Majority-rule opinion dynamics with differential latency

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Outline

Majority-Rule Models

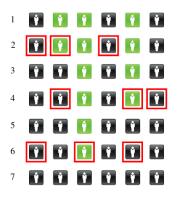
Differential Latecy

Experiments

Discussion and Conclusions

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Basic Model



Initial population configuration

Random team formation

Result after applying the majority rule

Random team formation

Result after applying the majority rule

Random team formation

Result after applying the majority rule

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Figure: The basic majority rule model

Dynamics

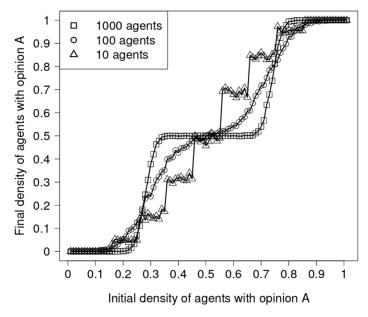
- The model will converge to the majority opinion
- As the number of agents increases the probability of converging to the original majority opinion increases for smaller and smaller majorities.

Model with latency

- After changing state, agents become *latent*.
 - Latent agents can influence the opinion of others, but cannot change their opinion (ie. can't be influenced)
- Each time a latent agent is selected to be part of a team, the agent becomes non-latent with probability λ
- ► For some λ it is possible for the proportion of agents favoring a particular opinion to fluctuate randomly, and never converge.

called "zero magnetization"

Dynamics with latency ($\lambda = 1/20$)



Motivation

- A swarm of robots may need to select some action to take
- It may be desirable for all robots to select the same swarm
- Bound the probability that the robots will converge to a particular opinion given some environmental conditions

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- Motivating examples
 - Use search algorithm A or algorithm B
 - Do or do not use task partitioning

Model

The extended model is influenced by the embodied nature of robots

- Latency is directly mapped to the time taken for a robot to perform an action.
- Latent agents are excluded during team formation
 - this means that they cannot be influenced by or *influence* the opinions of other agents.

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 Agents always go in to a latent state when their team makes a decision (since they go perform an action).

Exponential distribution dynamics

- the dynamics can be modeled as a system of differential equations,
 - one for the fraction of agents with opinion A
 - one for the number of agents with opinion A

$$\dot{f} = -rp + a \tag{1}$$
$$\dot{a} = -a + r(3p^2 - 2p^3)$$

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Figure: Dynamics of the system with exponentially distributed latency

Opinion dependent latency

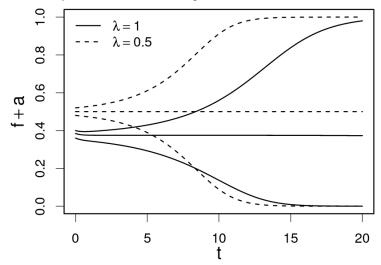


Figure: Despite starting as the minority opinion, because A has lower latnecy the population converges to opinion A (top solid line)

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Exponentially distributed latency

- the model is analyzed as a system of differential equations
- It is shown that under exponentially distributed latency periods, agents will always reach consensus.
- It is possible that the consensus may be the opinion that was the in the minority initially, if the alternative opinion has longer latency.

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Normally distributed latency

- The latency for each opinion is normally distributed mean/std. deviation for each opinion.
- The model with normally distributed latency periods is analyzed using a Monte Carlo simulation.

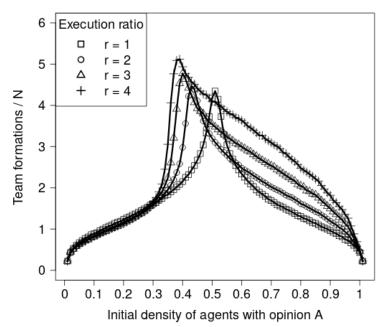
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Normal distribution results

- It is more likely that consensus will be on the opinion having the lowest mean latency. (Figure 6(c,e))
- Increasing the number of teams generally increases the likelihood of reaching consensus on the opinion with shortest latency.
- The overlap between mean latency effects the probability that agents will reach a consensus.
 - In general the system will reach a consensus regardless of the amount of overlap between the latency distributions

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Number of team formations



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Robot experiment

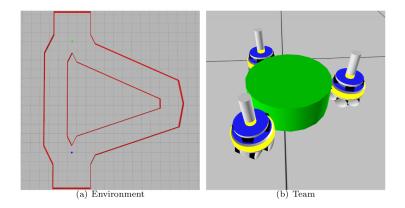


Figure: Collective transport with two alternative paths

video

Experiment Results

- Robots reach a consensus on the shorter path
- not perfect due to collisions forcing some robots to take the longer path on the return trip

Benefits to this approach

- Can be accomplished using only local sensing/communication.
- Can be thought of as roughly equivalent to pheromones,
 - but simpler to implement in real life
- Does not rely on measurement of distances or time, latency periods are implicitly measured by the rate at which agents become non-latent.
- Decision making is a continuous process; however, after a consensus has been reached decision making stops
 - cannot adapt to a changing environment.

Conclusion

- May be time to depart from biological inspiration.
- At the very least there are many effective algorithms to be discovered from an engineering point of view.