# Social learning strategies Collective problem solving

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# Collective problem solving

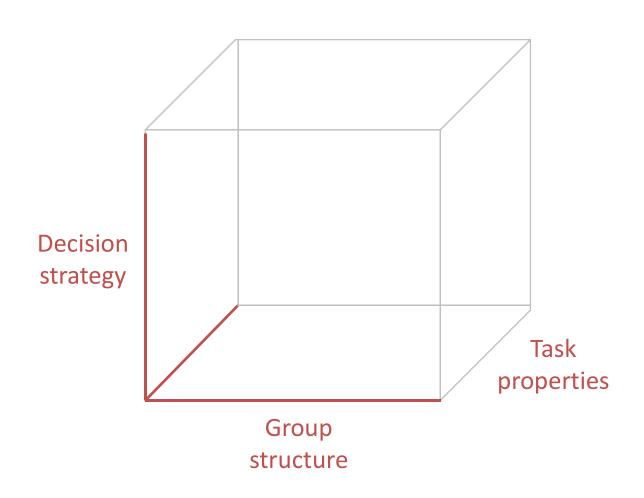




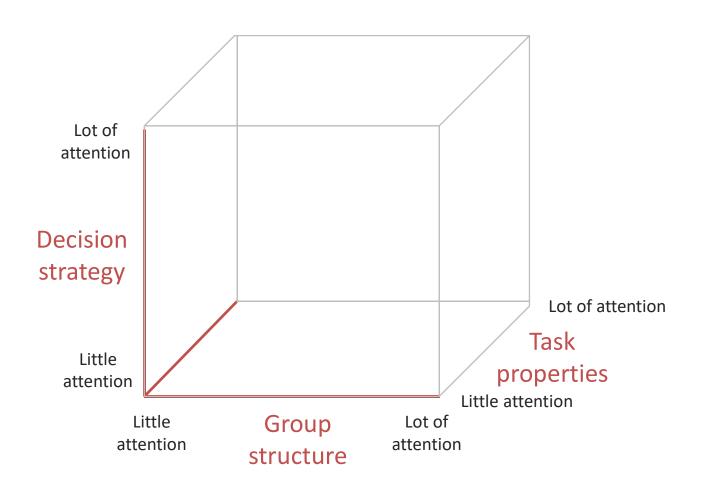




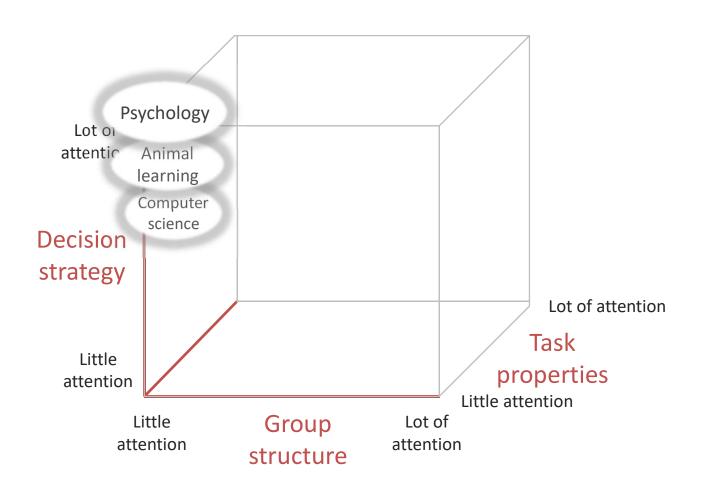
# Dimensions of collective problem solving



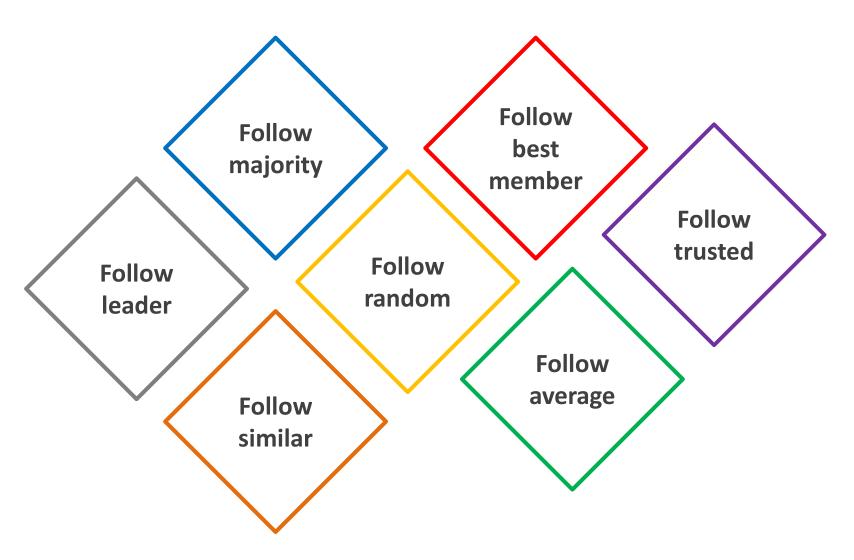
# Views of different disciplines



# Views of different disciplines



# Group decision strategies



## ... in social psychology

- The dark side of conformity
  - Asch (1955) and Milgram (1963) experiments



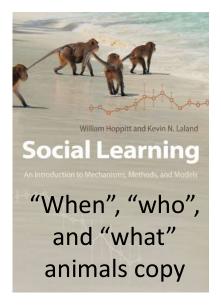


- And the good sides...
  - Social norm interventions





# ... in animal learning





Van de Waal, 2013, Science

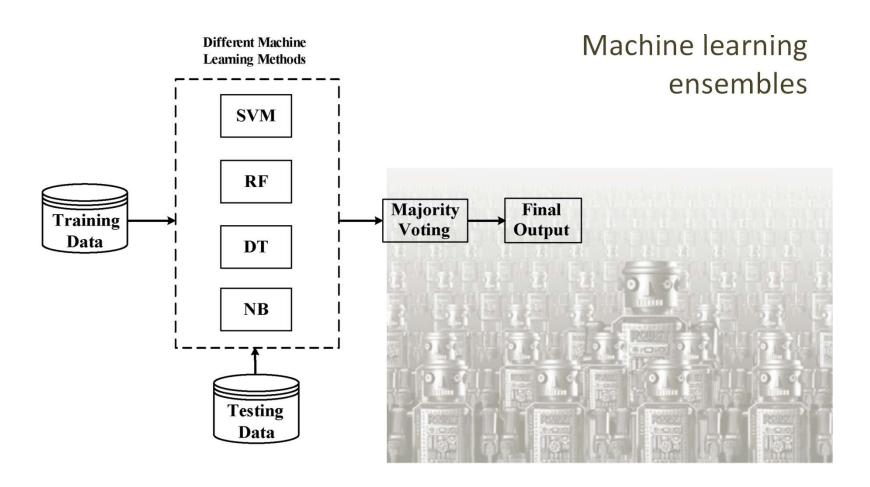


Gunhold, Animal Behavior, 2014



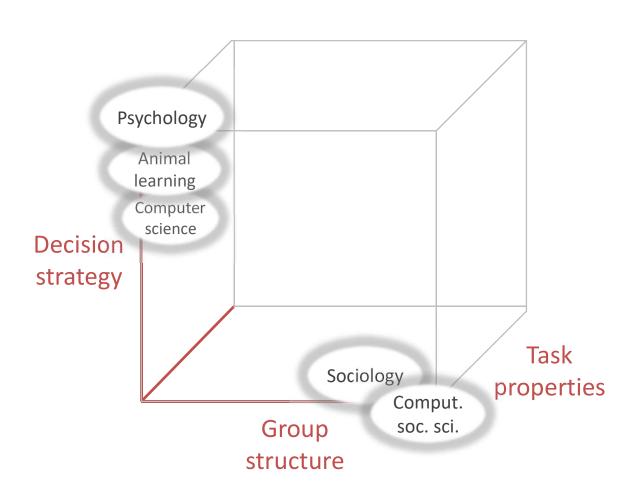
Aplin, 2015, Nature

# ... in computer science

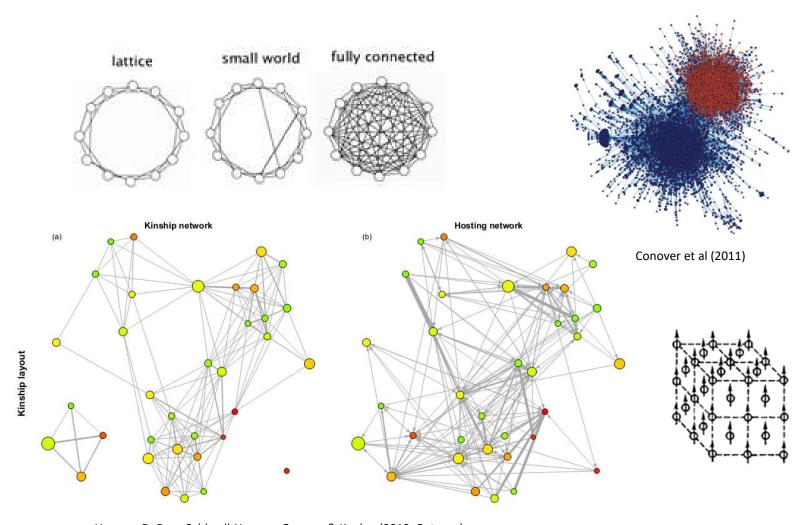


Saha et al, 2014; Rokach, 2012

# Views of different disciplines



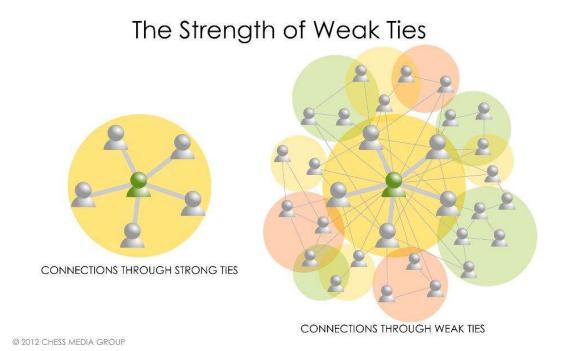
# Group structure



Hooper, DeDeo, Caldwell-Hooper, Gurven, & Kaplan (2013, Entropy)

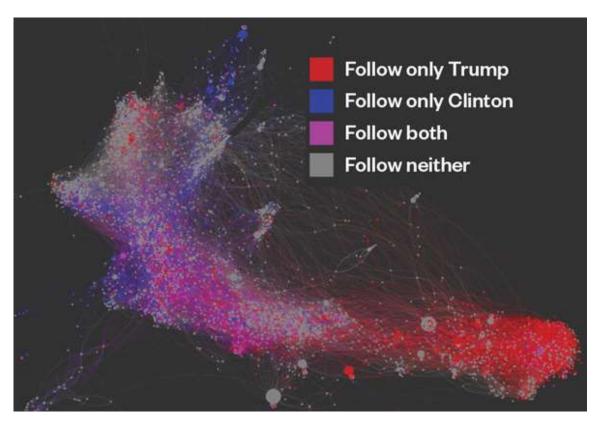
# ... in sociology

- Mechanisms of social contagion
  - Granovetter (1973, 1978), Centola (2007, 2010)



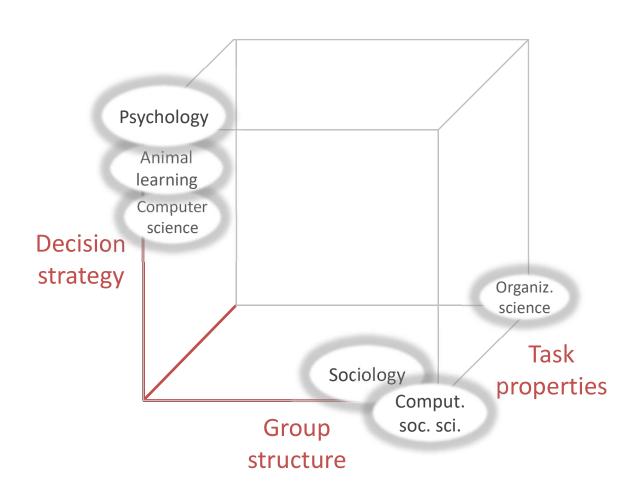
## ... in computational social science

Reconstructing social networks from big data

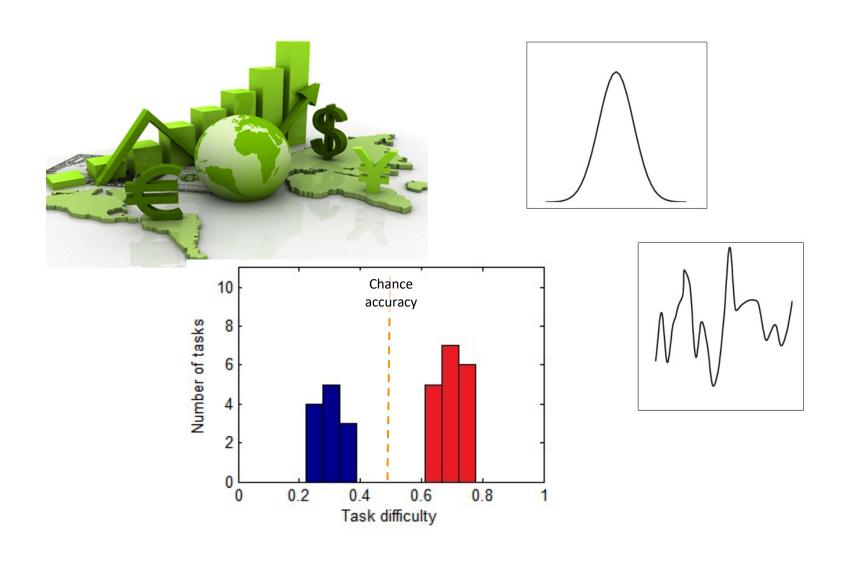


http://www.electome.org/, Twitter+MIT

# Views of different disciplines

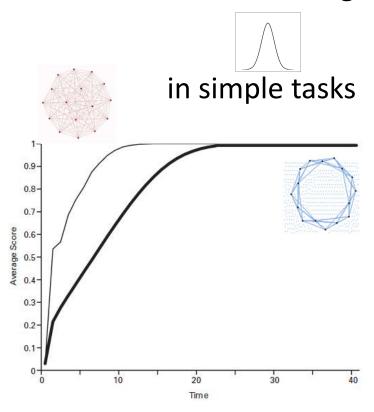


# Task properties

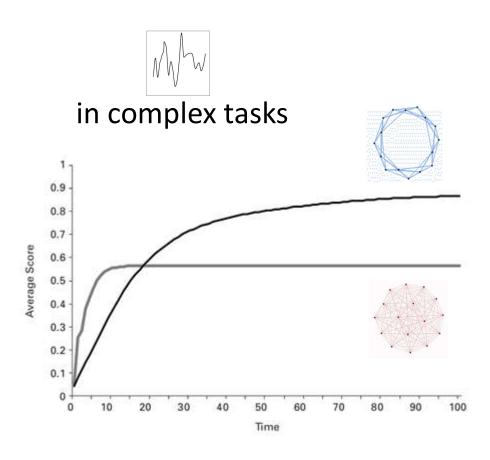


# ... in organization science

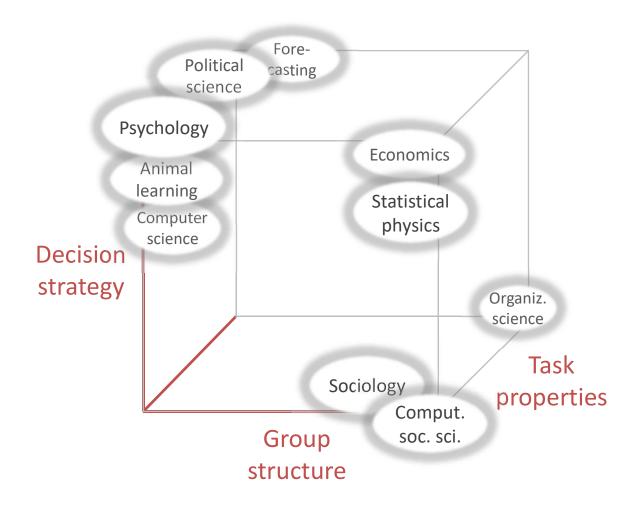
Performance of different group structures



Lazer & Friedman, 2007



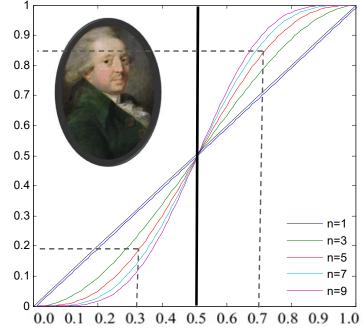
# Views of different disciplines



#### Political science

- Voting models for aggregation of preferences and information (Dewan & Shepsle, Annu Rev Polit Sci, 2011)
- Example: Condorcet Jury Theorem





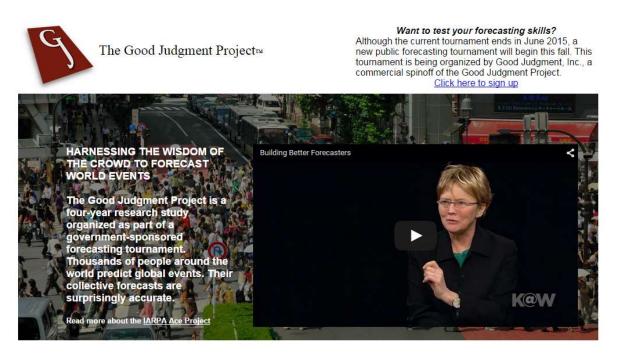
$$M = \sum_{i=m}^{n} {n \choose i} p^{i} (1-p)^{n-i}$$

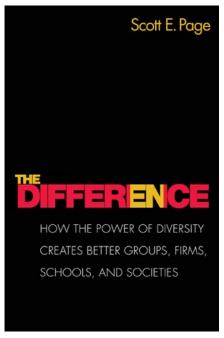
Grofman, Owen, & Feld (1983, Th Dec). 13 Theorems in Search of the Truth.

List & Goodin (2001, J of Pol Phil)

#### Forecasting

When and why groups perform better than individuals? How to elicit and aggregate expert forecasts?





#### **Economics**

#### Opinion formation on networks

Bayesian updating models

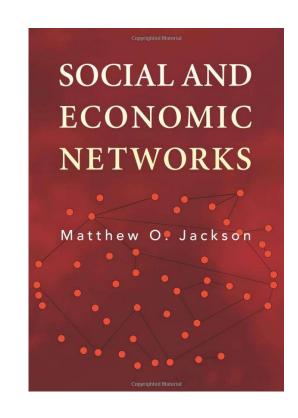
$$P(\theta|s) = \frac{P(s|\theta)P(\theta)}{P(s)}$$

$$P(\theta) - \text{prior belief}$$

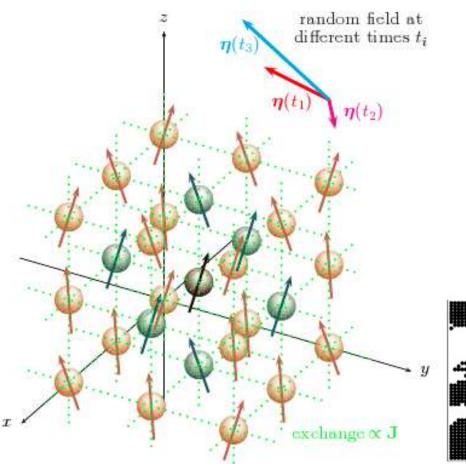
$$s - \text{social signal}$$

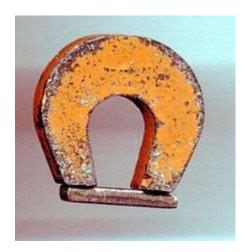
- Non-Bayesian models
  - DeGroot model

$$x_{i,t+1} = \sum_{j=1}^{n} T_{ij} x_{j,t}$$
  $x_{i}$  belief of agent  $i$   $T_{ij}$  – trust of  $i$  in  $j$ 's signal

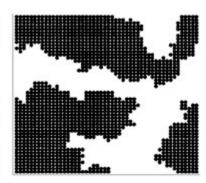


# Statistical physics

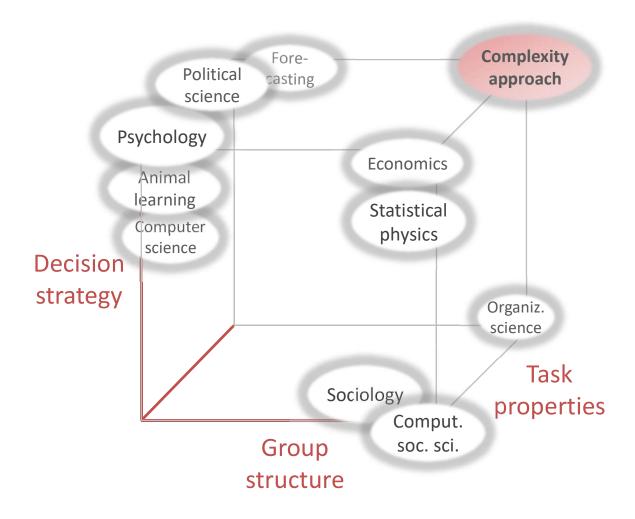








# Need for complexity approach



### Need for complexity approach

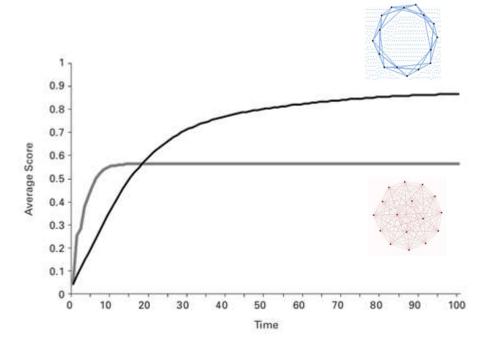
# Example 1: Which networks are best for collective problem solving?

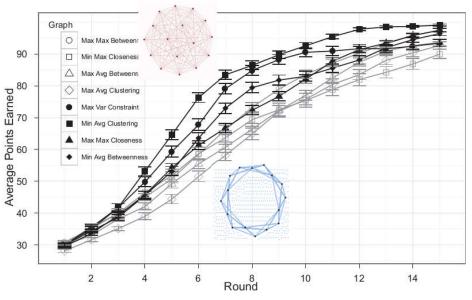




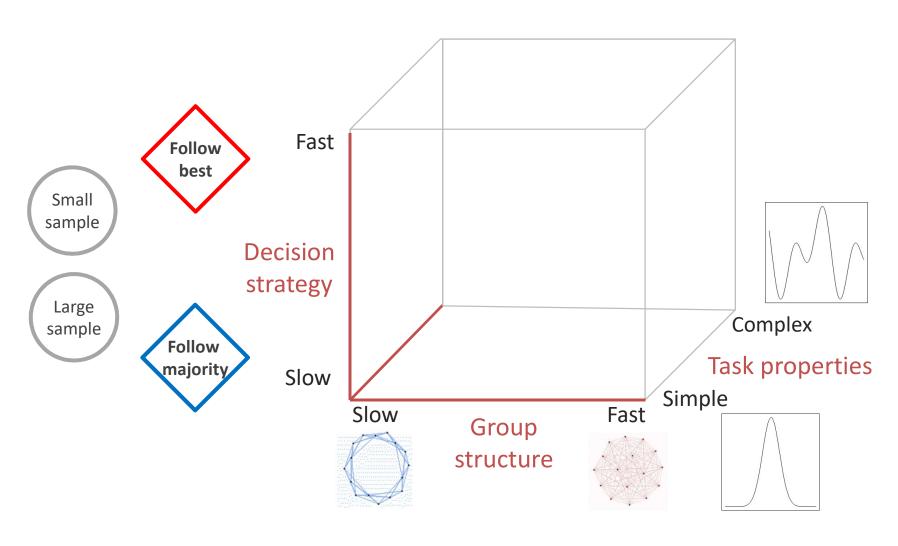
#### Studies disagree:

- Poorly-connected, slow networks better (Lazer & Friedman, 2007, ASQ; Derex & Boyd, 2016, PNAS)
- Well-connected, fast networks better (Mason & Watts, 2012, PNAS)
- → Studies focused on group structure; disregarded decision strategies and task properties





# Dimensions of collective problem solving



### Computational study

100 agents, in different networks, solve simple or complex tasks over 200 time steps:

Step 1. Imitate another agent, if better

Choice of agent depends on decision strategy

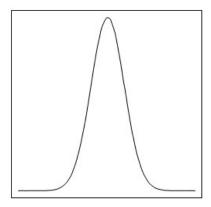
Step 2. Otherwise, explore alone

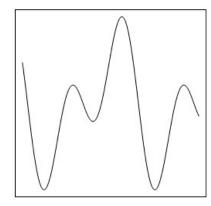
Record average payoff of agents on each time step.

### Task properties

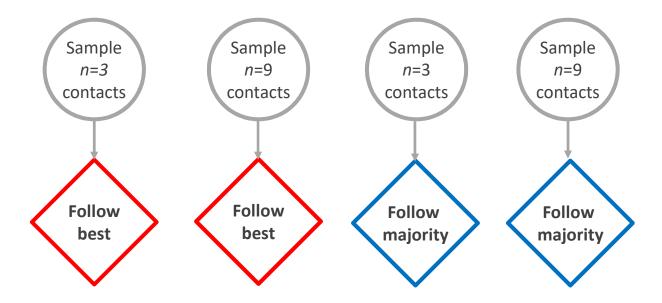
#### NK environments:

- Solutions are strings of N elements (000101001)
- Fitness of each element depends on K other elements
- N = 15
- K varies from 0 ("Simple environment") to 14 ("Complex environment")



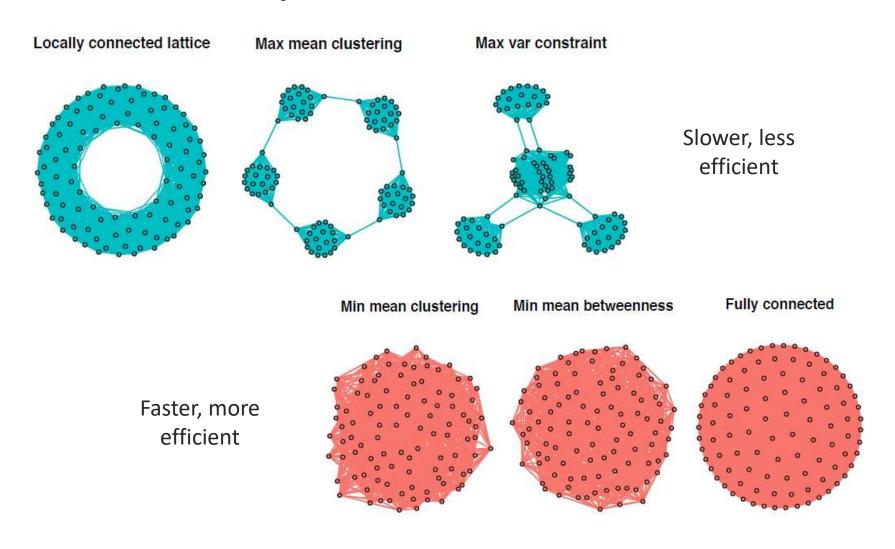


## Decision strategies



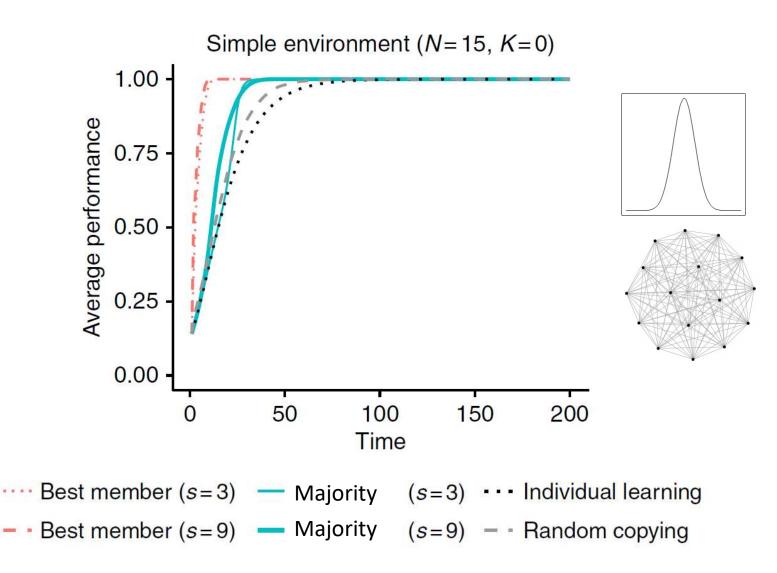
+ for comparison: random copying and individual learning

# Group structure - Networks

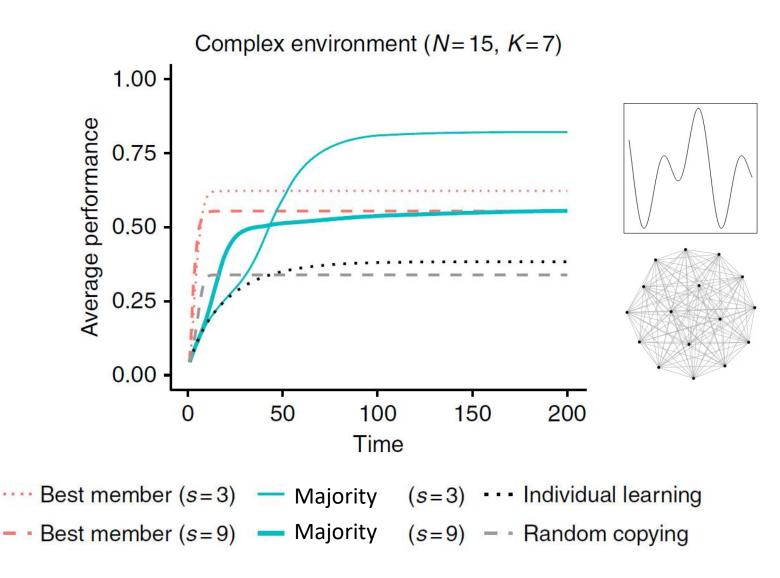


# Results

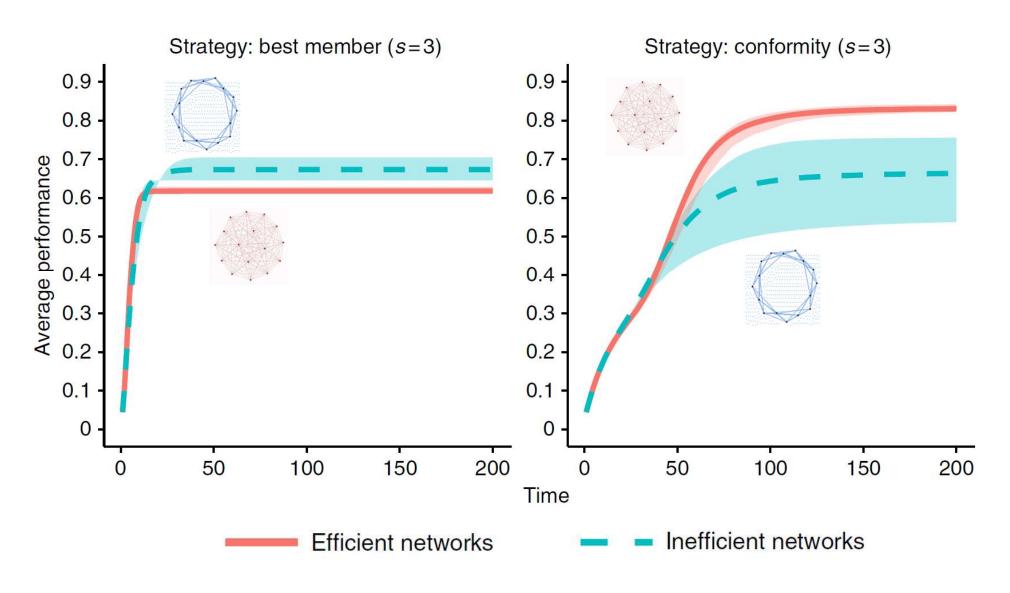
#### Fast strategy good when task simple



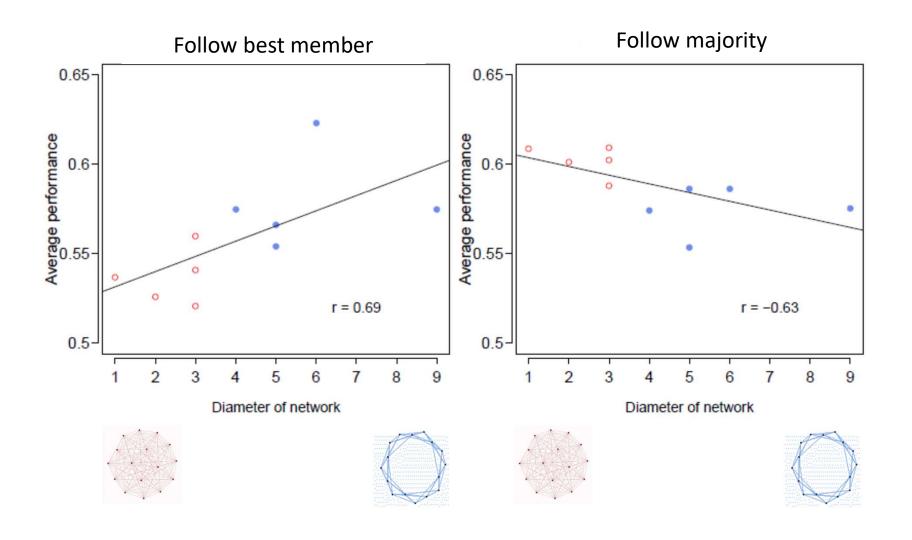
### Slow strategy good when task complex

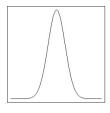


#### Fast networks better with slow strategies

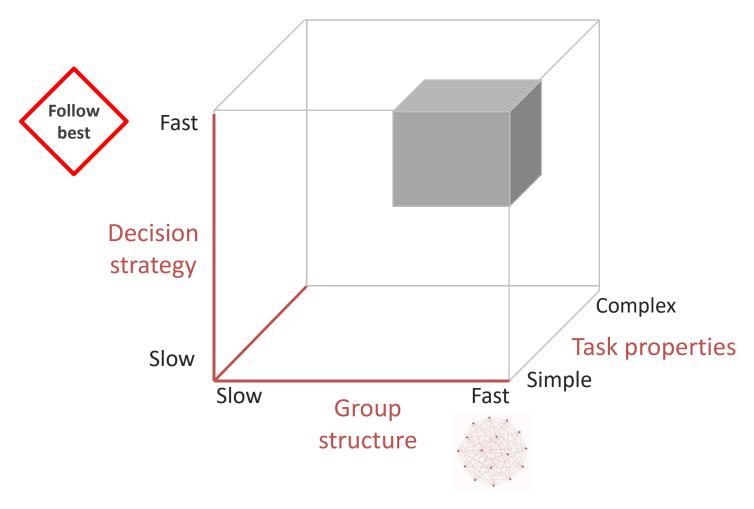


# Fast networks better with slow strategies



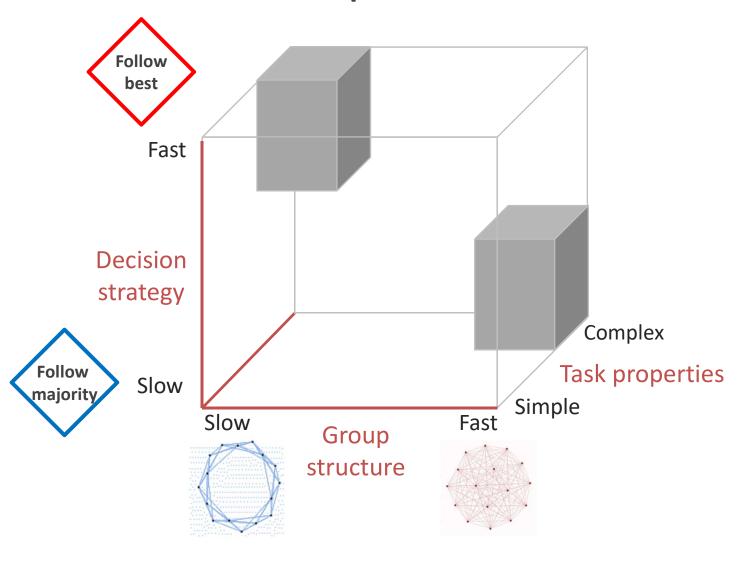


# Simple tasks





# Complex tasks



## **Implications**

- There is no one best group structure
- When designing an organization, it matters what are the desired (or available):
  - 1. Network/communication structures
  - 2. Decision making procedures/strategies
  - 3. Task properties

## Need for complexity approach

Example 2: What is the best size of committees that need to make many decisions over time?

## Typical committees



http://www.federalreserve.gov/



www.med.upenn.edu/criticalcare/



www.rhuddlantowncouncil.gov.uk/

Galesic, Barkoczi, & Katsikopoulos (2016), Decision

#### Typical committee sizes

Jury sizes in most countries: 6-15 people

Town councils in UK and Australia: 5-30

Parliamentary committees in US, EU, Australia: 20-40

US House and Senate subcommittees: 10-15

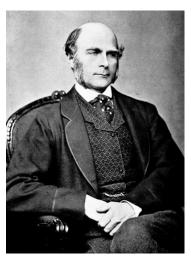
Central bank boards: 5-12

Number of close friends: 6 or less

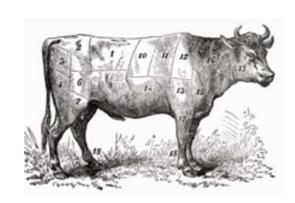
Number of online reviews read: average 5, max 30

If there is "wisdom of crowds", why are committees so small?

#### Wisdom of crowds









average of 800 guesses = 1,197 actual weight of the ox = 1,198

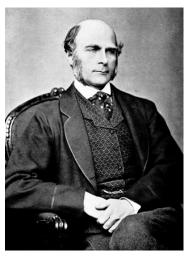
When a group estimates a quantity by averaging individual guesses

→ Larger group will always have smaller error (as long as it is diverse)

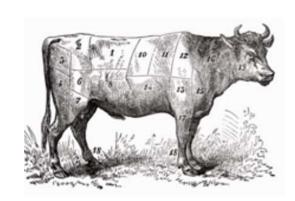
$$MSE = \overline{\text{bias}}^2 + \frac{1}{M}\overline{\text{var}} + \left(1 - \frac{1}{M}\right)\overline{\text{cov}} + \sigma_Y^2$$

→ Total error = bias + variance + covariance + irreducible error (M=group size)

#### Wisdom of crowds









average of 800 guesses = 1,197 actual weight of the ox = 1,198

When a group estimates a quantity by averaging individual guesses

→ Larger group will always have smaller error (as long as it is diverse)

But when a group needs to decide on one of several courses of action

\( \rightarrow \) Larger group is not always better

### Group error components

$$MSE = \overline{\text{bias}}^2 + \frac{1}{M} \overline{\text{var}} + \left(1 - \frac{1}{M}\right) \overline{\text{cov}} + \sigma_Y^2$$

### Typical committee tasks

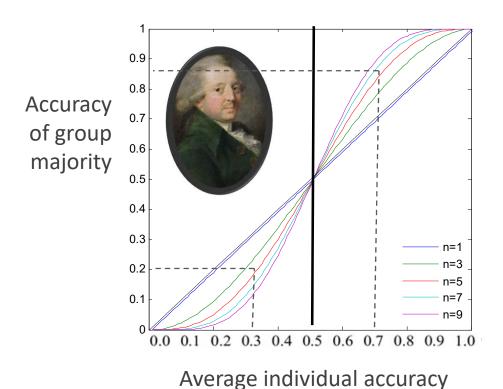
- Group of experts vote by simple majority to make decisions and predictions such as:
  - Will the economy grow or fall in the next period?
  - Which policy should we adopt?
  - Should we make this investment or not?
  - What is patient's diagnosis?
  - Should we attack or not?





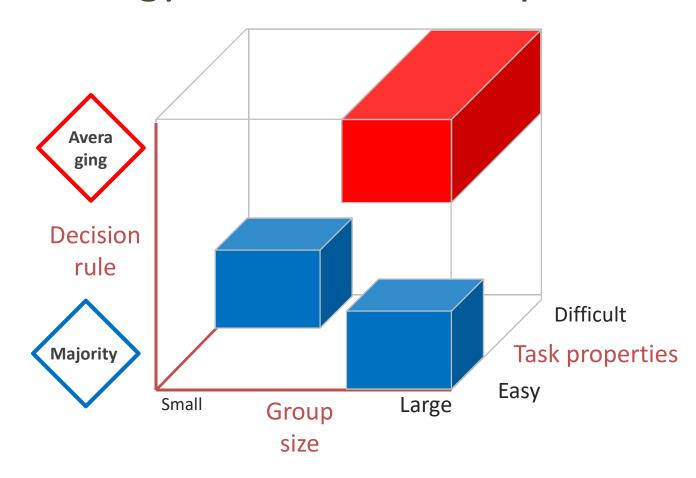
# Accuracy of simple majority rule for a single task

#### Condorcet Jury Theorem



$$M = \sum_{i=m}^{n} {n \choose i} p^{i} (1-p)^{n-i}$$

# Best group size depends on decision strategy and task difficulty

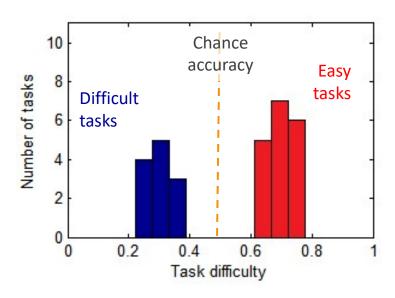


### Accuracy over many tasks?

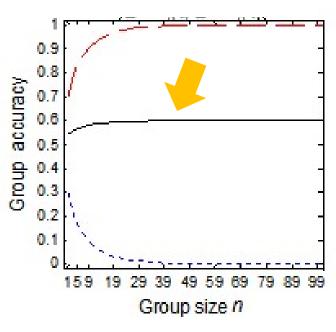
- Committees encounter many tasks over the course of their existence
  - Some tasks are easy
  - Some tasks are surprisingly difficult
  - We can't know in advance how easy or difficult the next task is going to be

# Accuracy of simple majority rule across many tasks

"Neutral" task environment



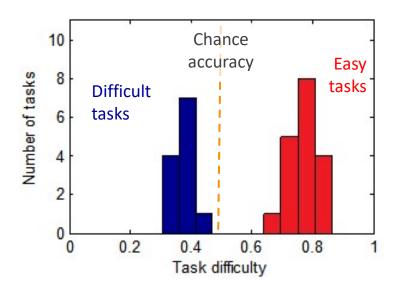
Proportion of easy tasks: e = 0.6



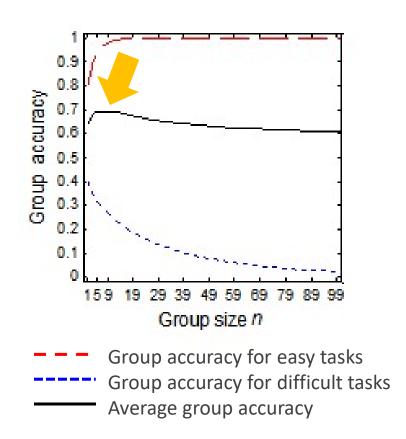
Group accuracy for easy tasks
Group accuracy for difficult tasks
Average group accuracy

# Accuracy of simple majority rule across many tasks

"Friendly" task environment

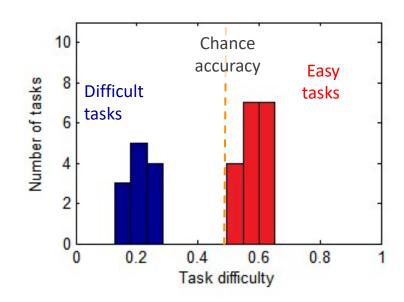


Proportion of easy tasks: e = 0.6

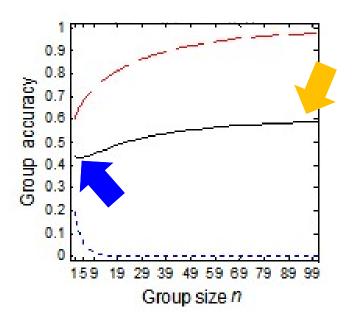


# Accuracy of simple majority rule across many tasks

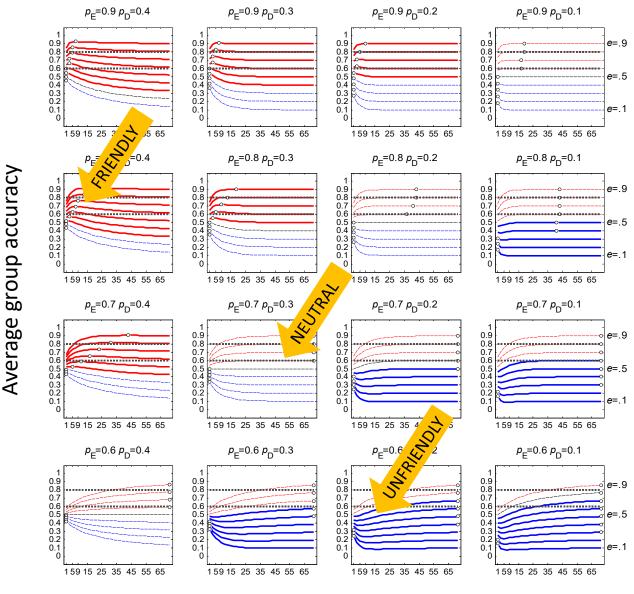
"Unfriendly" task environment



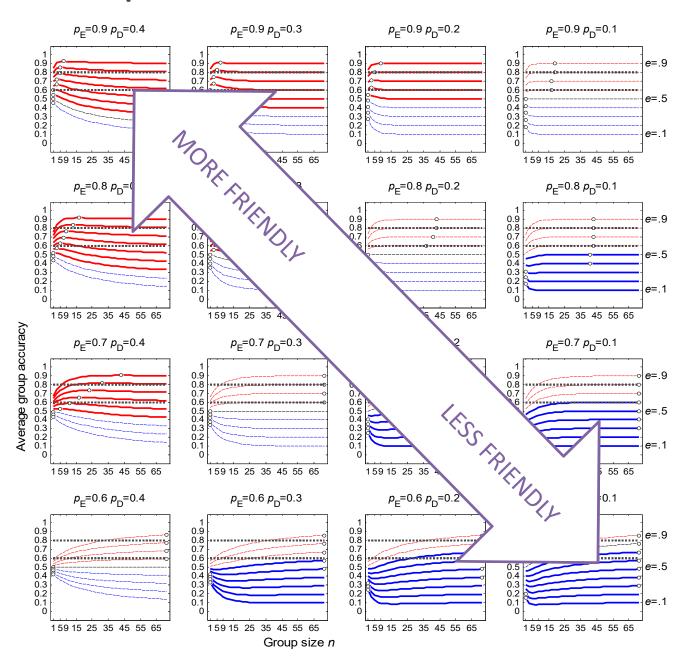
Proportion of easy tasks: e = 0.6

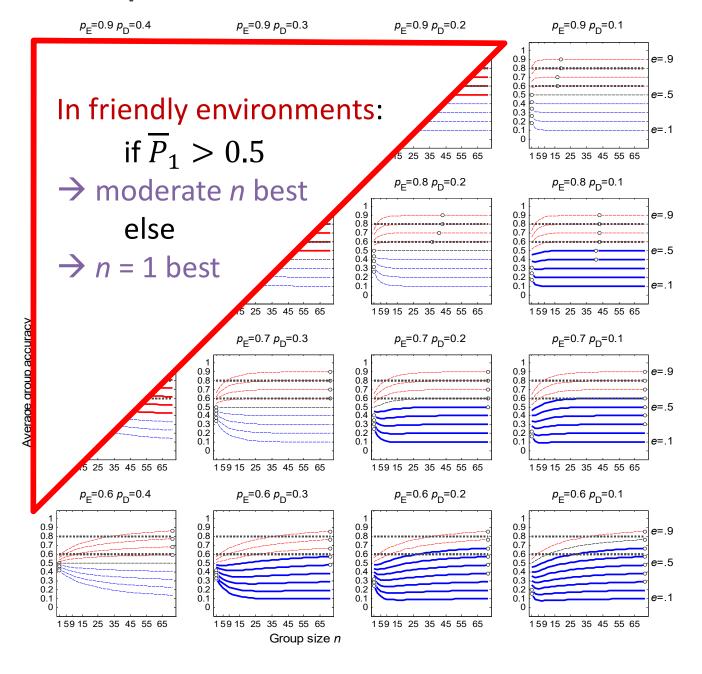


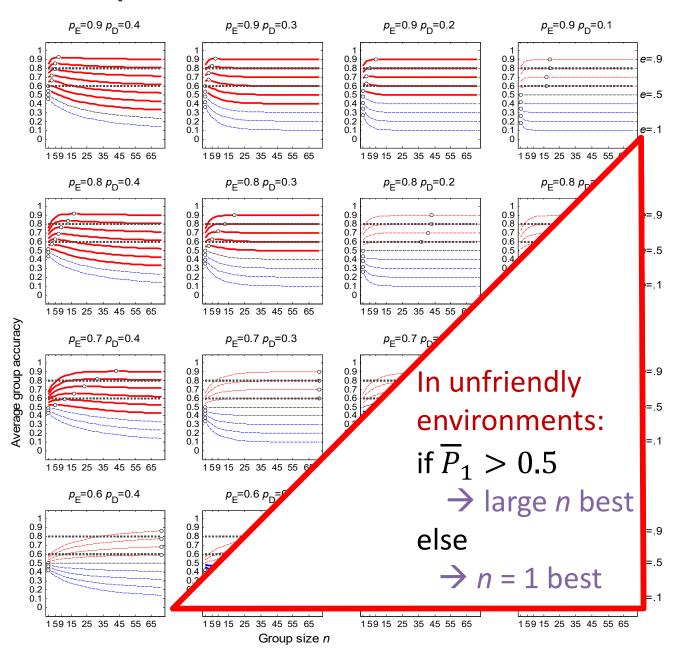
Group accuracy for easy tasks
Group accuracy for difficult tasks
Average group accuracy



Group size *n* 







## More formally

Average group accuracy across many tasks:

$$\overline{P}_n = e P_{E,n} + (1 - e) P_{D,n}$$

 $\bar{P}_n \rightarrow$  average accuracy of group of size n across tasks  $e \rightarrow$  proportion of easy tasks that group needs to solve  $P_{\mathrm{E},n}\left(P_{\mathrm{D},n}\right) \rightarrow$  accuracy of group of size n on easy (difficult) tasks

Friendly environment:

$$\overline{p}_{\rm E} - 0.5 > 0.5 - \overline{p}_{\rm D} \quad \Rightarrow \quad \overline{p}_{\rm E} + \overline{p}_{\rm D} > 1$$

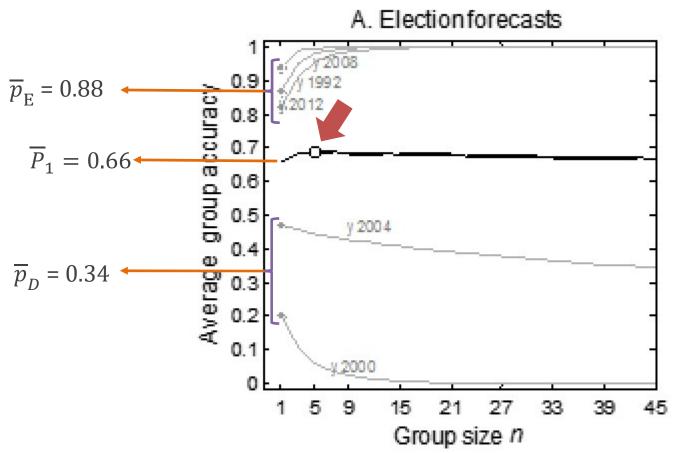
 $\overline{p}_{\rm E} \, (\overline{p}_{\rm D}) o$  individual accuracy on easy (difficult) tasks

Moderate group sizes better whenever friendly environment and  $\overline{P}_1 > 0.5$ 

# Is the real world friendly or unfriendly?

#### Real-world task environments: Elections

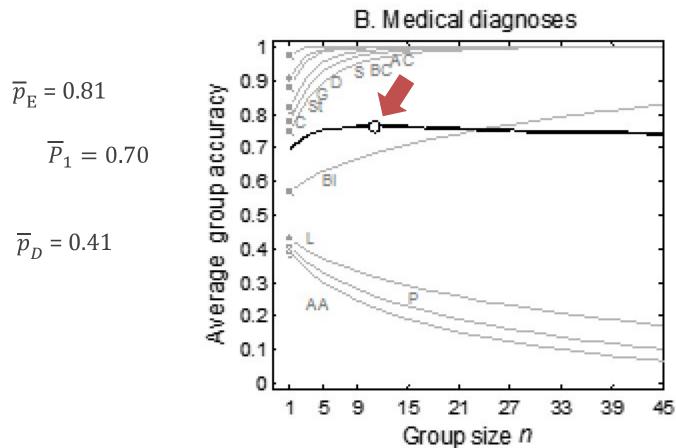
(Graefe, 2014)



 $ightarrow \overline{p}_{\rm E} + \overline{p}_{\rm D} > 1$  and  $\overline{P}_1 > 0.5$  (friendly environment, and average expert more accurate than chance across tasks)

#### Real-world task environments: Medicine

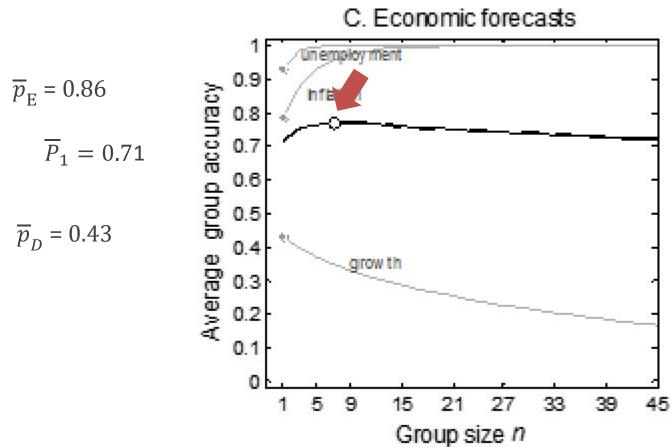
(Schiff et al, 2009)



 $ightarrow \overline{p}_{\rm E} + \overline{p}_{\rm D} > 1$  and  $\overline{P}_1 > 0.5$  (friendly environment, and average expert more accurate than chance across tasks)

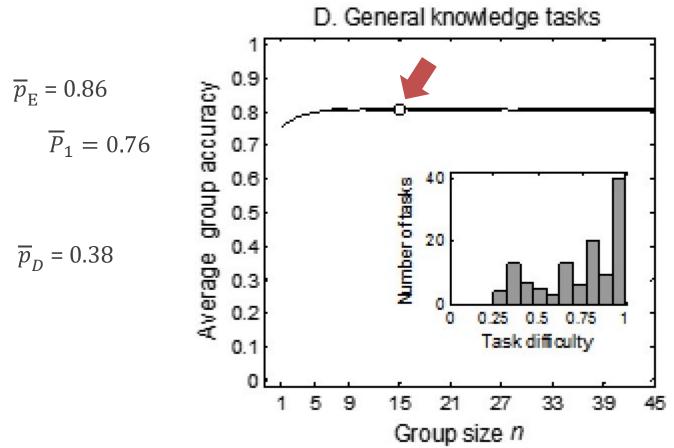
#### Real-world task environments: Economics

(Hilsenrath & Peterson, 2013)



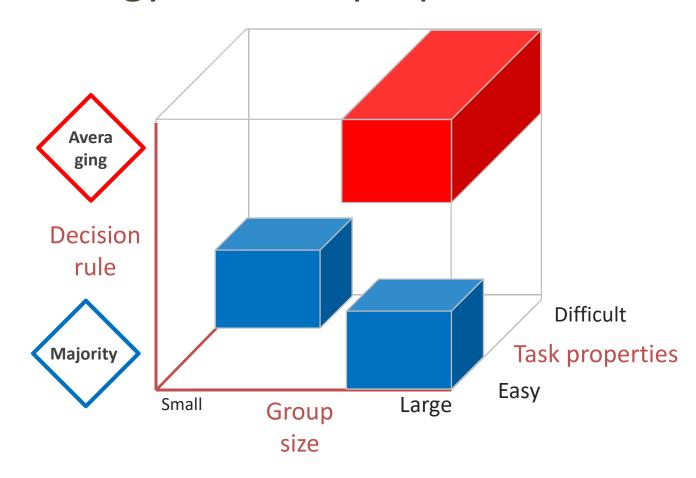
 $ightarrow \overline{p}_{\rm E} + \overline{p}_{\rm D} > 1$  and  $\overline{P}_1 > 0.5$  (friendly environment, and average expert more accurate than chance across tasks)

## Real-world task environments: General knowledge (Juslin, 1997)

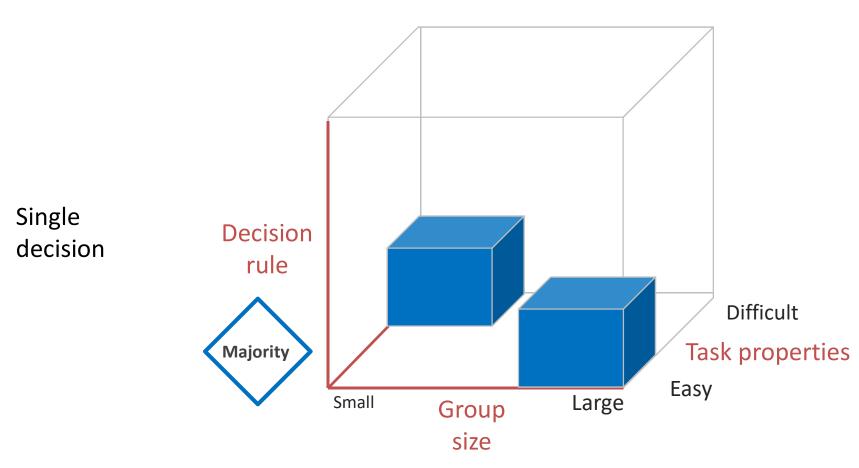


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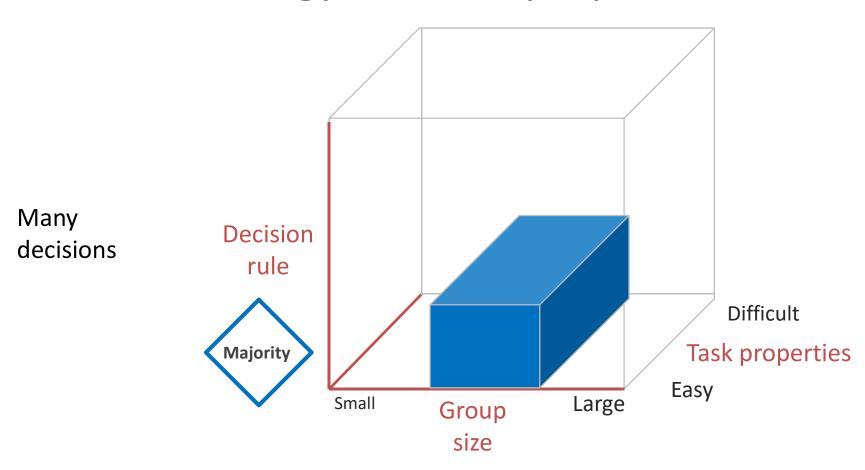
# Best group size depends on decision strategy and task properties



# Best group size depends on decision strategy and task properties



# Best group size depends on decision strategy and task properties



## **Implications**

- When designing committees, larger is not always better
- If there is a chance of surprising mistakes,
   5-9 member-team can be as good as or better than larger teams

#### In sum: Collective problem solving as a complex system

