

# Game Theoretic Pragmatics

L&G Project 2017

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ILLC, Amsterdam, 2017.01.17

- On-line language use as a game between interlocutors
- *Computational* model of back-and-forth reasoning
- Bridge between theoretical descriptive analysis and quantitative experimental data
- Predictions about different types of interlocutor behavior
  - Level of sophistication (bounded rationality)
  - Preferences
  - Choice rule
  - Priors

# Gricean Maxims (quick recap)

## 1. Maxim of Quality

- Do not say what you believe to be false
- Do not say that for which you lack adequate evidence

## 2. Maxim of Quantity

- Make your contribution as informative as is required
- Do not make your contribution more informative than is required

## 3. Maxim of Relevance

## 4. Maxim of Manner

(avoid ambiguity & obscurity, be brief & orderly)

• Paul Grice. *Logic and conversation*.

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

## Grice

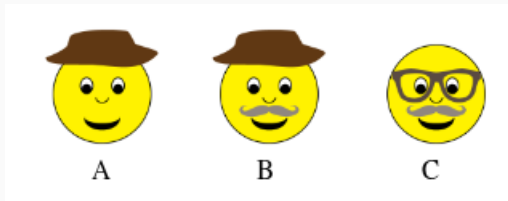
Conjecture about rational foundation of maxims  
(+ cooperativity assumption)

## Game Theoretic Pragmatics

Can use be accounted for as (approximate) rational behavior?

- Not restricted to cooperation
- No mention or refinement of maxims

## The guy with the mustache...



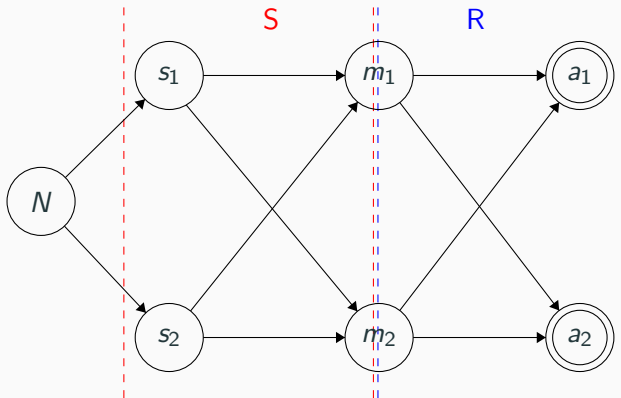
- (1) Some students passed the exam.
- a.  $\rightarrow$  More than one student passed the exam.
  - b.  $\rightsquigarrow$  Some but not all students passed the exam.

## Away from static solution concepts...

- Model agent-internal perspective
- What behavioral predictions follow from which assumptions about agents'...
  - (higher order) beliefs
  - rationality
  - dispositions to act or revise beliefs
  - ...

- Games as conversational *context* models
- Iterated X-Response
  - Iterated Best Response
  - Iterated Quantal Response
- Recent developments & quantitative modelling

## Reminder: Lewisian signaling



· David Lewis. *Convention: A Philosophical Study*.  
Harvard University Press, Cambridge, 1969



# Setup I

- Two players: Sender & Receiver;
- States  $S$
- Messages  $M$
- Acts  $A$

In the following:  $S = A$

Interpretation game

## Setup II

- Beliefs  $\mathbf{p} = \Delta^+(|S|)$ ,  
 $\Delta^+(n) = \{\langle p_0, \dots, p_{n-1} \rangle \mid \sum p_i = 1 \text{ and } p_i > 0\}$
- Conventional meaning as Boolean matrix  $L$ ,  $(|S|, |M|)$ ,  
where  $L_{ij}$  is the value of  $m_j$  in  $s_i$ ;
- $U_R$  and  $U_S$  are  $(|S|, |A|)$ -matrices;
- cost vector  $\mathbf{c}$  of length  $|M|$

In the following:  $U_R = U_S = U$

Cooperativity

- Semantic conventions as a starting point  
Usually: a shared language fragment
- Game represents relevant conversational context

(2) Some students passed the exam.

$\rightsquigarrow$  Some but not all students passed the exam

- *S?*
- *M?*
- **p?**
- **c?**
- ...

# Scalar implicatures

(3) Some students passed the exam.

$\rightsquigarrow$  Some but not all students passed the exam

- $S = \{s_{\exists \rightarrow \forall}, s_{\forall}\}$
- $M = \{m_{\text{some}}, m_{\text{all}}\}$
- $\mathbf{p} = \left\langle \frac{1}{2}, \frac{1}{2} \right\rangle$

$$L = \begin{array}{cc} & \begin{array}{cc} m_{\text{some}} & m_{\text{all}} \end{array} \\ \begin{array}{c} s_{\exists \rightarrow \forall} \\ s_{\forall} \end{array} & \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix} \end{array}$$

# Manner implicatures

(4) Mercader killed Trotsky.

↪ Mercader killed Trotsky in a stereotypical way

(5) Mercader caused Trotsky to die.

↪ Mercader killed Trotsky in a non-stereotypical way

# Manner implicatures

(6) Mercader killed Trotsky.

$\rightsquigarrow$  Mercader killed Trotsky in a stereotypical way

(7) Mercader caused Trotsky to die.

$\rightsquigarrow$  Mercader killed Trotsky in a non-stereotypical way

- $S = \{s, s^*\} = A$

- $M = \{m, m^*\}$

- $\mathbf{p} = \left\langle \frac{1}{2} + \epsilon, \frac{1}{2} - \epsilon \right\rangle$

- $\mathbf{c} = \langle 0, \delta \rangle, \delta > 0$

$$L = \begin{array}{cc} & \begin{matrix} m & m^* \end{matrix} \\ \begin{matrix} s \\ s^* \end{matrix} & \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \end{array}$$

# Lifted games for epistemic states

- (8) Alice ate some of the cookies.
- Speaker believes that Alice didn't eat all of the cookies.
  - Speaker is uncertain whether Alice ate all of the cookies.

- $S = \{s_{[\exists \neg \forall]}, s_{[\forall]}, s_{[\exists \neg \forall, \forall]}\}$
- $\mathbf{p} = \langle p_1, p_2, p_3 \rangle$
- Under competence assumption:  $p_1, p_2 > p_3$
- Uncertain competence:  $p_1 = p_2 = p_3$

$$L_E = \begin{array}{cc} & \begin{array}{cc} m_{\text{some}} & m_{\text{all}} \end{array} \\ \begin{array}{c} s_{\exists \neg \forall} \\ s_{\forall} \\ s_{\exists \neg \forall, \forall} \end{array} & \begin{pmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 0 \end{pmatrix} \end{array}$$



# Iterated x-Response

**IxR-models:** Family of pragmatic reasoning models,  
 $x$  is *best* (IBR), *cautious* (ICR), or *quantal* (IQR)

- Sophistication as strategic player type
- Strategic types are hierarchically organized: level-0, ..., level- $n$
- Level-0 players are naïve/literal/honest
- level- $k + 1$  agents have a behavioral belief about his interlocutor being of level- $l$ ,  $l \leq k$

· Michael Franke. *Signal to Act: Game Theoretic Pragmatics*.

PhD thesis, University of Amsterdam, 2009

· Michael Franke and Gerhard Jäger. *Pragmatic back-and-forth reasoning*.

*Semantics, Pragmatics and the Case of Scalar Implicatures*. Ed. by Salvatore Pistoia Reda. *Palgrave Studies in Pragmatics Language and Cognition*. New York: Palgrave MacMillan, pages 170–200, 2014

## Strategies

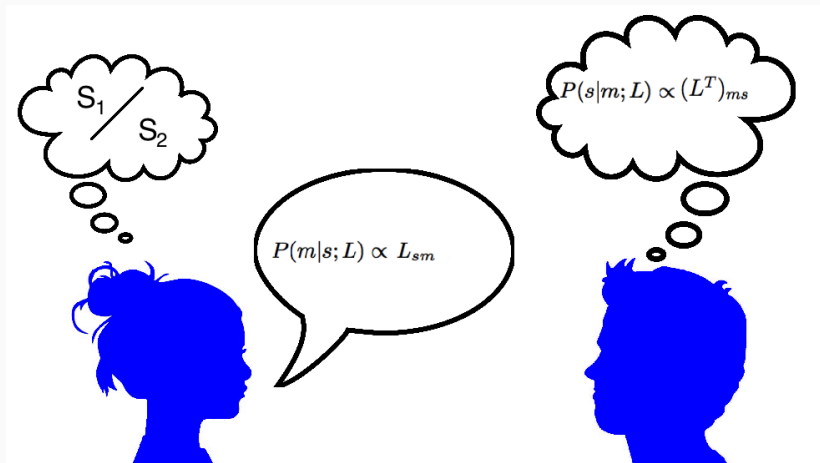
A sender strategy  $\sigma$  is a row-stochastic  $(|S|, |M|)$ -matrix

$$\sigma = \begin{array}{cc} & \begin{array}{cc} m_{\text{some}} & m_{\text{all}} \end{array} \\ \begin{array}{c} t_{\exists \rightarrow \forall} \\ t_{\forall} \end{array} & \left( \begin{array}{cc} .7 & .3 \\ .4 & .6 \end{array} \right) \end{array}$$

A receiver strategy  $\rho$  is a row-stochastic  $(|M|, |A|)$ -matrix

$$\rho = \begin{array}{cc} & \begin{array}{cc} t_{\exists \rightarrow \forall} & t_{\forall} \end{array} \\ \begin{array}{c} m_{\text{some}} \\ m_{\text{all}} \end{array} & \left( \begin{array}{cc} .1 & .9 \\ .5 & .5 \end{array} \right) \end{array}$$

## Strategic types I: Literal behavior



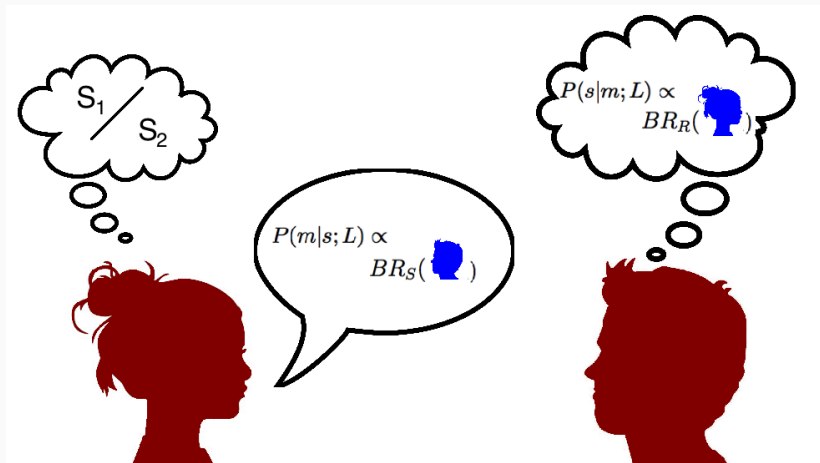
## Strategic types I: Literal behavior

$$\sigma_0 = \{\text{Norm}(L)\}$$

$$\rho_0 = \{\text{Norm}(L^T)\},$$

where, for  $(m,n)$ -matrix  $A$ ,  $\text{Norm}(A)$  maps  $A$  to another matrix such that  $\text{Norm}(A)_i \propto A_i$  if  $\sum_j (A_{ij}) > 0$  and  $\text{Norm}(A)_{ij} = \frac{1}{n}$  otherwise.

## Strategic types II: level-1 behavior



## Strategic types II: Sophisticated/Gricean behavior

Let  $X$  be an ordered set of strategies and  $|X| = d$ .

Unconstrained belief

$$\Pi(X) = \left\{ \sum_{x_i \in X} p_i x_i \mid \langle p_1, \dots, p_n \rangle \in \Delta(d) \right\}$$

Cautious belief

$$\Pi^c(X) = \left\{ \sum_{x_i \in X} p_i x_i \mid \langle p_1, \dots, p_n \rangle \in \Delta^+(d) \right\}$$

Unbiased belief

$$\Pi^u(X) = \left\{ \sum_{x_i \in X} \frac{1}{d} x_i \right\}$$

## Best response

$$BR_S(\rho) = \{s \in S \mid s_{ij} = 1 \Rightarrow j \in \arg_k \max(\rho^T - c)_{ik}\}$$

$$S_{k+1} = BR(\Pi^u(R_k)) \quad (\text{IBR}) \quad \text{or} \quad S_{k+1} = BR(\Pi^c(R_k)) \quad (\text{ICR})$$

$$\begin{aligned} BR_R(\sigma) = \{r \in R \mid (r_{ij} = 1 \wedge \max \sigma_j^T > 0 \Rightarrow \\ j \in \arg_k \max(\sigma^T \times \mathbf{p})_{ik}) \wedge \\ (r_{ij} = 1 \wedge \max \sigma_j^T = 1 \Rightarrow L_{ij} = 1)\} \end{aligned}$$

## Example I: Scalar implicatures

- $S = \{s_{\exists \neg \forall}, s_{\forall}\}$
- $M = \{m_{\text{some}}, m_{\text{all}}\}$
- $S = A$
- $U_S = U_R$
- $\mathbf{p} = \left\langle \frac{1}{2}, \frac{1}{2} \right\rangle$

$$L = \begin{array}{cc} & \begin{array}{cc} m_{\text{some}} & m_{\text{all}} \end{array} \\ \begin{array}{c} s_{\exists \neg \forall} \\ s_{\forall} \end{array} & \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix} \end{array}$$



## Example I: Scalar implicatures

$$L = \begin{array}{cc} & \begin{array}{cc} m_{\text{some}} & m_{\text{all}} \end{array} \\ \begin{array}{c} t_{\exists \rightarrow \forall} \\ t_{\forall} \end{array} & \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix} \end{array}$$

$$\sigma_0 = \left\{ \begin{pmatrix} 1 & 0 \\ .5 & .5 \end{pmatrix} \right\}$$

$$\rho_0 = \left\{ \begin{pmatrix} .5 & .5 \\ 0 & 1 \end{pmatrix} \right\}$$

$$\Pi^u(\sigma_0) = \left\{ \begin{pmatrix} 1 & 0 \\ .5 & .5 \end{pmatrix} \right\}$$

$$\Pi^u(\rho_0) = \left\{ \begin{pmatrix} .5 & .5 \\ 0 & 1 \end{pmatrix} \right\}$$

$$\sigma_1 = \left\{ \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right\}$$

$$\rho_1 = \left\{ \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right\}$$

## Example II: Horn games

(9) Mercader killed Trotsky.

$\rightsquigarrow$  Mercader killed Trotsky in a stereotypical way

(10) Mercader caused Trotsky to die.

$\rightsquigarrow$  Mercader killed Trotsky in a non-stereotypical way

- $T = \{t, t^*\} = A$

- $M = \{m, m^*\}$

- $\mathbf{p} = \left\langle \frac{1}{2} + \epsilon, \frac{1}{2} - \epsilon \right\rangle$

- $\mathbf{c} = \langle 0, \delta \rangle, \delta > 0$

$$B = \begin{array}{cc} & m & m^* \\ \begin{array}{c} t \\ t^* \end{array} & \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \end{array}$$

## Horn game up to divergence

$$\sigma_0 = \left\{ \begin{pmatrix} .5 & .5 \\ .5 & .5 \end{pmatrix} \right\}$$

$$\rho_0 = \left\{ \begin{pmatrix} .5 & .5 \\ .5 & .5 \end{pmatrix} \right\}$$

$$\sigma_1 = \left\{ \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix} \right\}$$

$$\rho_1 = \left\{ \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix} \right\}$$

Under  $\sigma_1$   $m^*$  is a surprise message. Hence:

$$\sigma_2 = \left\{ \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix} \right\}$$

$$\rho_2 = \left\{ \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right\}$$

## Unbiased Horn Game (IBR)

$$\Pi^u(\sigma_2) = \left\{ \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix} \right\}$$

$$\Pi^u(\rho_2) = \left\{ \begin{pmatrix} 1 & 0 \\ .5 & .5 \end{pmatrix} \right\}$$

$$\sigma_3 = \left\{ \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right\}$$

$$\rho_3 = \left\{ \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right\}$$

$$\sigma_4 = \left\{ \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right\}$$

$$\rho_4 = \left\{ \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right\}$$

# Iterated Quantal Response

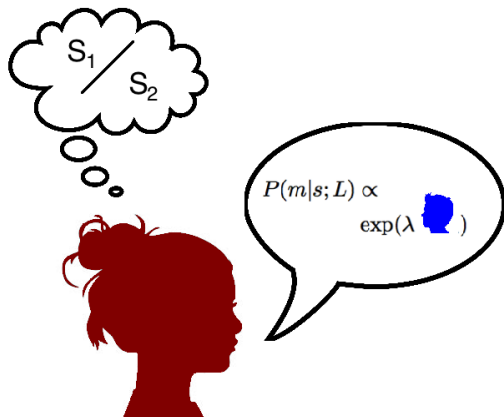
- IBR/ICR maximizes expected utility; response as function from expected utilities to choice probabilities

For arbitrary expected utility matrix  $U$ ;

$$BR(U) = \text{Norm}(\text{maxrow}(U))$$

- However, mistakes may happen in calculating  $U$  or in implementing  $BR(\cdot)$
- IQC as probabilistic member of the IxR family with no assumption of fully rational best responses (cf. RSA models)





# Iterated Quantal Response

- $QR_\lambda(U)_{ij} \propto \exp(\lambda U_{ij})$ , where  $\lambda$  is a rationality parameter
- Higher rationality with higher  $\lambda$ -values  
Random for  $\lambda = 0$  and  $\lim_{\lambda \rightarrow \infty} QR_\lambda(U) = BR(U)$

· Duncan R. Luce. *Individual choice behavior: a theoretical analysis*.  
Wiley, 1959

· Richard S. Sutton and Andrew G. Barto. *Introduction to Reinforcement Learning*.  
MIT Press, Cambridge, MA, USA, 1998



## Some-all IQR

Assuming a single  $\lambda$  for all players and that each player believes his opponent to be quantal,  $\lambda \in \{0, 0.5, 1, 5\}$  yields

$$QR_0(L) = \left\{ \begin{pmatrix} .5 & .5 \\ .5 & .5 \end{pmatrix} \right\} \qquad QR_{.5}(L) = \left\{ \begin{pmatrix} .622 & .378 \\ .5 & .5 \end{pmatrix} \right\}$$

$$QR_1(L) = \left\{ \begin{pmatrix} .731 & .269 \\ .5 & .5 \end{pmatrix} \right\} \qquad QR_5(L) = \left\{ \begin{pmatrix} .993 & .007 \\ .5 & .5 \end{pmatrix} \right\}$$

## Some IQR-sequences

Keeping track of the diagonal, we get

$$\lambda = .5$$

$$\sigma_0 \approx \langle .622, .5 \rangle$$

$$\rho_1 \approx \langle .514, .517 \rangle$$

$$\sigma_2 \approx \langle .503, .504 \rangle$$

$$\rho_3 \approx \langle .5, .5 \rangle$$

$$\lambda = 5$$

$$\sigma_0 \approx \langle .993, .5 \rangle$$

$$\rho_1 \approx \langle .839, .992 \rangle$$

$$\sigma_2 \approx \langle .984, .984 \rangle$$

$$\rho_3 \approx \langle .992, .992 \rangle$$

That is, IQR gives a parametrized prediction of the probability of scalar inferences, inter alia

## Three experiments using MT

- Experiment I: Comprehension
- Experiment II: Production
- Experiment III: Comprehension (with cost)

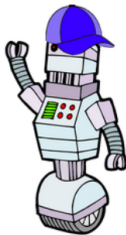
## Trials

- Simple (single step) implicatures
- Complex (two step) implicatures
- Ambiguous fillers
- Unambiguous fillers

Judith Degen, Michael Franke, and Gerhard Jäger. *Cost-based pragmatic inference about referential expressions*. In *Proceedings of the 35th Annual Conference of the Cognitive Science Society*, pages 376–381, 2013

# Comprehension trial: simple implicature

The previous participant said:

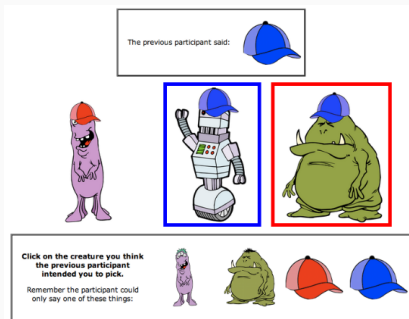


**Click on the creature you think  
the previous participant  
intended you to pick.**

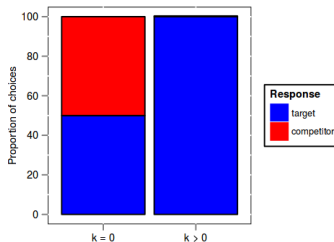
Remember the participant could  
only say one of these things:



# Comprehension prediction: simple implicature



- IBR predictions for distribution of responses over target and competitor:



# Comprehension trial: complex implicature

The previous participant said:

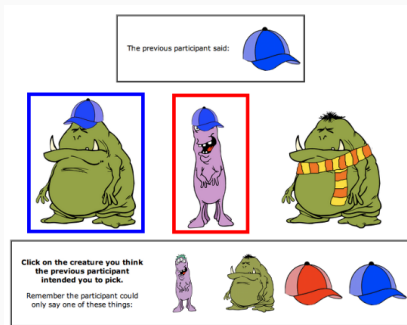


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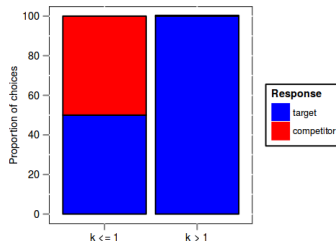
Remember the participant could  
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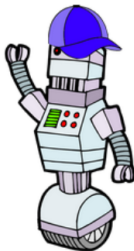
# Comprehension prediction: complex implicature



- IBR predictions for distribution of responses over target and competitor:



# Production trial: simple implicature

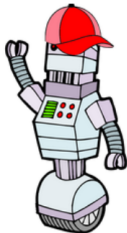


Your task is to get another worker to pick out the highlighted creature. It's not highlighted on their display.  
**Click on one of the following four messages to send it to the other worker and get them to pick out the right creature.** The other worker knows you can only send these messages.





# Production trial: complex implicature

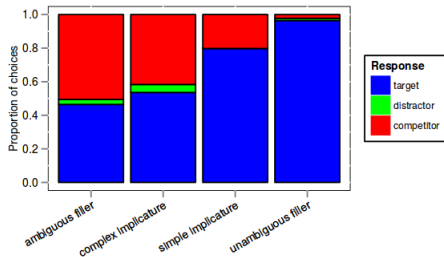


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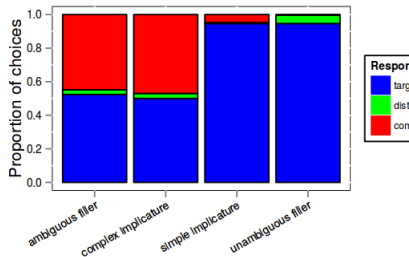


# Experiments I and II: Results

Experiment 1  
(comprehension)

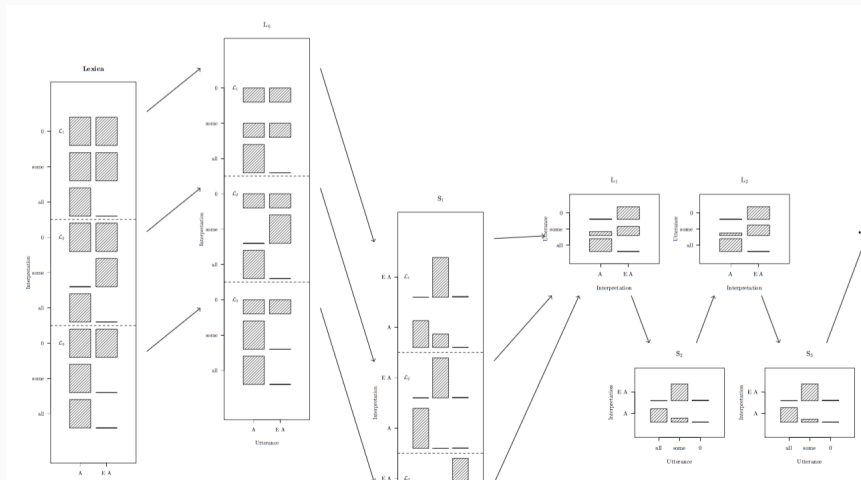


Experiment 2  
(production)



- IBR does not provide a good fit  
(particularly for complex implicatures)
- Asymmetry between production and comprehension
- IQR provides a means to fit the data

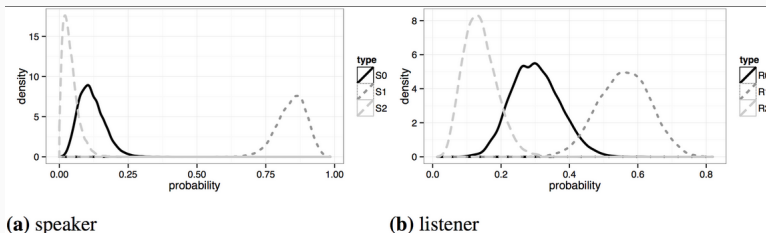
# What's new in probabilistic rational models of language use



· Leon Bergen, Roger Levy, and Noah D Goodman. **Pragmatic reasoning through semantic inference.**

*Semantics and Pragmatics*, 2016

# What's new in probabilistic rational models of language use



· Michael Franke and Judith Degen. Reasoning in reference games: Individual-vs. population-level probabilistic modeling.  
*PloS one*, 11(5), 2016

# What's new in probabilistic rational models of language use

- Lexical uncertainty
- Compositional signaling games
- Homogeneous populations
- Lack of common priors
- Machine learning – IQR/RSA as a trained classifier
- Applications to other implicatures, irony, metaphors, politeness, ...

## Some open issues

- Modeler degrees of freedom for RSA/IQR-style models
- Level of description?
- Choice rules?
- Which implicatures should be derived online? What about fossilization?
- What is the right representation of the game being played? (Alternatives matter - a lot)

## Final remarks

- IxR-models target reasoning under goal-oriented rational agency
- Inferences not derived from maxims but from behavioral principles
- Potentially interesting for predictions and fit of experiments
- Allow for variation in terms of rationality, preferences, cooperativity, ...
- Straightforward to plug into other dynamics  
(see tomorrow's session)





Leon Bergen, Roger Levy, and Noah D Goodman.

**Pragmatic reasoning through semantic inference.**

*Semantics and Pragmatics*, 2016.



Judith Degen, Michael Franke, and Gerhard Jäger.

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