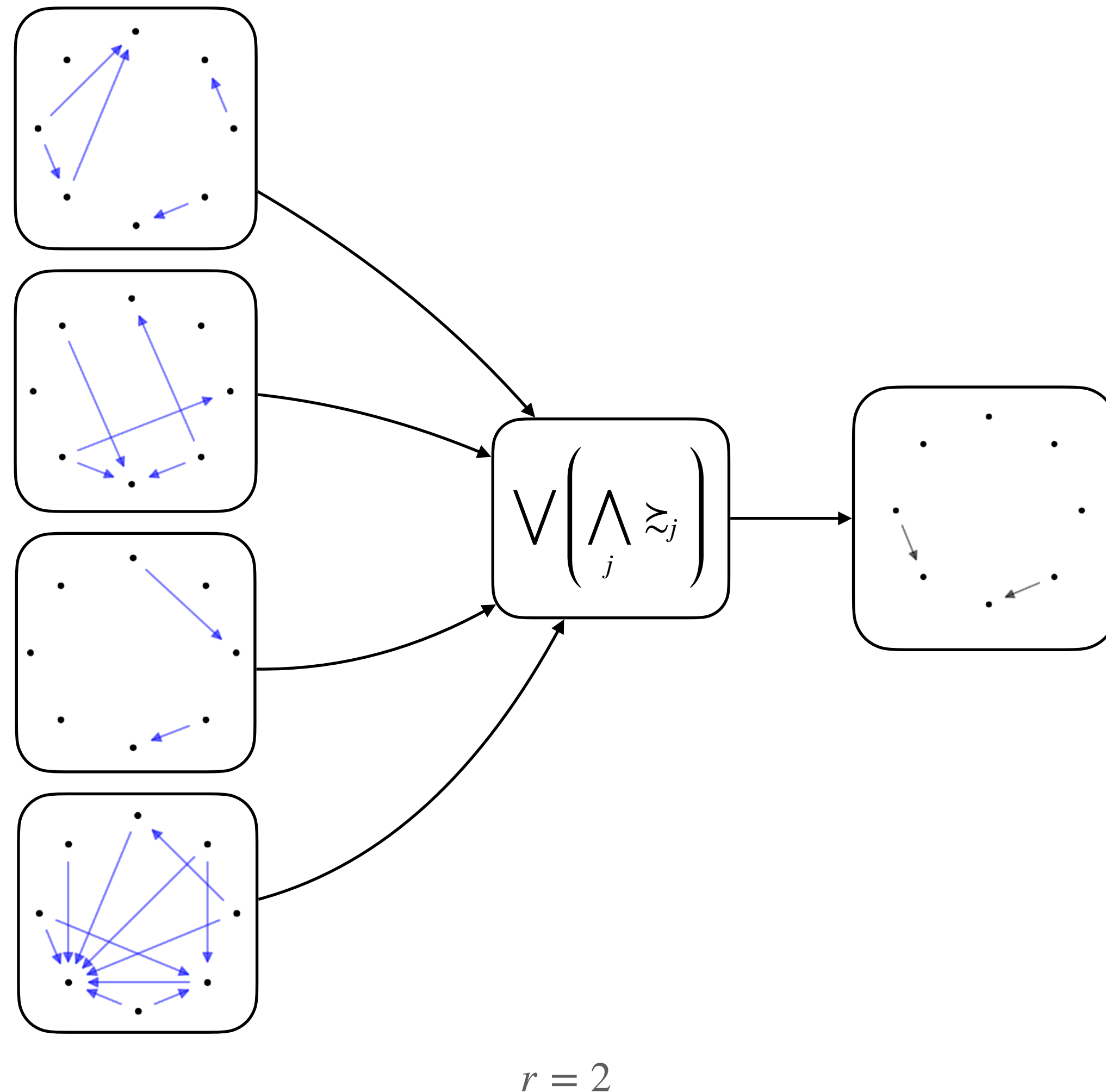
$r \in \{1,2,\dots,N\}$, majority threshold

- Join consensus, $\text{Median}_1(\pi) = \bigvee_{j \in \mathcal{N}} \succsim_j$
- Meet consensus, $\text{Median}_N(\pi) = \bigwedge_{j \in \mathcal{N}} \succsim_j$

Properties:

- anonymity*
- unanimity*
- r-middle*

Can we extend the lattice median to the decentralized setting?



Aggregating Preferences: Decentralized

Message-passing preference dynamics:

$$\tilde{z}'_i = \varphi\left(\tilde{z}_i, \text{Median}_r\left[\psi(\tilde{z}_i, \tilde{z}_j)\right]_{j \in \mathcal{N}_i}\right)$$

- ▶ ψ , message function
 - $\psi(\tilde{z}_i, \tilde{z}_j) = \tilde{z}_j$ (truthful)
 - $\psi(\tilde{z}_i, \tilde{z}_j) = \tilde{z}_i$ (mimicking)
- ▶ φ , updates function
 - $\varphi(\tilde{z}_i, \tilde{z}) = \tilde{z}_i \wedge \tilde{z}$ (confirming)
 - $\varphi(\tilde{z}_i, \tilde{z}) = \tilde{z}_i \vee \tilde{z}$ (learning)
 - $\varphi(\tilde{z}_i, \tilde{z}) = \tilde{z}_i$ (single-minded)
 - $\varphi(\tilde{z}_i, \tilde{z}) = \tilde{z}$ (open-minded)

Local updates lead to operator on preference profiles:

$$F : \text{Pre}(\mathcal{A})^N \rightarrow \text{Pre}(\mathcal{A})^N$$

Algorithm.

Inputs: Initial preference relations

1. Agent j sends “message” $\psi(\tilde{z}_i, \tilde{z}_j)$ (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 \tilde{z}_i with $\varphi(\tilde{z}_i, -)$
5. Go to back to 1.

Operator Theory on Lattices

- Suppose \mathcal{L} is a lattice

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

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

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

- We apply this result to our preference dynamics model

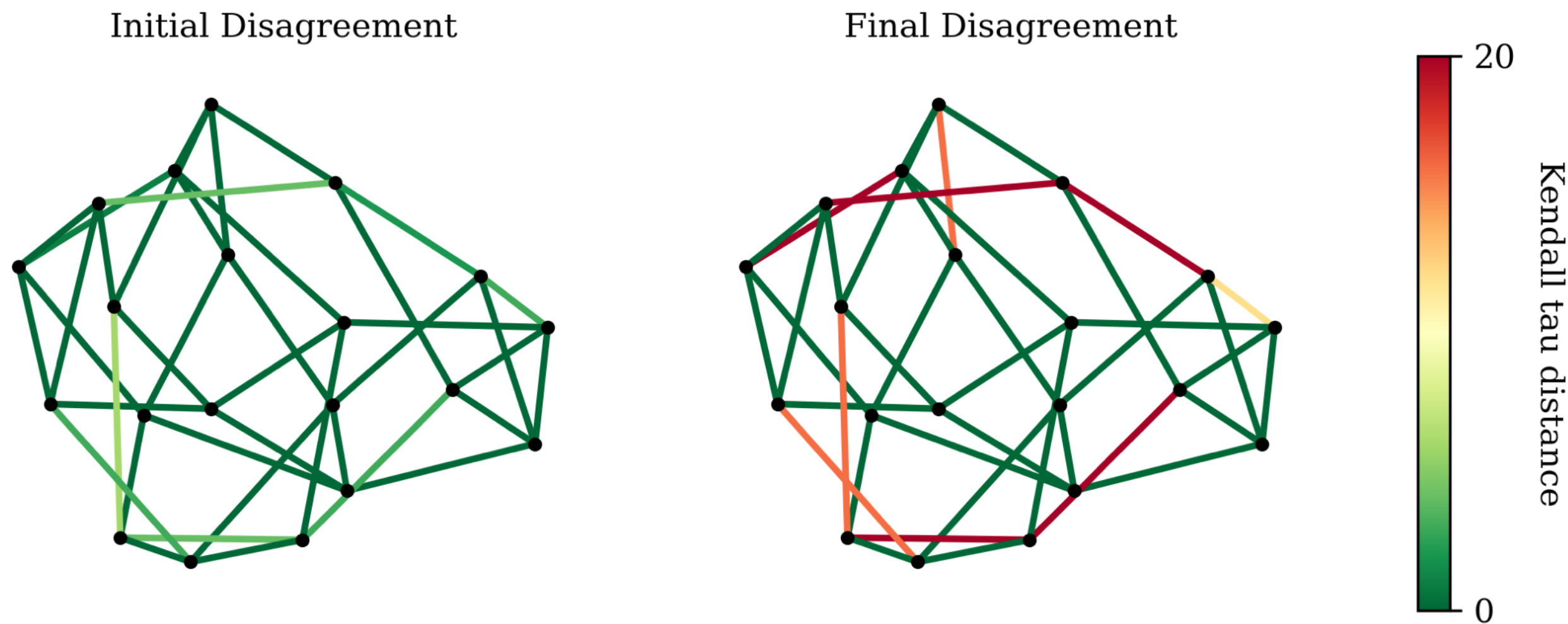
Theorem. Suppose the message (ψ) and update (φ) functions are monotone. Then, the equilibrium points (preference profile) of the global message-passing preference dynamical system form a (non-empty) lattice.

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

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

Evaluating Preference Consensus Algorithms

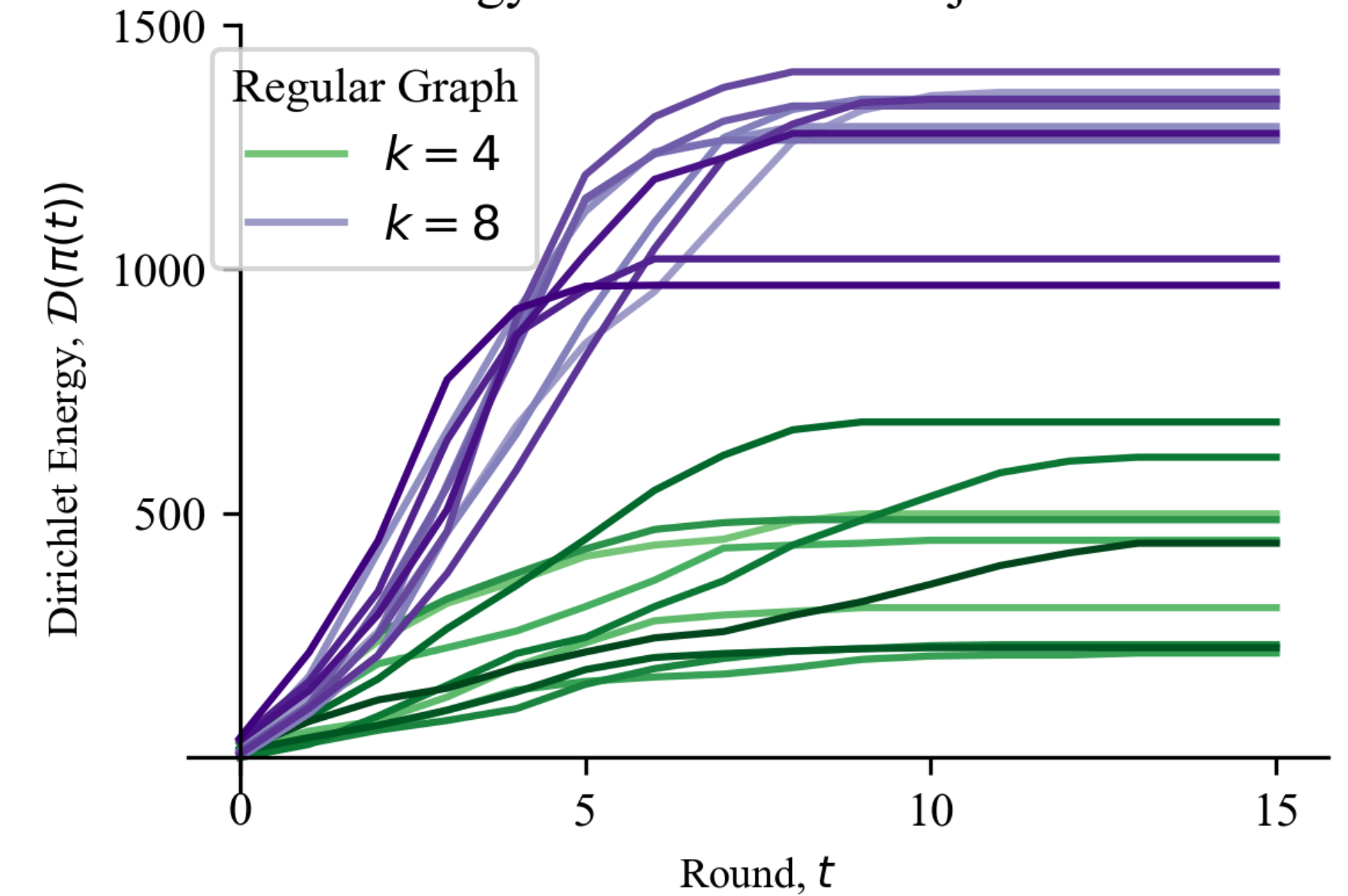
Comparison of Initial and Final Disagreement



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

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

Energy of Preference Trajectories



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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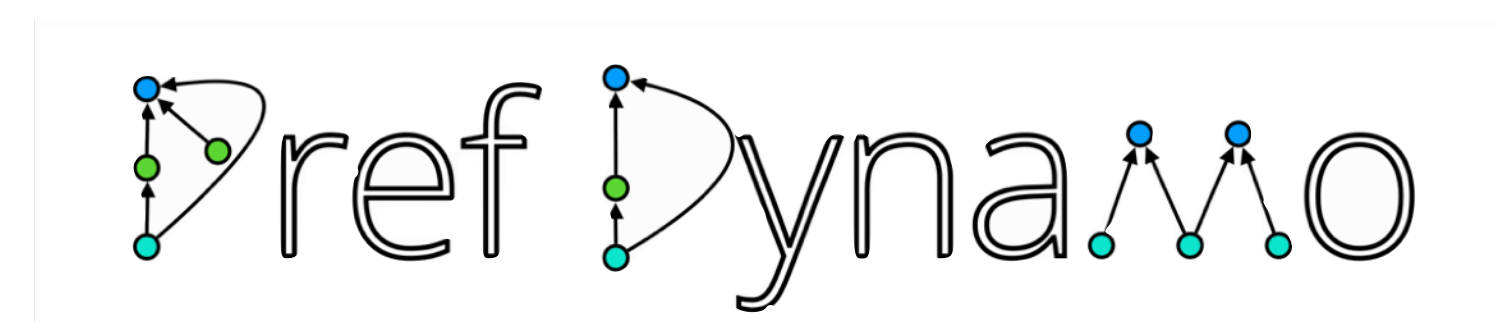
“Network Preference Dynamics using Lattice Theory”



Paper: <https://hansriess.com/papes/acc2024.pdf>

Slides: <https://hansriess.com/slides/acc2024.pdf>

Code: <https://github.com/hans-riess/prefdynamo/>



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