Noisy population dynamics lead to efficiently compressed semantic systems





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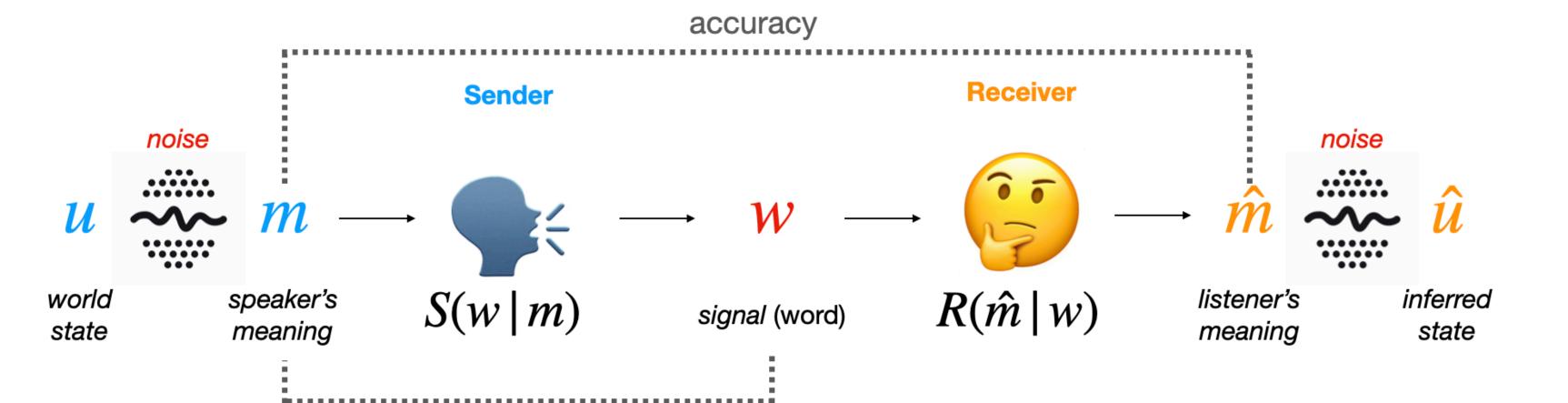
complexity

TÜBINGEN

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How do efficient vocabularies evolve?

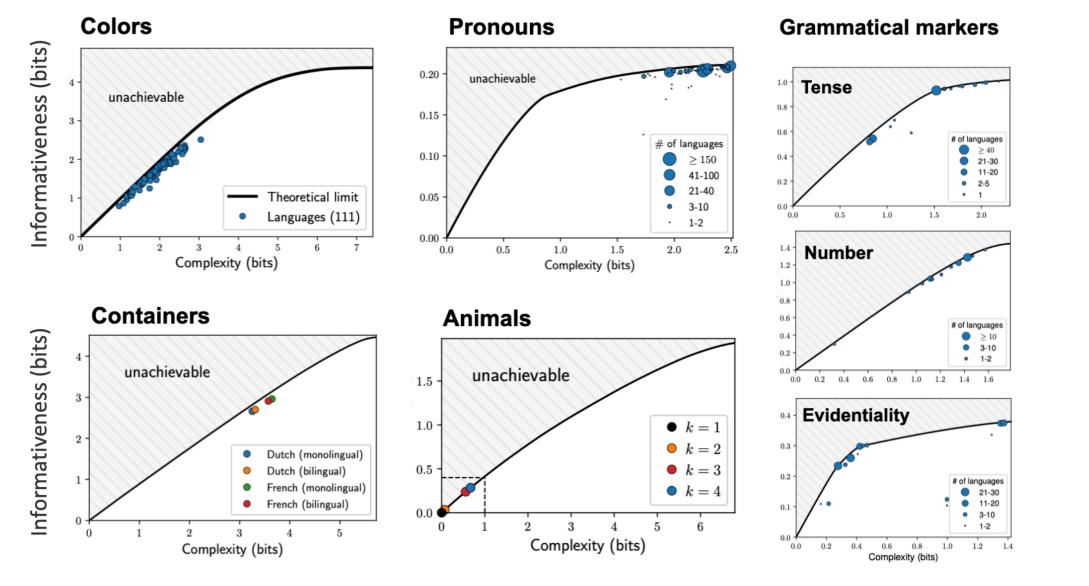
Human vocabularies optimally balance communicative and cognitive cost [2-5]



What agent-based dynamics might drive efficient trade-offs?

The Information Bottleneck

IB principle shown to shape semantic systems [1-5]



characterizes systems' optimal, fixed point behavior **IB** objective

$$\min_{S(w|m)} I(M; W) - \beta I(W; U)$$

The Replicator Dynamics

- foundational model in evolutionary game theory [6]
- a noisy discrete-time replicator dynamic [7,8]:

Sender update

$$S^{(t+1)}(w \mid u) \propto \sum_{m} p(m \mid u) \cdot S^{(t)}(w \mid m) \cdot f_{S}(w, m)$$

Receiver update

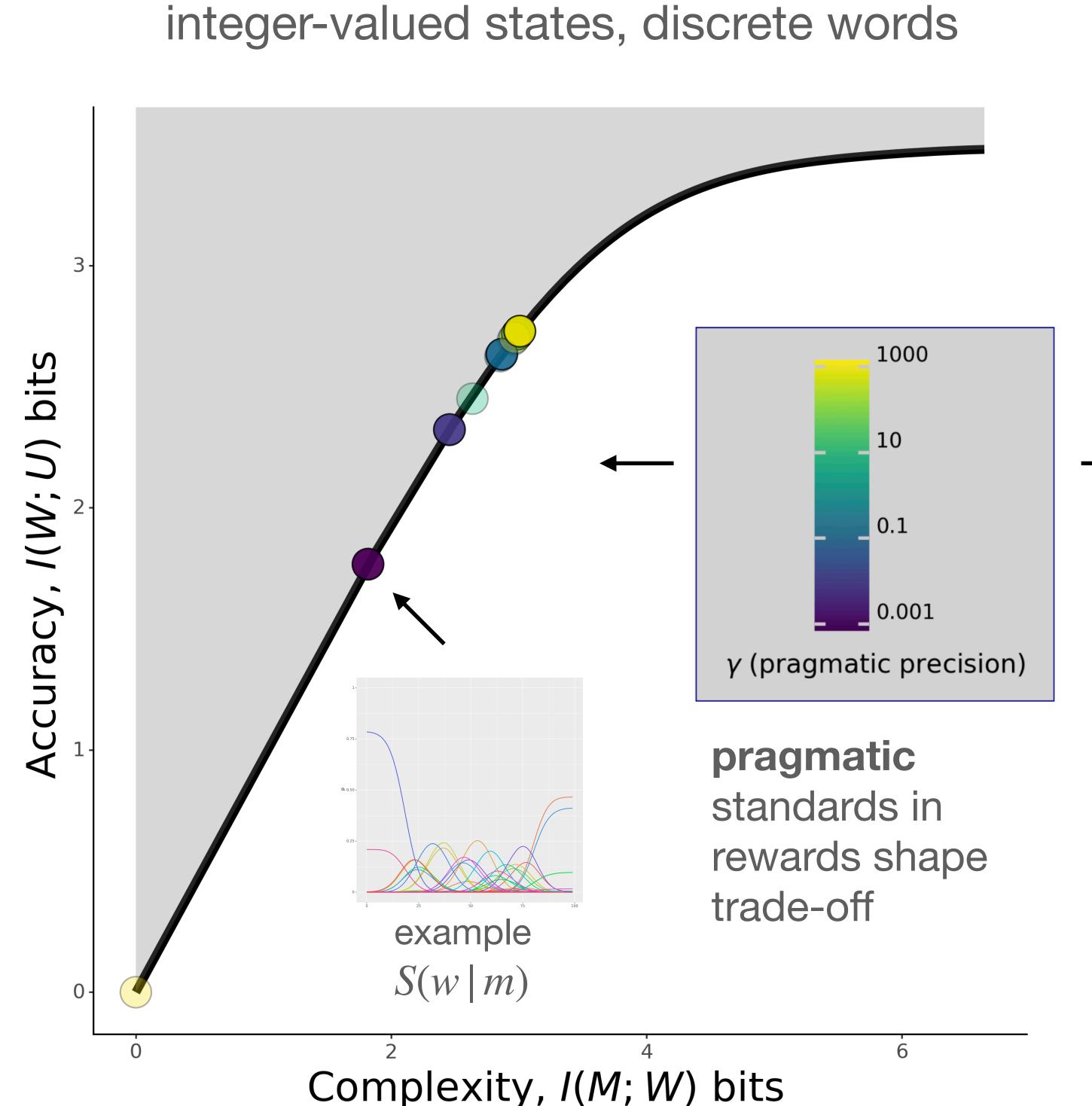
$$R^{(t+1)}(\hat{u} \mid w) \propto \sum_{\hat{m}} p(\hat{m} \mid \hat{u}) \cdot R^{(t)}(\hat{m} \mid w) \cdot f_R(w, \hat{m})$$

Fitness functions

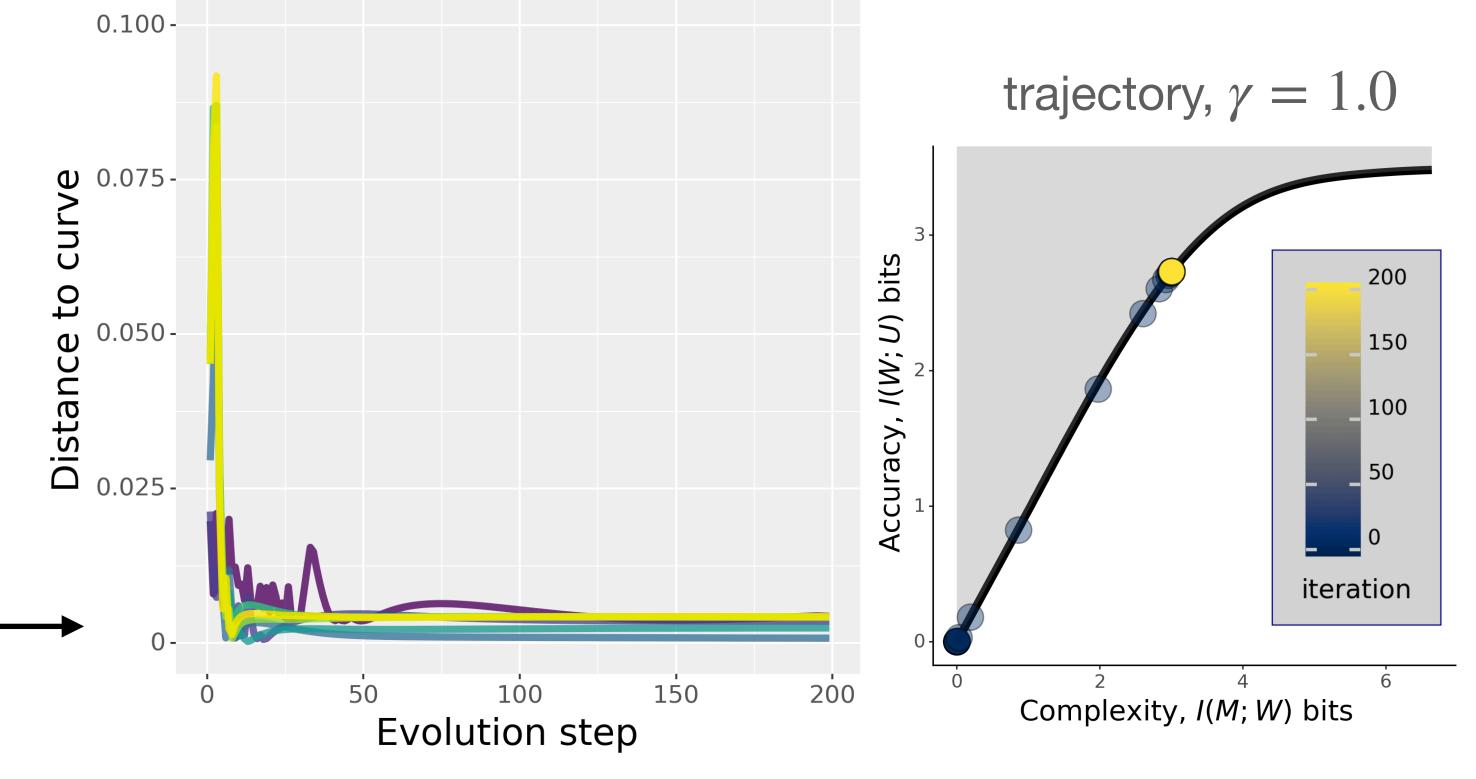
$$f_R(w, \hat{m}) = \sum_{m} P(m) \cdot S^{(t)}(w \mid m) \cdot \operatorname{sim}_{\gamma}(m, \hat{m}) \qquad f_R(w, m) = \sum_{\hat{m}} R^{(t)}(\hat{m} \mid w) \cdot \operatorname{sim}_{\gamma}(m, \hat{m})$$

The replicator dynamics leads to near-optimal compression

Numerical simulations with 100



systems tend to evolve towards bound



Conclusions

efficiency can emerge via replicator dynamics pragmatic pressures shape trade-off solutions future work: test empirically, explore formal links between game dynamics and IB