Learning biases may prevent lexicalization of pragmatic inferences

A case study combining iterated (Bayesian) learning and functional selection

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The semantics-pragmatics distinction

Semantics

Literal meaning (truth-conditional)

Pragmatics

Information beyond literal meaning (e.g. defeasible inferences)

Scalar inferences

- (1) $\langle some, many, most, all \rangle$
 - a. All students came to class
 → Some students came to class
 - b. Some students came to class
 - → Not all students came to class
- (2) $\langle may, should, must \rangle$
- (3) $\langle one, two, three, ... \rangle$
- (4) $\langle or, and \rangle$
- (5) ...

The use of a less informative expression when a more informative one could have been used* can license a defeasible inference that stronger alternatives do not hold

*The hearer assumes the speaker to be knowledgeable and cooperative

- · Laurence R. Horn. On the Semantic Properties of Logical Operators in English. Indiana University Linguistics Club, Bloomington, IN, 1972
- Gerald Gazdar. Pragmatics, Implicature, Presuposition and Logical Form.
 Academic Press, New York, 1979
- · Paul Grice. Logic and conversation.

In Studies in the Ways of Words, chapter 2, pages 22-40. Harvard University Press, Cambridge, MA, 1975

- Why are (pragmatically inferred) upper-bounds of weak(er) alternatives not part of semantics?
- 2. What justifies semantic structure in light of pragmatic enrichment?

Today's talk:

- Propose a model to address (2) and analyze dynamics of linguistic pressures more broadly
- Use (1) as a case study for (2)

Explanations for a lack of semantic upper-bounds I

(1) Signal no commitment to stronger alternatives when knowledge/cooperativity are not given

Conditions that may pressure for English-like semantics:

- Contextual cues are very reliable
- Morphosyntactic disambiguation is not frequently necessary
- Morphosyntactic disambiguation is not very costly
- Cost of larger lexica higher than morphosyntactic disambiguation

Explanations for a lack of semantic upper-bounds II

(2) Lack of upper-bounds provides learnability advantage

Model

Components I: Probabilistic (pragmatic) language users

- Varied lexica
- Varied lexica
 Varied production and comprehension behavior

- · Anton Benz, Gerhard Jäger, Robert Van Rooii, and Robert Van Rooii, editors, Game theory and pragmatics. Springer, 2005
- · Leon Bergen, Roger Levy, and Noah D Goodman, Pragmatic reasoning through semantic inference, Semantics and Pragmatics, 2016
- · M. C. Frank and N. D. Goodman. Predicting pragmatic reasoning in language games. Science, 336(6084):998-998, 2012
- · Michael Franke and Gerhard Jäger, Pragmatic back-and-forth reasoning. Semantics, Pragmatics and the Case of Scalar Implicatures., pages 170-200, 2014

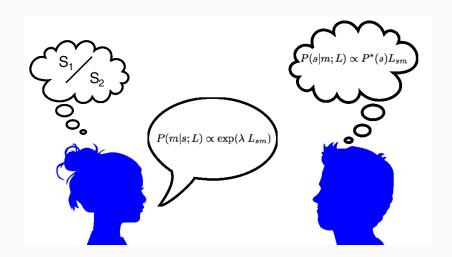
Lexica

- s₁: Bill read some but not all books
- s₂: Bill read all books

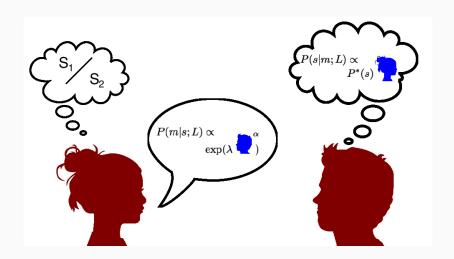
$$L_a = egin{array}{ccc} m_{
m all} & m_{
m some} \ s_1 & 0 & 1 \ s_2 & 1 & 1 \end{array}
ight)$$

$$L_b = egin{array}{ccc} & m_{
m all} & m_{
m some} \ & s_1 \left(egin{array}{ccc} 0 & 1 \ 1 & 0 \end{array}
ight) \end{array}$$

Literal behavior



Pragmatic behavior



Components II: Cultural transmission

Two competing pressures:

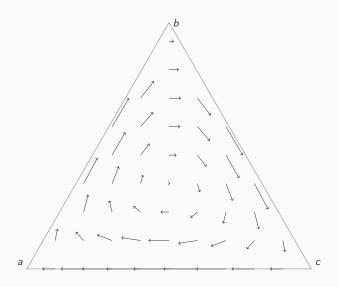
- 1. Communicative efficiency
 - ... as replicator dynamics
- 2. Learnability
 - ... iterated Bayesian learning as mutator dynamics

Replicator-mutator dynamics

- · Thomas L. Griffiths and Michael L. Kalish. Language evolution by iterated learning with bayesian agents. Cognitive Science, 31(3):441–480, 2007
- · Josef Hofbauer and Karl Sigmund. Evolutionary game dynamics. Bulletin of the American Mathematical Society, 40(04):479–520, 2003
- · M. A. Nowak and D. C. Krakauer. The evolution of language.

 Proceedings of the National Academy of Sciences, 96(14):8028-8033, 1999

Functional pressure (replicator dynamics); $\hat{x}_i = \frac{x_i f_i}{\Phi}$



Iterated learning (mutator dynamics)

- $Q_{ji} \propto \sum_d P(d|t_j)P(t_i|d)$
- $d = \langle \langle s_h, m_n \rangle, ..., \langle s_l, m_o \rangle \rangle$ of length k
- $P(d|t_j)$
- $P(t_i|d) \propto [P(t_i)P(d|t_i)]^l$, $l \geq 1$
- Prior encodes learning biases of players prior to data exposure

Replicator-mutator dynamics

$$\hat{x}_i = \sum_j Q_{ji} \frac{x_j f_j}{\Phi}$$

Analysis

Lexica, signaling behavior & types

<u>Lexica</u>

$$L_1 = \begin{pmatrix} 0 & 0 \\ 1 & 1 \end{pmatrix} \quad L_2 = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} \quad L_3 = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}$$

$$L_4 = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \quad L_5 = \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix} \quad L_6 = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}$$

Signaling behavior

Literal or pragmatic

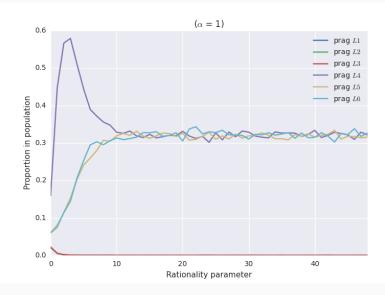
Types

12 types (2 behaviors \times 6 lexica)

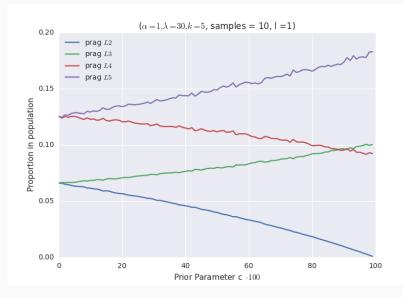


- ullet $P(t_i)$ managed by parameter $c \in [0,1]$
- Signaling behavior: λ , α
- Learning c, k, # of sequences of length k, I

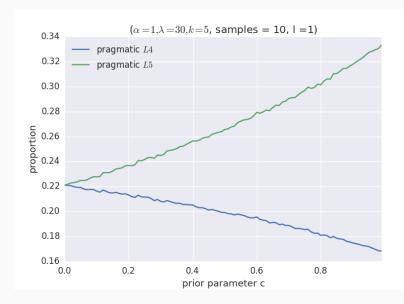
Fitness only



Learning only



Fitness and learning



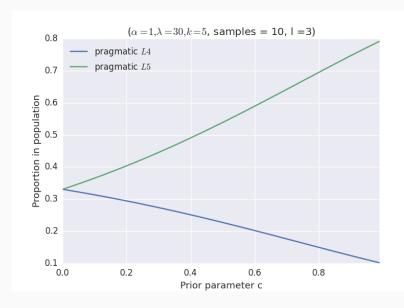
Interim remarks

- Lack of semantic-upper bounds can overcome pressures and stabilize in a population provided...
 - Bias for simpler representations
 - Pragmatics to compensate lack of upper-bounds in use

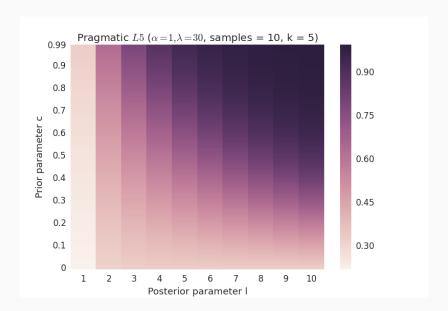
But

- Highly polymorphic populations even for high c
- Role of learning mechanisms, rationality, and learning observations unexplored

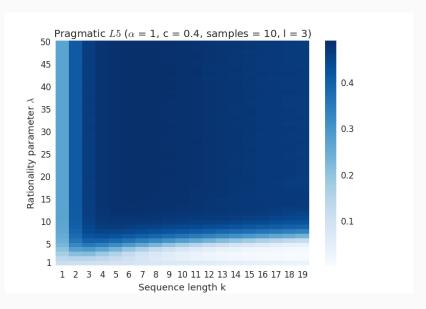
Effect of prior with more posterior maximization



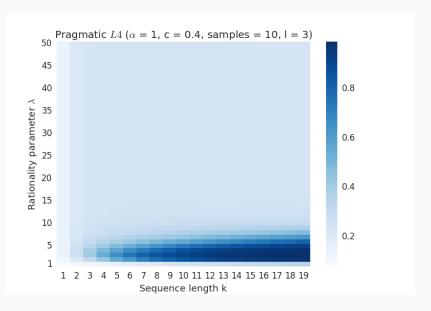
Prior and posterior



Rationality and sequence length I



Rationality and sequence length II



Concluding remarks: Application

- Learnability steers language towards simpler semantics
- Pragmatics compensates for potential loss in expressivity

Lack of semantic upper-bounds

- Functional advantages
 - Reliability of contextual cues to cancel implicatures
 - Lexicon size
 - ...

Provided

- Some degree of rationality in learning & choice
- Relation between bounds and simplicity

Concluding remarks: Model

- Combines
 - Functional pressure
 - · Learning pressure
 - (Probabilistic) hearer & speaker models
 - Distinct languages

Future directions

- Generalization
- Frequency effects
- Larger lexica & uncertainty
- Further applications

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