Heterogeneity and Robustness in Social Learning

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Heterogeneity in Biological & Human Systems

Threshold Response Models





Division of Labour



Behaviour Types

Different





Social vs Individual Learning

Individual Learning



Rodin: Thinker



Raphael: The School of Athens

Waggle Dance



Kilobots

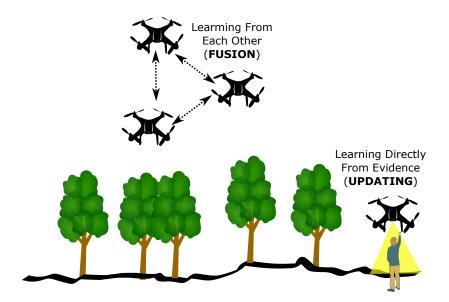


& Consensus

Best-of-N Problem

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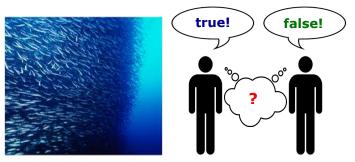
The Two Processes View of Social Learning



True, False & Uncertain: A Route to Consensus

- Consider a simple social learning problem where a population must determine whether a hypothesis H is true or false.
- At any time each individual is in one of three belief states: true (t), uncertain (u) or false (f).

Disagreement Leads to Doubt

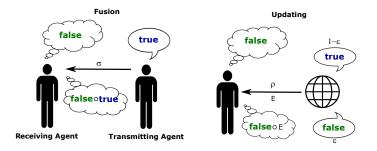


Undecided Individuals Help Consensus in Fish Schools

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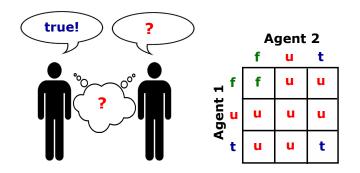
A Simple Two Process Model

- The aim is to learn whether a hypothesis is true or false.
- We assume that the hypothesis is in fact true
- Key parameters: σ (fusion rate), ρ (evidence rate), ϵ (noise).

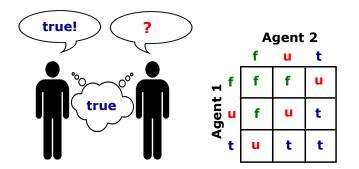


- P_t = (P_t(f), P_t(u), P_t(t)) denotes the proportion of agents in each belief state at time t
- Difference Equation: $\mathbf{P}_{t+1}^T = U(F^{\mathbf{P}_t}\mathbf{P}_t^T)$

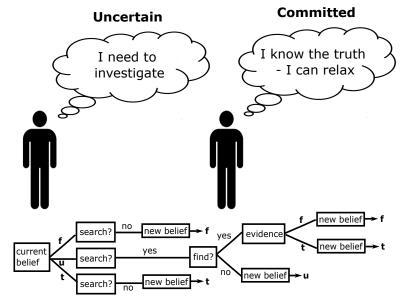
Uncertainty dominates over certainty so that if a receiving agent with committed belief state \mathbf{t} or \mathbf{f} interacts with a transmitting agent with uncertain truth state \mathbf{u} they will abandon their committed position and change their belief to \mathbf{u}



Certainty dominates over uncertainty so that if the receiving agent with belief state \mathbf{u} interacts with a transmitting agent with committed belief states \mathbf{t} or \mathbf{f} , they adopt the latter.

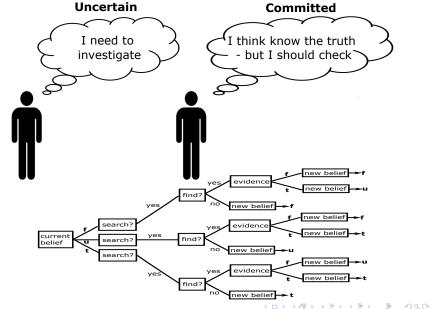


Behaviour Types: Confident Updating



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Behaviour Types: Inquisitive Updating



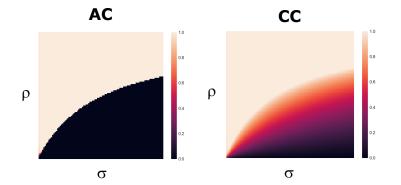
Four Behaviour Types for Social Learning

Cautious:
$$F_{C}^{\mathbf{P}} = \begin{pmatrix} 1 - \sigma + \sigma P(\mathbf{f}) & 0 & 0\\ \sigma(1 - P(\mathbf{f})) & 1 & \sigma(1 - P(\mathbf{t}))\\ 0 & 0 & 1 - \sigma + \sigma P(\mathbf{t}) \end{pmatrix}$$
Adventurous:
$$F_{A}^{\mathbf{P}} = \begin{pmatrix} 1 - \sigma P(\mathbf{t}) & \sigma P(\mathbf{f}) & 0\\ \sigma P(\mathbf{t}) & 1 - \sigma(P(\mathbf{t}) + P(\mathbf{f})) & \sigma P(\mathbf{f})\\ 0 & \sigma P(\mathbf{t}) & 1 - \sigma P(\mathbf{f}) \end{pmatrix}$$
Confident:
$$U_{C} = \begin{pmatrix} 1 & \rho \epsilon & 0\\ 0 & 1 - \rho & 0\\ 0 & \rho(1 - \epsilon) & 1 \end{pmatrix}$$
Inquisitive:
$$U_{I} = \begin{pmatrix} 1 - \rho + \rho \epsilon & \rho \epsilon & 0\\ \rho(1 - \epsilon) & 1 - \rho & \rho \epsilon\\ 0 & \rho(1 - \epsilon) & 1 - \rho \epsilon \end{pmatrix}$$
Cautious Adventurous
Confident
Inquisitive:
$$\frac{\mathbf{CC} : \mathbf{P}_{t+1}^{T} = U_{C}(F_{C}^{\mathbf{P}t}\mathbf{P}_{t}^{T}) \quad \mathbf{AC} : \mathbf{P}_{t+1}^{T} = U_{C}(F_{A}^{\mathbf{P}t}\mathbf{P}_{t}^{T})}{\mathbf{CI} : \mathbf{P}_{t+1}^{T} = U_{I}(F_{C}^{\mathbf{P}t}\mathbf{P}_{t}^{T}) \quad \mathbf{AI} : \mathbf{P}_{t+1}^{T} = U_{I}(F_{A}^{\mathbf{P}t}\mathbf{P}_{t}^{T})$$

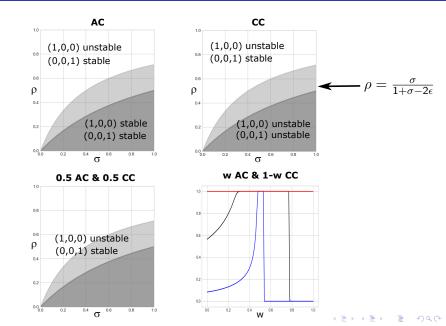
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Adventurous vs Cautious Fusion

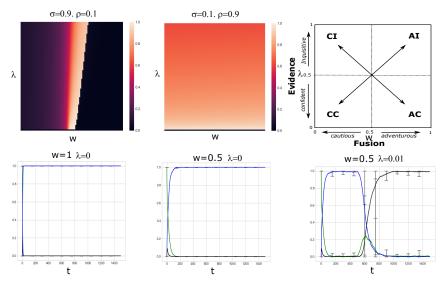
Assume updating is confident: AC vs CC
Initial Proportion: P₀ = (0.9, 0, 0.1)



Fixed Points & Mixtures



Mixtures & Dynamic Environments



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- We have presented social learning as a combination of two processes; fusion and updating.
- We have considered four overall behavioural types generated by independently combining conservative and open-minded approaches to both processes.
- Different behaviour types have been shown to have different convergence and consensus properties.
- Certain heterogeneous mixtures perform best in a range of different learning scenarios.
- A 50/50 mix of adventurous and cautious fusion combined with a mix of 1% inquisitive and 99% confident evidential updating, is highly robust especially in dynamic environments.