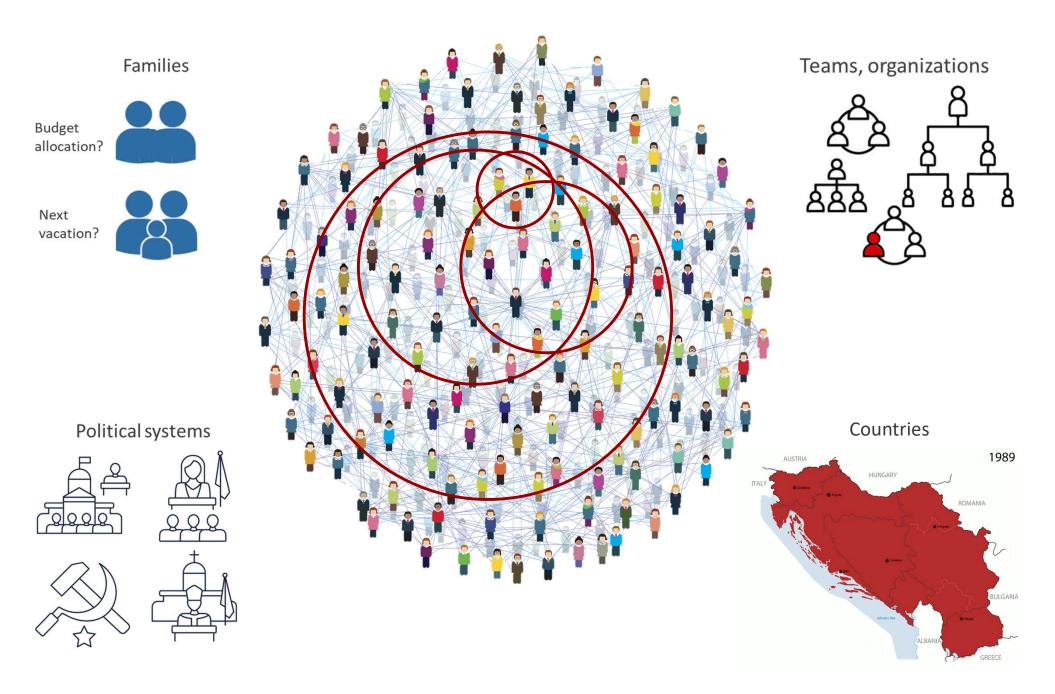
Beyond collective intelligence: Collective adaptation

Mirta Galesic and Henrik Olsson

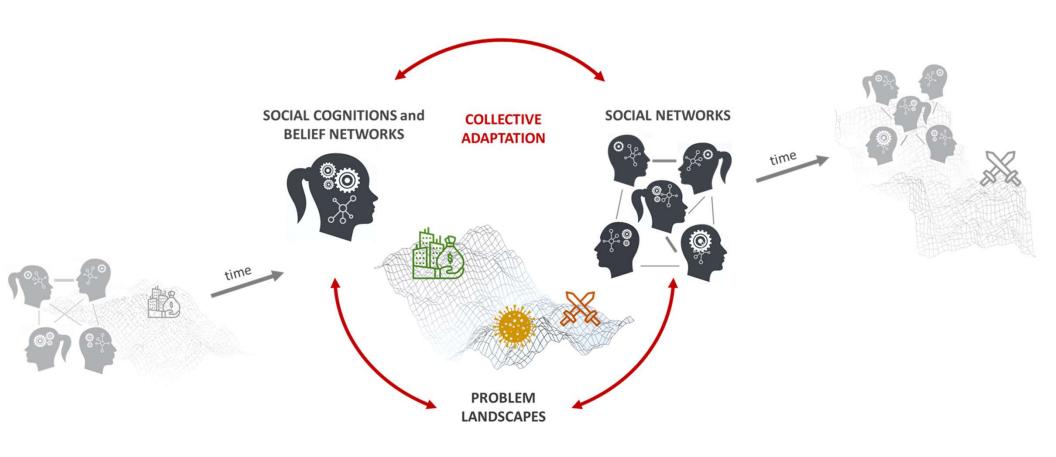
with: Daniel Barkoczi (South Denmark University), Andrew Berdahl (U Washington), Dora Biro (U Oxford, UK), Giuseppe Carbone (Polytechnic U Bari, Italy), Ilaria Giannoccaro (Polytechnic U Bari, Italy), Robert Goldstone (U of Indiana), Cleotilde Gonzalez (CMU), Anne Kandler (Max Planck Institute, Germany), Albert Kao (SFI), Rachel Kendal (Durham U, UK), Michelle Kline (Brunel University London, UK), Eun Lee (U Colorado Boulder), Giovanni Francesco Massari (Polytechnic U Bari, Italy), Alex Mesoudi (U of Exeter, UK), Niccolo Pescetelli (NJIT), Sabina Sloman (CMU), Paul E. Smaldino (UC Merced), Daniel L. Stein (NYU)

Royal Society Interface (2023). <u>https://doi.org/10.31235/osf.io/5f2ad</u>

Adaptive collectives: networks and cognitions



Collective adaptation of complex social systems



Beyond collective intelligence

• Collective intelligence: What cognitive strategies and social structures are best to solve a specific, well-defined task?





Barkoczi & Galesic, 2018; Galesic et al., 2019 Woolley et al., 2010, 2015; Centola 2023

- Collective adaptation: What trajectories can societies take while navigating multiple and ever-changing problems?
 - Instead of "Who is stupid/intelligent?" we ask
 "How and why did we get here?" "Where are we going"?

Collective adaptation at the summer school









Multiple simultaneous goals

Learning, collaborating, making friends ... impossible to optimize

Dynamic problem landscape

Importance of different goals is changing over time

Adapting your connections and cognitive strategies

 To accomplish these goals, you team up with different people and use different strategies to make group decisions

Path dependence

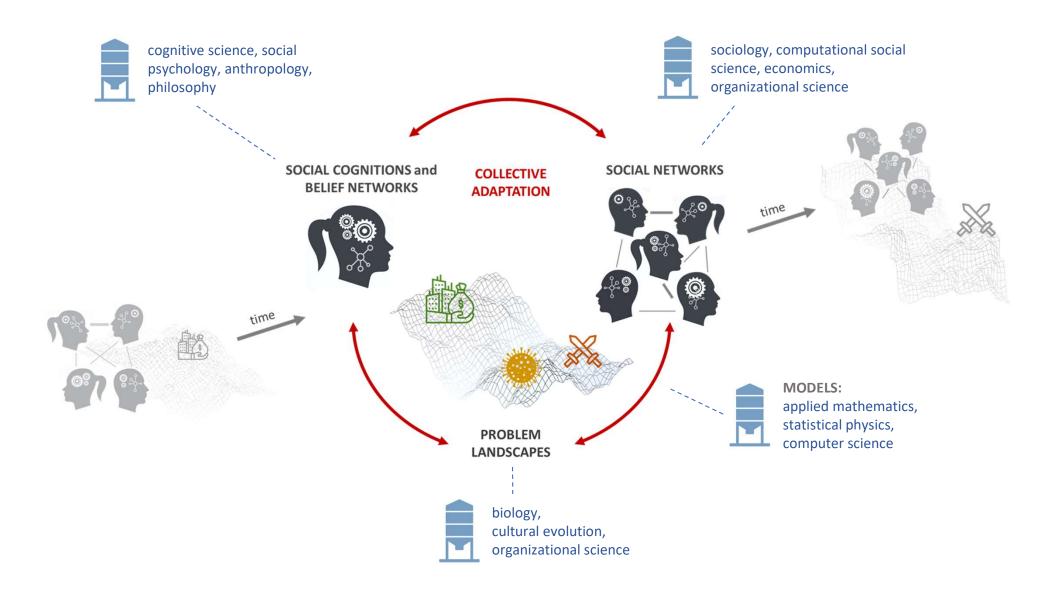
 Groups and strategies chosen to solve early problems will affect the way you solve later problems

Collective myopia

- Not always obvious what is the best way to structure the group or make group decisions

Studying collective adaptation: State of the art

Collective adaptation: Disciplinary silos

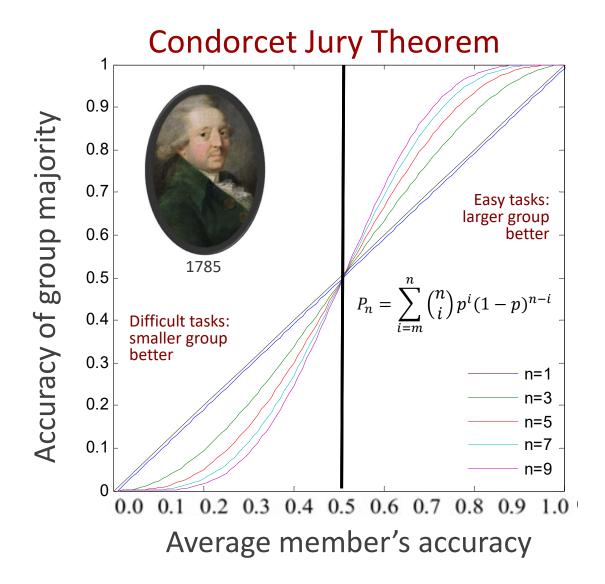


Overlapping views, different labels

- Collective intelligence (Graves, 1842, Woolley et al, 2010, Malone & Bernstein, 2015)
- Social learning (Bandura, 1977; Hoppitt & Laland, 2013; Kendal et al, 2018; Yaniv, 2004)
- Collective problem solving (Conradt & Roper, 2010; Gonzalez et al., 2015; Hills et al., 2015; Mehlhorn et al., 2015)
- Wisdom of crowds (Condorcet, 1783; Galton, 1907; Davis-Stober et al., 2014; Mellers et al., 2014; Budescu & Chen, 2015)
- Group decision making (Page, 2008; Stasser & Titus, 1985)
- Belief dynamics (Centola & Macy, 2007; Epstein, 2014; Galesic et al., 2021; Pentland, 2014; Proskurnikov & Tempo, 2017; Vallacher, Read, & Nowak, 2017)
- Cultural evolution (Boyd & Richerson, 1985; Mesoudi, 2016)
- Game theory (Friedman, 1998; Newton, 2018; Ostrom, 2010)
- **Group minds** (Goldstone & Theiner, 2017; Hinsz et al., 1997)
- ...

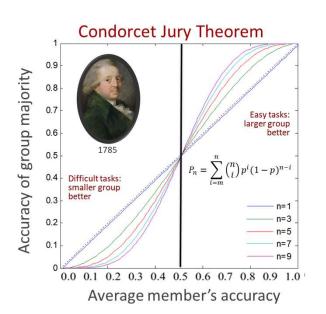
Parallel efforts: an example

Accuracy of majority rule depends on group size:



Parallel efforts: an example

Accuracy of majority rule depends on group size



Rediscoveries:

- political science (Grofman et al., 1984)
- cultural evolution (Boyd & Richerson, 1985)
- statistical physics (Krapivsky & Redner, 2003)
- psychology (Hastie & Kameda, 2005)
- sociology (Centola & Macy, 2007)
- biology (King & Cowlishaw, 2007)

Smaller groups can outperform larger groups facing several tasks:

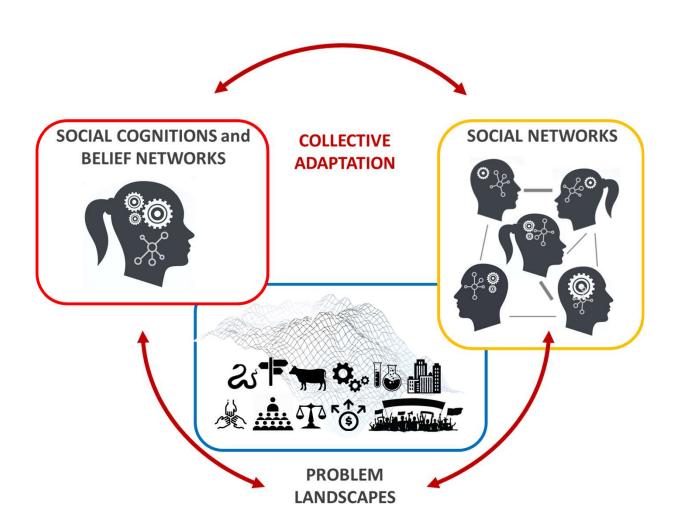
- political science (Grofman et al., 1984)
- biology (Kao & Couzin, 2014)
- cognitive science (Galesic et al., 2018)

Building blocks of collective adaptation

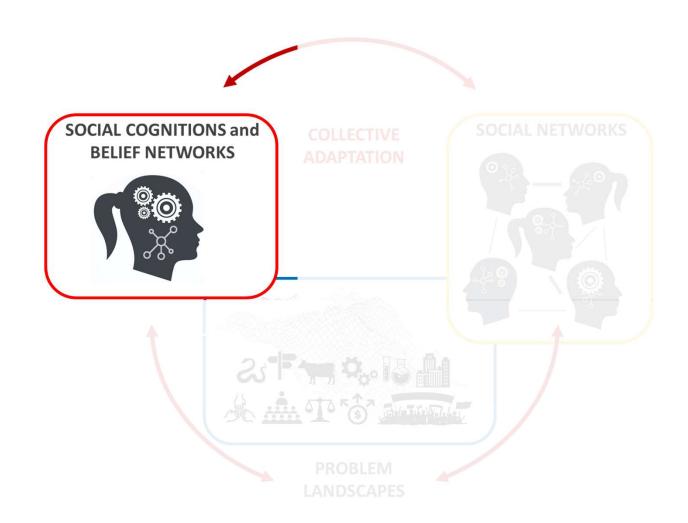
Building blocks







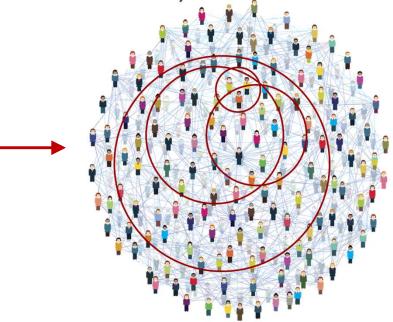
Building blocks



Social cognitions

Strategies for social interactions (Hertwig & Hoffrage, 2013):

- integration of social information (Hoppit & Laland, 2013)
- coordination (Grice, 1975; Moussaid et al., 2011)
- **cooperation** (Axelrod, 1984; Bowles & Gintis, 2013)
- exploration (Hills, et al., 2015; Mehlhorn et al., 2015)
- network building and revision (Jackson, 2010)
- innovation, etc.



Studied under different labels:

- Social learning strategies
- Belief updating strategies
- Group decision-making rules
- Voting procedures
- Aggregation procedures

Strategies for integrating social information

Three basic classes:

• Frequency-dependent strategies: majority, plurality, unanimity, minority rules, complex contagion

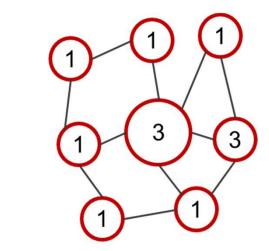
Condorcet 1785; political science (Heinberg, 1932); statistics (Penrose, 1946); psychology (Asch, 1955); economics (Plott, 1967); cultural evolution (Boyd & Richerson, 1985); computer science (Parhami, 1994); statistical physics (Krapivsky & Redner, 2003); biology (King & Cowlishaw, 2007); sociology (complex contagion; Centola & Macy, 2007)

• Averaging strategies: with or without weights, e.g. advice taking, voter model, contagion rules, blending inheritance

Galton, 1907; economics (DeGroot, 1974; Golub & Jackson, 2010); advice taking (Molleman et al., 2020; Yaniv, 2004); statistical physics (Ising models; Castellano et al., 2009); cultural evolution (blending inheritance: Boyd & Richerson, 1985); network science (contagion; Newman, 2003)

• Model-based strategies: follow leader, expert, similar, confident, liked, best

cultural evolution (high status: Henrich & Gil-White, 2001), social psychology (liking, authority: Cialdini & Trost, 1998), cognitive psychology (similarity, Wisdom et al., 2013), law (confidence; Penrod & Cutler, 1995)



2

Different beliefs about important problems

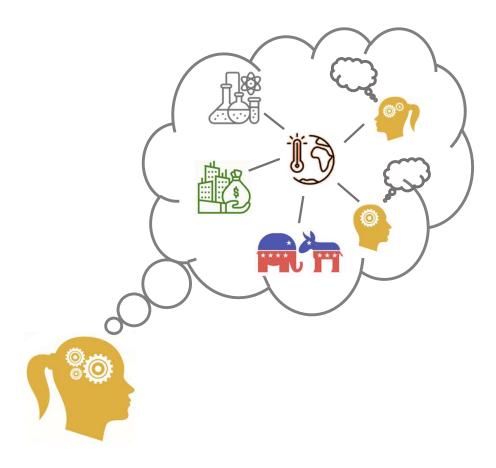








Different belief networks





Models of belief dynamics

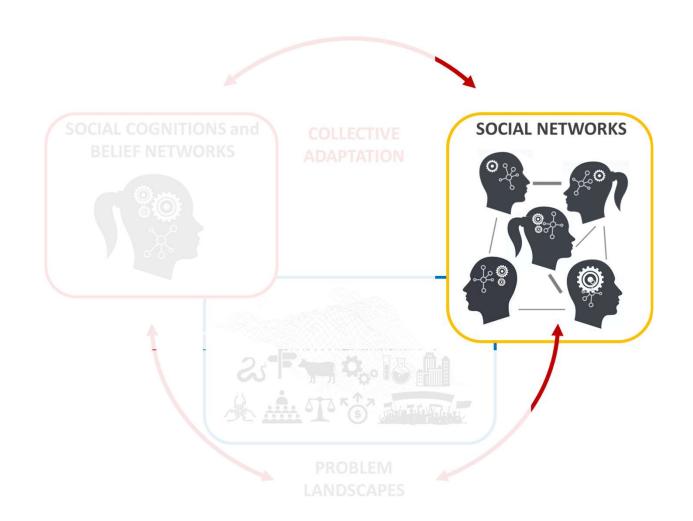
Many analogies (Olsson & Galesic, in prep, Trends in Cog Sci):

- **Epidemiological models,** where transmission of belief is like a transmission of disease (Newman, 2003; Cooney et al., 2022)
- Ferromagnetic models, where beliefs align with each other like spins in a crystal lattice (Castellano, 2009; Dalege et al., 2022)
- **Percolation**, where beliefs seep through a society like liquid through a substance (Duffie et al., 2010; Li & Wang, 2019)
- **Balance**, where beliefs and individuals align in a way that leads to most consistent relationships on the level of pairs and triads (Heider, 1958; Pham et al., 2020)
- **Expected utility**, where beliefs change in line with a weighted average of different cognitions (Ajzen, 1991; Friedkin & Bullo, 2017)
- **Evolution,** where beliefs evolve in the process of cultural learning (Richerson & Boyd, 2008; Anderson & Creanza, 2022)
- **Bayesian networks**, where networks of beliefs change in line with their conditional dependencies (Cook & Lewandowsky, 2016; Pallavicini, 2021)

...

• **Forces,** where belief change under combined influence of several distinct social forces (Latane, 1981; Harton et all, 2022)

Building blocks



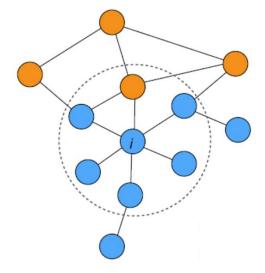
Social environments

• Social networks

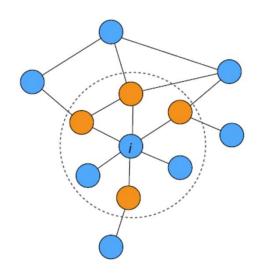
- Perceived vs. objective networks (Thomas theorem, 1928; Denrell, 2005; Gonzalez et al., 2015)
- Size and connectivity (Derex & Boyd, 2016; Lazer & Friedman, 2007; Mason et al., 2008; Giannoccaro et al. 2018)
- Homophily (McPherson et al., 2001; Karimi et al., 2018; Lee et al, 2019)
- Centrality (Barabasi & Albert, 1999; Becker et al., 2017)
- Directed vs. undirected
- Social artifacts
 - Languages and scripts
 - Communication channels
 - Institutions



High homophily \rightarrow False consensus

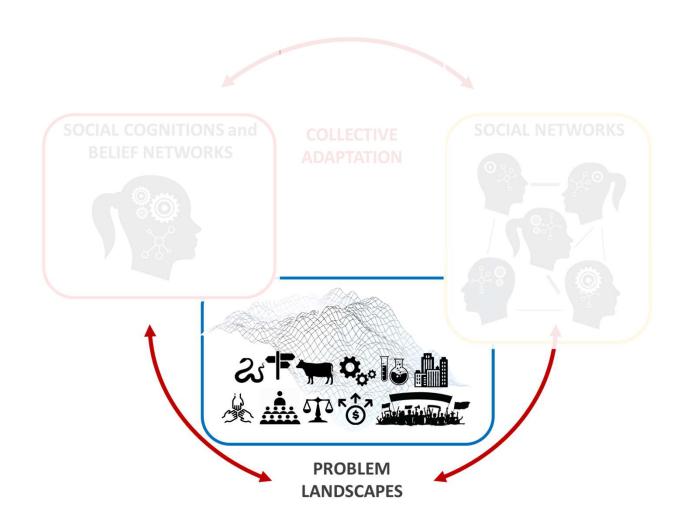


Low homophily \rightarrow False uniqueness



Lee et al., 2019; Galesic, Olsson, Rieskamp, 2018

Building blocks



Problem environments

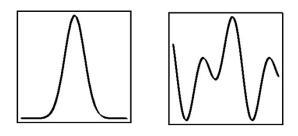
- Structural properties of problems
 - Categorical vs. continuous judgments

Simple and complex task landscapes

One-shot and repeated problems

- Global environment
 - Economic, political, cultural factors that change payoffs of different options, feasibility of different strategies and networks



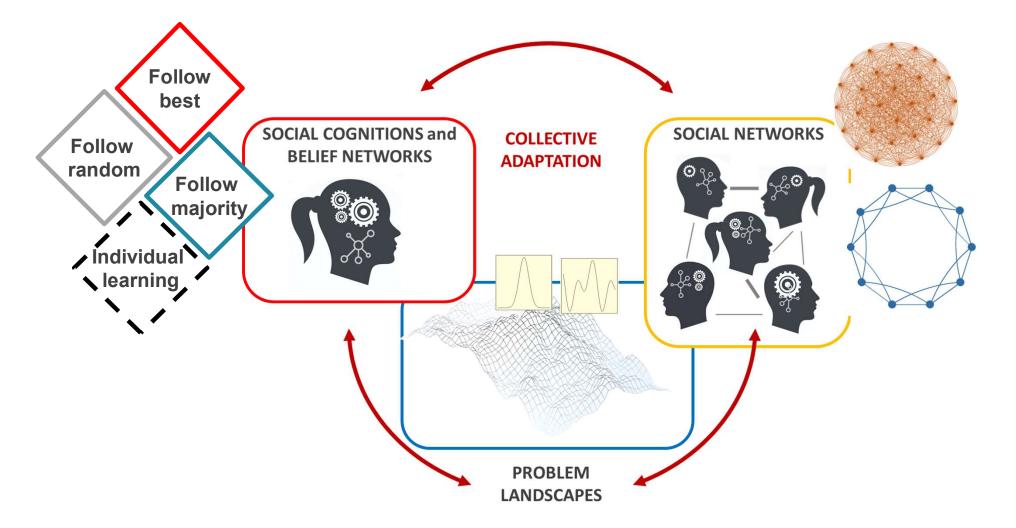






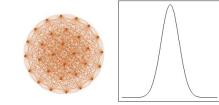
Interaction of building blocks: an example

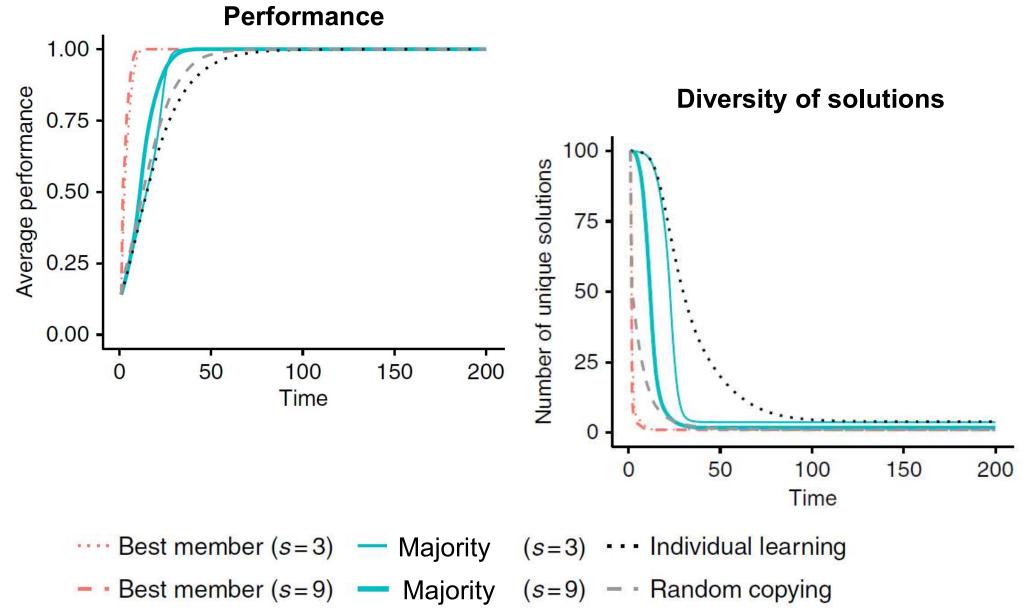
Interaction of building blocks: an example



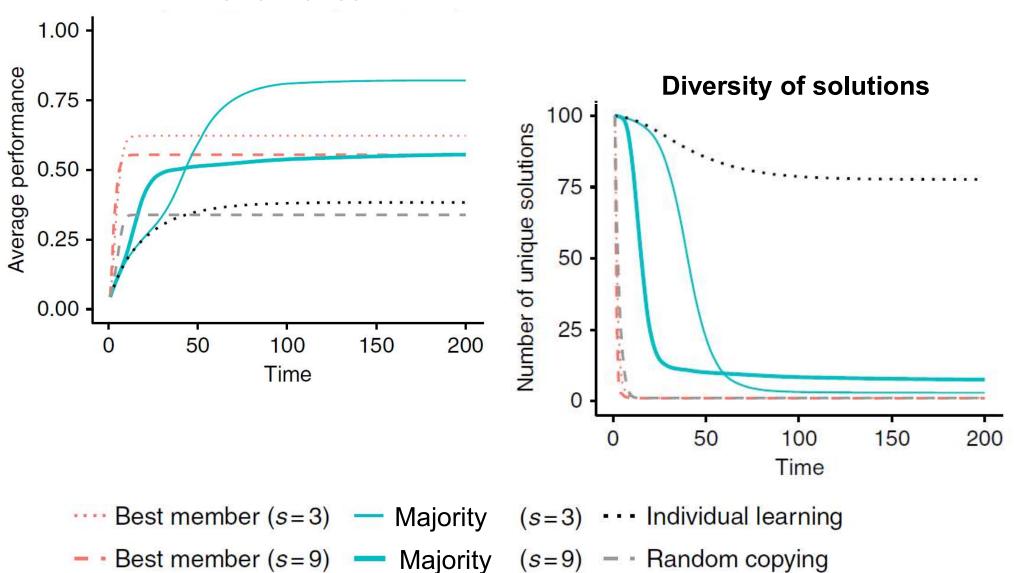
Interaction of building blocks: an example 1 0.9 **Contradictory findings about** 0.8 0.7 network structure and wisdom Average Score 0.6 of crowds 0.5 0.4 0.3 0.2 • Poorly-connected, slow 0.1 networks better (Lazer & 10 20 30 50 70 Time Friedman, 2007, ASQ; Derex & Boyd, 2016, PNAS) Graph 0 Max Max Betweennes 90 Jay Closeness Ava Retweennes Earned 08 ax Avg Clustering Well-connected, fast Var Constraint Ava Clustering Average Points networks better (Mason & Max Max Closeness Min Avg Betweenne Watts, 2012, PNAS) 40 30 6 10 12 2 14 Round

Simple tasks: fast learning strategies better





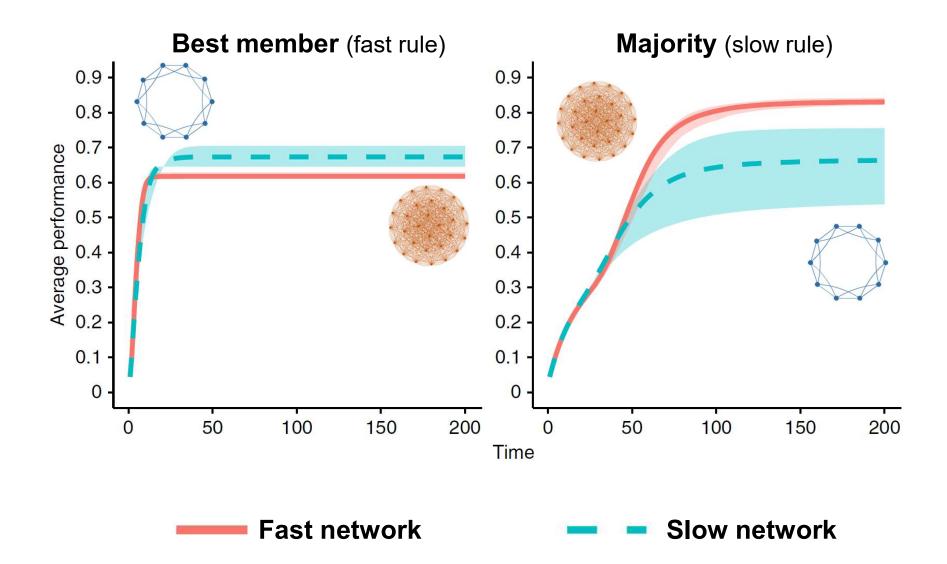
Complex tasks: slow learning strategies better



Performance

Interplay of networks and learning strategies





Interaction of building blocks: an example 0.9 0.8 Contradictory findings about 0.7 0.6 network structure and wisdom **Used best-member** of crowds strategy 0.2 Poorly-connected, slow 0.1 networks better (Lazer & 10 20 0 30 40 50 70 60 80 Time **Apparent contradictions can be** resolved by taking into account Graph the whole system of cognition Max Max Betweenne 90 Min Max Closeness and networks Ava Retweennes Earned 08 Ava Clusterina Well-connected, fast Var Constraint Min Ava Clusterina ⁵oints networks better (Mason & Max Closenes Watts, 2012, PNAS) **Used frequency**based strategy 30 10

2

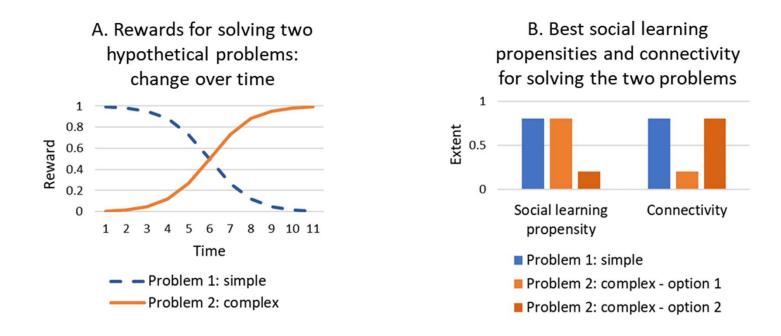
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Round

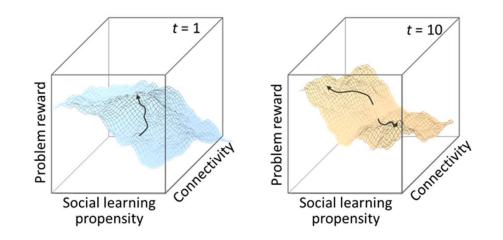
14

Emergence of collective adaptation

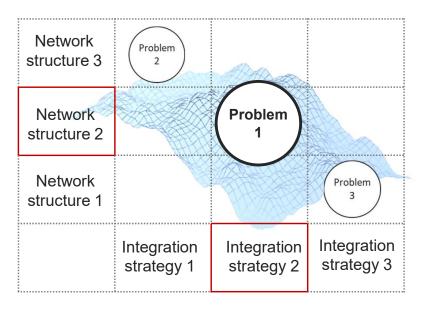
A simple example



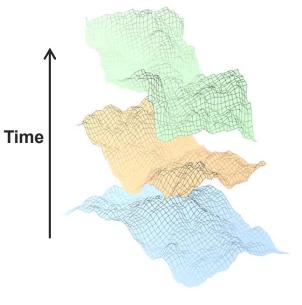
C. The resulting problem landscapes



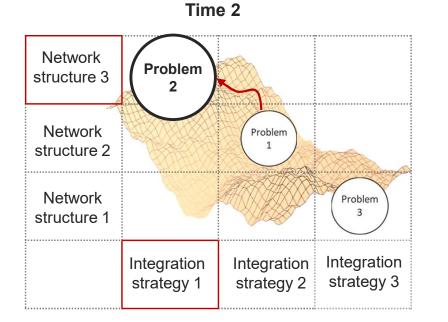
Emergence of collective adaptation

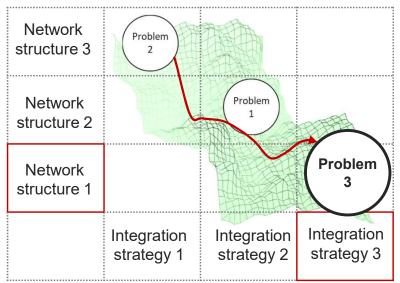


Time 1

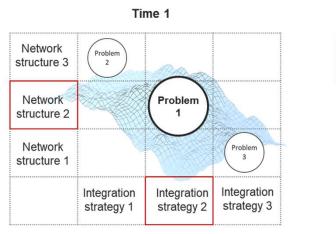


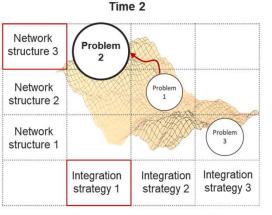
Time 3

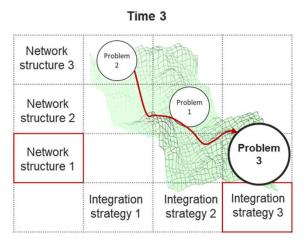




Five implications







- 1) Path dependence
- 2) Not one "collective intelligence" (c.f. van der Maas et al., 2006)
- 3) Multi-task satisficing
- 4) Collective myopia
- 5) Collective imagination: Simulating long jumps



Quantitative models of collective adaptation

Modeling challenges

Models of collective adaptation ...

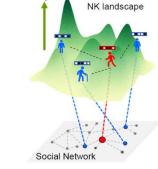
- Should be simple, but grounded in theories of human cognition and sociality
- Should represent the dynamic interplay of cognition, network, and problem structures
- Should produce quantitative predictions that can be tested by empirical data

Analogies for modeling collective adaptation

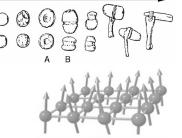
- No analogy is completely correct but some are useful
- Adaptive landscapes
 - Easy to understand and flexible, but can lead to wrong intuitions for high-dimensional, constantly changing spaces (Gavrilets, 2004; Agarwala & Fisher, 2019, Fragata et al., 2019)
- Cultural evolution
 - From tools to institutions, typically no network structure (but see Smolla & Akçay, 2019)
- Statistical physics
 - Reducing dissonance on individual and collective level, no meta-level rules for switching strategies and structures when problems change
- Reinforcement learning
 - Of social learning strategies (Ha & Jeong, 2022), of networks
 - Of individual welfare functions (Wolpert & Tumer, 2001)
- Ecosystem modeling
 - Dynamic interaction networks
- Combinations of analogies

Different analogies for different building blocks, for example:

- Drift diffusion model of learning + an evolutionary process of adaptation to different groups structures and costs of errors (Tump et al., 2022)
- Epidemiological + evolutionary models to study evolution on sociality on different time scales (Cooney et al., 2022)









Analogies for modeling collective adaptation

- No analogy is completely correct but some are useful
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• Cultural evolution

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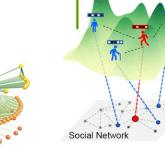
Reinforcement learning

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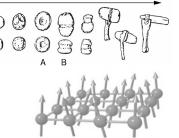
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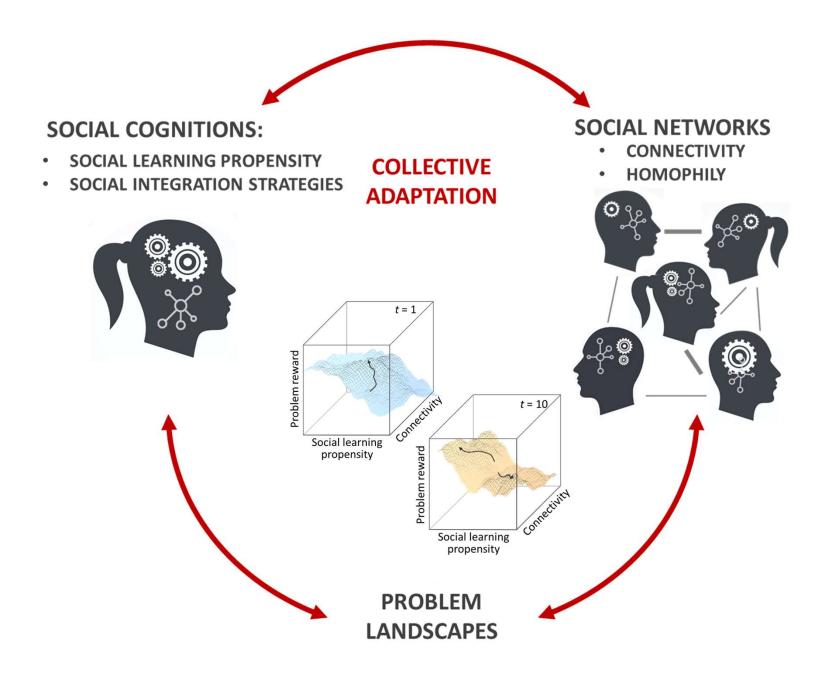
NK landscape



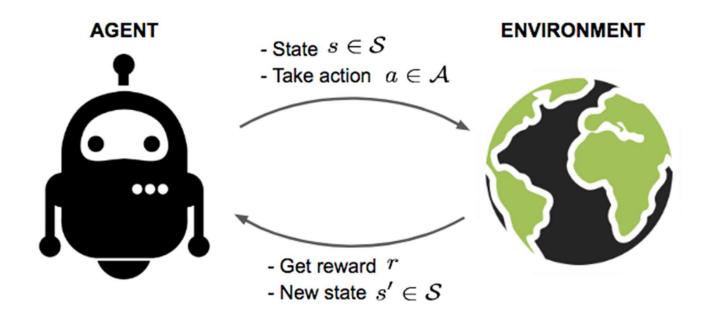




Modeling examples



Modeling: Reinforcement learning of suitable social cognitions and social networks



Modeling: Reinforcement learning of suitable social cognitions and social networks

State of the world: learned value of different cognitions and networks $W_{f,t} = W_{f,t-1} + \alpha (R_{x,t} - V_{x,t-1})$...weight of a feature, e.g. connectivity $V_{x,t} = \sum_{f \in x} W(f)$...overall value of a particular point in the problem environment

Action: choosing the combination of cognitions and networks that currently seems best

 $p(x)_t = \frac{V_{x,t}}{\sum_{x \in X} V_{x,t}}$...choosing the point with the highest value

Reward: success in solving the important current problems

 $R_{x,t,i} = R_{x,t,i}^{ind} + R_{x,t,i}^{col} + \eta \sum_{j \neq i} R_{x,t,j}$...reward includes individual and collective parts

Modeling: Reinforcement learning of suitable social cognitions and social networks

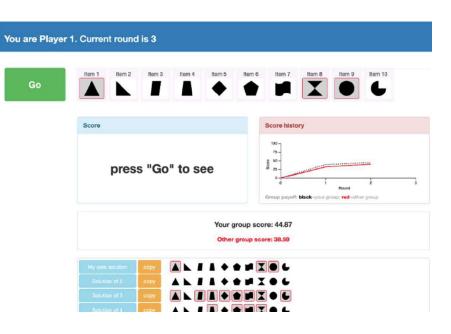
- + Cultural evolution on longer time scales
- to learn about possible cognitions and networks
- to learn the best value of parameter α (speed of adaptation)

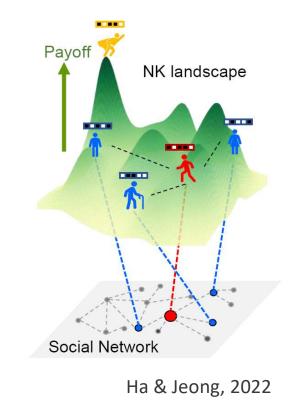
$$W_{f,t} = W_{f,t-1} + \alpha (R_{x,t} - V_{x,t-1})$$



Data: Group experiments

- Participants solve problems in groups
- Problems change over time
- Can learn from each other
- Can rewire their networks





Galesic et al., 2022

2. Different beliefs about important problems



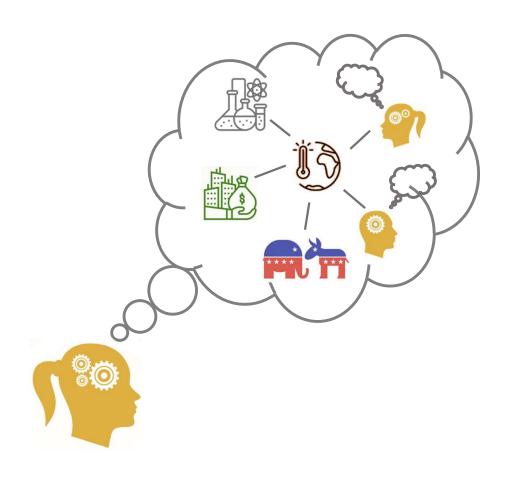


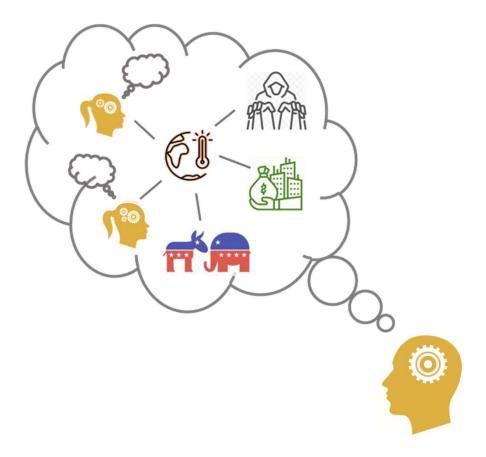




2. Different beliefs about important problems

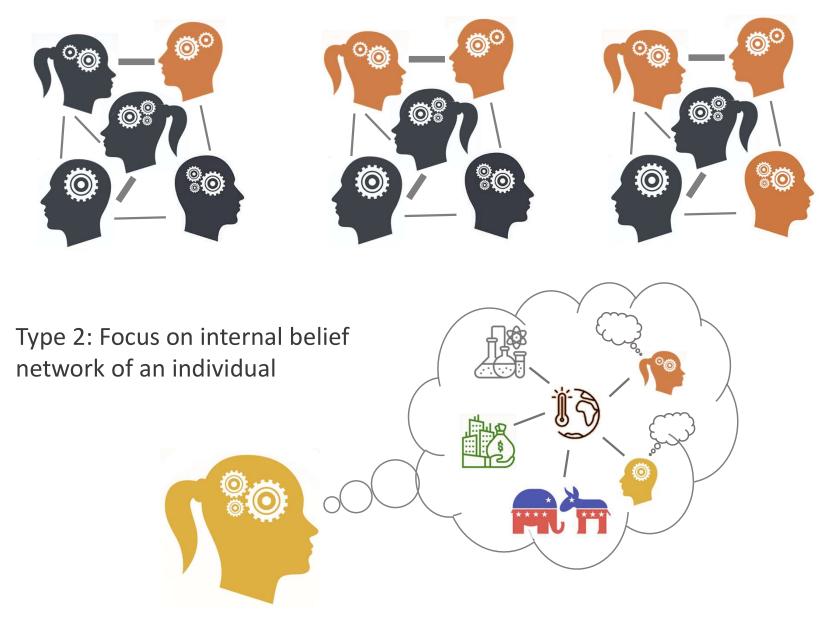
 \rightarrow Different belief networks



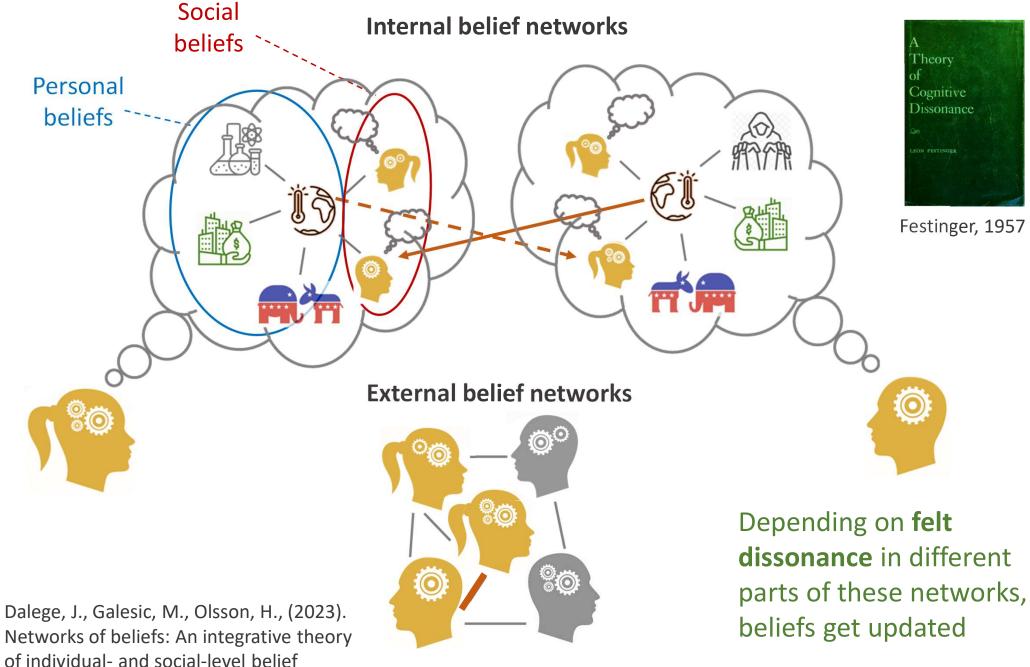


Models of belief dynamics

Type 1: Focus on single beliefs in social networks

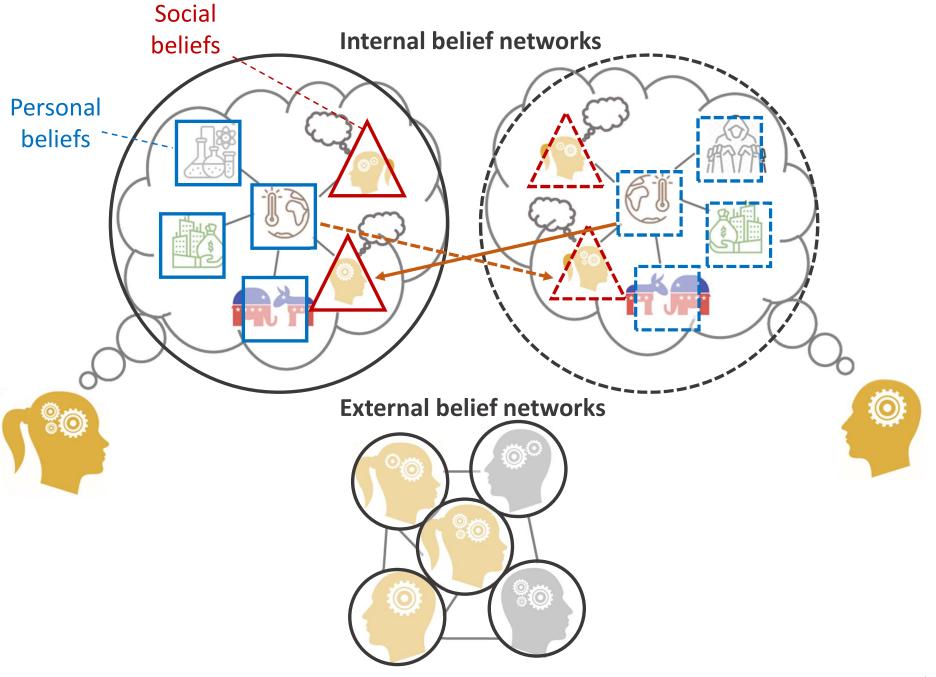


Integrating social and belief networks: Networks of beliefs

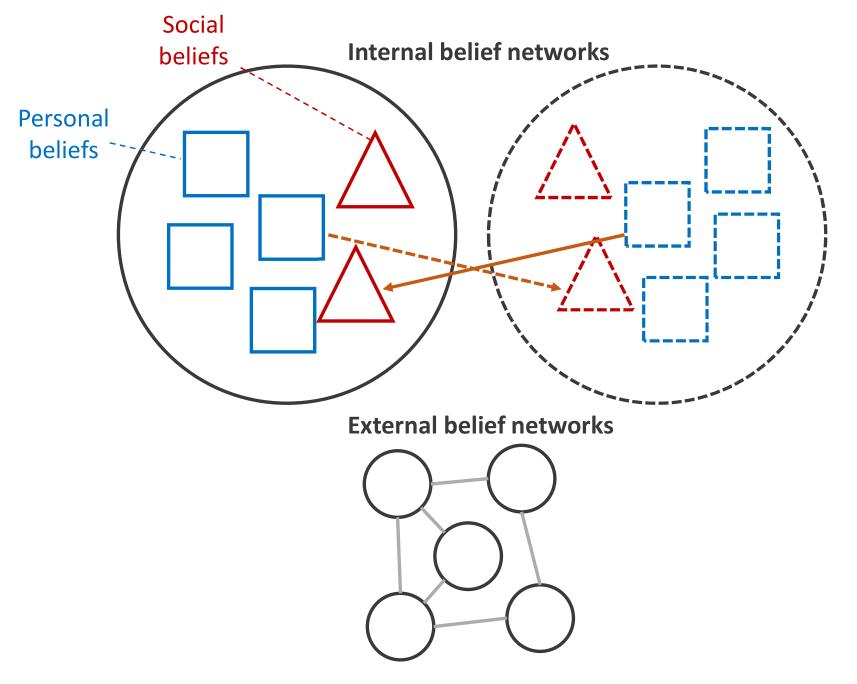


dynamics. <u>https://osf.io/368jz/</u>

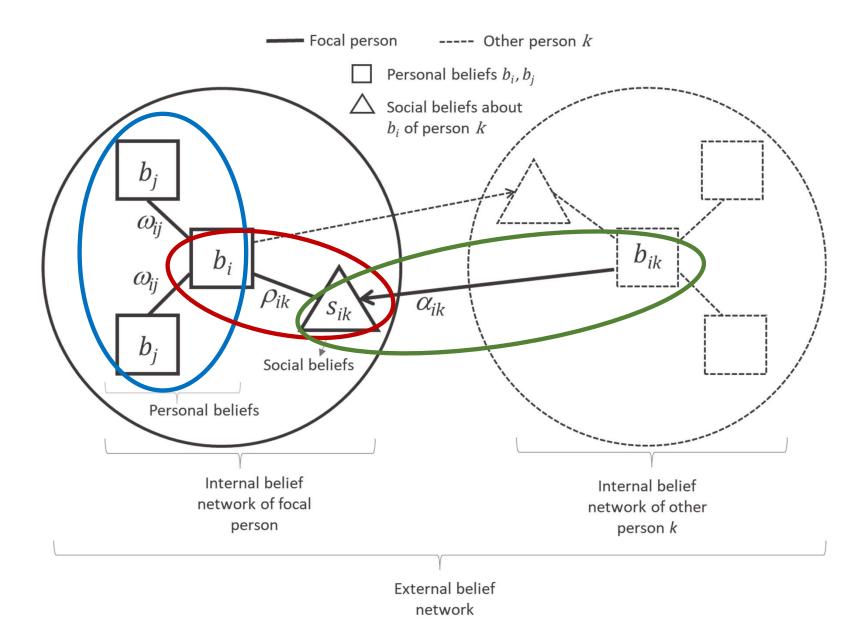
Networks of beliefs: Visual notation



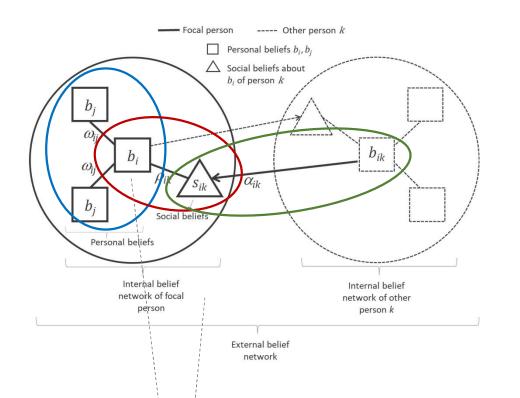
Networks of beliefs: Visual notation



Networks of beliefs: Mathematical notation



Networks of beliefs: Mathematical notation



Updating personal beliefs

$$P(b_i \rightarrow b'_i) \approx 1/1 + e^{\Delta(\beta_{pers}H_{pers} + \beta_{soc}H_{soc})}$$

Updating social beliefs

$$P(s_i \rightarrow s_i') \approx 1/1 + e^{\Delta(\beta_{soc}H_{soc} + \beta_{ext}H_{ext})}$$

Potential dissonances

Personal dissonance

$$H_{pers} = -\sum_{ij} \omega_{ij} b_i b_j$$

Social dissonance

$$H_{soc} = -\sum \rho_{ik} b_i s_{ik}$$

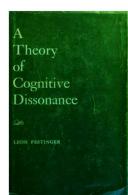
External dissonance

$$H_{ext} = -\sum_{i} \sum_{k} \alpha_{k} s_{ik} b_{ik} ,$$

Felt dissonances

Depend on **attention** to personal, social, and external dissonances

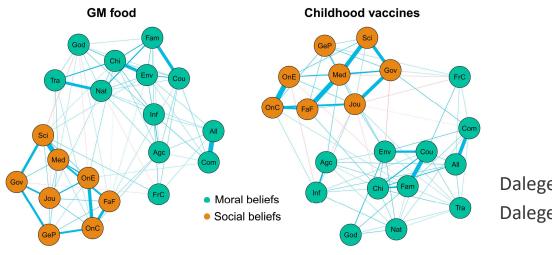
 $\beta_{pers}, \beta_{soc}, \beta_{ext}$



Networks of beliefs: Empirical tests

Data

• Networks of beliefs measured in longitudinal surveys



Dalege & van der Does, 2022, *Science Advances* Dalege, Galesic, & Olsson, 2023, <u>https://osf.io/368jz/</u>

 Networks of topics discussed in comment sections of news sites across the US political spectrum, each month, over 5 to 8 years



BERTopic networks on Gateway Pundit, Jaksic et al., 2023



What we know



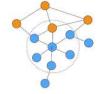
Collective adaptation: Disciplinary silos

socialdge, comparational social since, economics, projections and social cognitions and

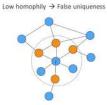
Social environments

Social networks

- Perceived vs. objective networks (Thomas theorem, 1928; Denrell, 2005; Gonzalez et al., 2015)
- Size and connectivity (Derex & Boyd, 2016; Lazer & Friedman, 2007; Mason et al., 2008; Giannoccaro et al. 2018)
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- Directed vs. undirected
- Social artifacts
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- Communication channels
- Institutions
- -----



High homophily → False consensus



Lee et al., 2019; Galesic, Olsson, Rieskamp, 2018

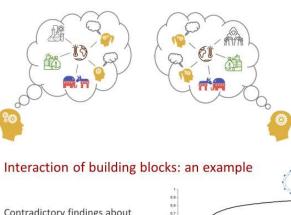
Strategies for integrating social information

Three basic classes:

- Frequency-dependent strategies: majority, plurality, unanimity, minority, complex contagion... rules
 Condorcet 1785; political science (Heinberg, 1932); statistics (Penrose, 1946); psychology (Asch, 1955); economics (Plott, 1967); cultural evolution (Boyd & Richerson, 1985); computer science (Parhami, 1994); statistical physics (Krapivsky & Redner, 2003); biology (King & Cowlishaw, 2007); sociology (complex contagion; Centola & Macy, 2007)
- Averaging strategies: with or without weights, e.g. advice taking, voter model, contagion rules, blending inheritance Galton, 1907; economics (DeGroot, 1974; Golub & Jackson, 2010); advice taking (Molleman et al., 2020; Yaniv, 2004); statistical physics ((sing models; Castellano et al., 2020); cultural evolution (blending inheritance: Boyd & Richerson, 1985); network science (contagion; Newman, 2003)
- Model-based strategies: follow leader, expert, similar,
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cultural evolution (high status: Henrich & Gil-White, 2001), social psychology (liking, authority: Cialdini & Trost, 1998), cognitive psychology (similarity, Wisdom et al., 2013), law (confidence; Penrod & Cutler, 1995)

Different belief networks



Problem environments

Structural properties of problems

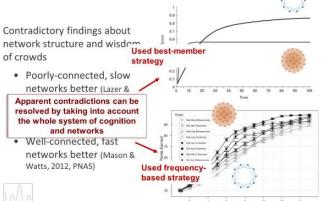
- Categorical vs. continuous judgments
- Simple and complex task landscapes



One-shot and repeated problems

Global environment

 Economic, political, cultural factors that change payoffs of different options, feasibility of different strategies and networks



What we are working on

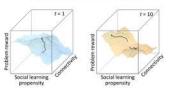
A simple example



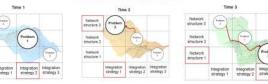
B. Best social learning propensities and connectivity for solving the two problems Social learning Connectivit propensity Problem 1: simple Problem 2: complex - option 1

Problem 2: complex - option 2

C. The resulting problem landscapes



Five implications



Path dependence 1)

- Not one "collective intelligence" (c.f. van der Maas et al., 2006) 2)
- 3) Multi-task satisficing
- 4) Collective myopia
- 5) Collective imagination: Simulating long jumps



Analogies for modeling collective adaptation

- · No analogy is completely correct but some are useful
- Adaptive landscapes
 - Easy to understand and flexible, but can lead to wrong intuitions for high-dimensional, constantly changing spaces (Gavrilets, 2004; Agarwala & Fisher, 2019, Fragata et al., 2019)

Cultural evolution

From tools to institutions, typically no network structure (but see 5molla & Akcay, 2019)

Statistical physics

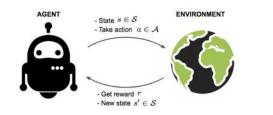
Reducing dissonance on individual and collective level, no meta-level rules for switchin strategies and structures when problems change

· Reinforcement learning

- Of social learning strategies (Ha & Jeong, 2022), of networks - Of individual welfare functions (Wolpert & Tumer, 2001)
- Ecosystem modeling
- Dynamic interaction network
- Combinations of analogies
 - Different analogies for different building blocks, for example:
 - Drift diffusion model of learning + an evolutionary process of adaptation to different groups structures and costs of errors (Tump et al., 2022)
 - Epidemiological + evolutionary models to study evolution on sociality on different time scale (Cooriey et al., 2022)

1. Co-adaptation of social cognitions and social networks

Modeling: Reinforcement learning of suitable social cognitions and social networks



Esternal ballet retwork Updating personal beliefs

Updating social beliefs $P(s_i \rightarrow s_i') \approx 1/1 + e^{\Delta(\beta_{soc}H_{soc} + \beta_{ext}H_{ext})}$

Networks of beliefs: Mathematical notation

Social dissonance

 $H_{soc} = -\sum \rho_{ik} b_i s_{ik}$

```
H_{ext} = -\sum_{i} \sum_{\nu} \alpha_k s_{ik} b_{ik},
```

Depend on attention to personal, social, and external dissonances

Bpers, Bsoc, Bext

Dalege, Galesic, & Olsson, (2023)

Networks of beliefs: Empirical tests

Data

Networks of beliefs measured in longitudinal surveys



· Networks of topics discussed in comment sections of news sites across the US political spectrum, each month, over 5 to 8 years



BERTopic networks on Gateway Pundit, Jaksic et al., 2023

0.0

ps://illianweng.github.ic/posts/2018-02-19-rl-overview/

Potential dissonances Personal dissonance $H_{pers} = -\sum_{ij} \omega_{ij} b_i b_j$





$P(b_i \rightarrow b'_i) \approx 1/1 + e^{\Delta(\beta_{pers}H_{pers} + \beta_{soc}H_{soc})}$

What we hope to understand

- 1. How do collectives change their integration strategies and network structures to adapt to different problems?
- 2. How do radically different beliefs about what problems are important affect collective adaptation?
- 3. Why is it sometimes hard for collectives to reach seemingly obvious solutions to a particular problem?
- 4. Can we anticipate new problems that might emerge because of the way societies adapted to past problems?
- 5. Can we reduce less desirable consequences of collective adaptation to emerging problems?