

# Computational pragmatics: Introducing the Rational Speech Act framework

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Can you reach the top shelf?



... please?



... or should you be working the register?

Same string, many possible interpretations depending on *context*.

When a diplomat says *yes*, he means 'perhaps';

When he says *perhaps*, he means 'no';

When he says *no*, he is not a diplomat.

— Voltaire

Context: common-ground information, social dynamics, beliefs about speaker intentions/identity, 'norms' of linguistic interaction...

Pragmatics: study of speaker meaning and its relationship to context.

“My friend has glasses”



If I had meant...



... there were 'better' things I could have said.

So, I probably meant



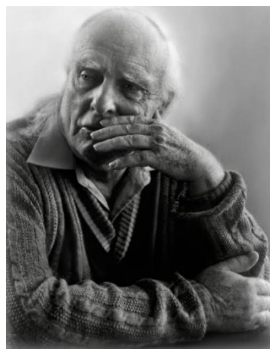
(But this is obviously quite context-dependent!)

# Grice's approach

Pragmatic inference results from counterfactual reasoning about *alternative utterances* a speaker could have produced.

Reasoning assumes that speakers are rational and cooperative.

Assumptions expressed as *maxims* of conversation.

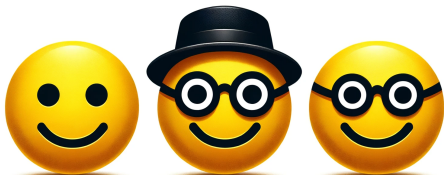


Maxim of *quality*: don't say things that you know to be false (or for which you lack evidence).

Maxim of *quantity*: say all you need to say (and no more).

Grice (1975)

“My friend has glasses”



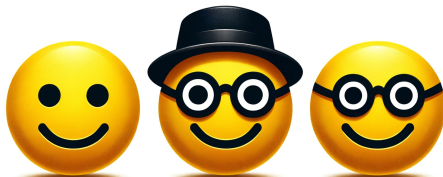
- (a) *Contextual premise*: ↑ these are S's three relevant possible referents.
- (b) *Contextual premise*: it is mutual, public information that S has complete knowledge re: who they intend to refer to.
- (c) Assume S is being cooperative insofar as they are obeying the maxims of Quality and Quantity.
- (d) Then S will assert what is maximally relevant, informative, and true.
- (e) By (a), *My friend has a hat* (hat) is more informative (and just as relevant) as *My friend has glasses*.
- (f) Therefore, S must lack sufficient evidence to assert hat.
- (g) By (b), S must lack evidence for hat because it's false.

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Inspired by Chris Potts' handout on conversational implicature:

<https://web.stanford.edu/class/linguist130a/materials/ling130a-handout-02-20-implicature.pdf>

# The Classic View of Pragmatics



“My friend has glasses” implicates  $\neg$ hat.

Meaning treated as categorical: implicatures computed or not.

Inferences arise categorically if premises are met.

Gradience (and population-level variation!) ignored, or accommodated by exceptions or processing constraints.

Degen (2023)



# The 'probabilistic' turn in pragmatics

Probabilistic pragmatics: computational approach integrating insights from formal semantics, psycholinguistics, cognitive science

Key advances:

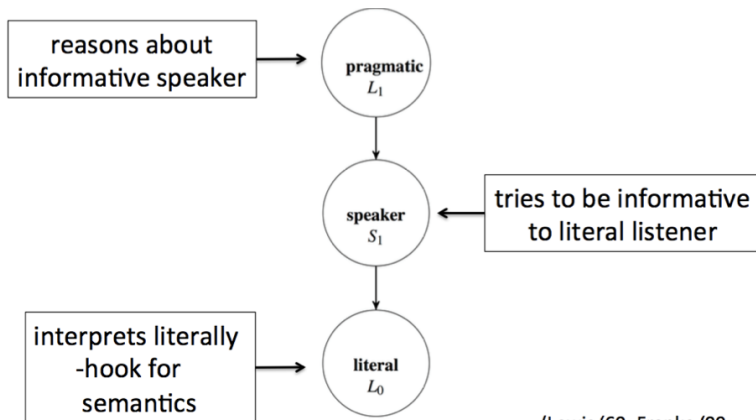
- Formalizes general principles of conversation (including Grice's 'maxims')

- Treats language production and interpretation as probabilistic, boundedly rational processes

- Allows linguistic knowledge to interact with communicative pressures and world knowledge

Rational Speech Act (RSA) framework: influential probabilistic approach

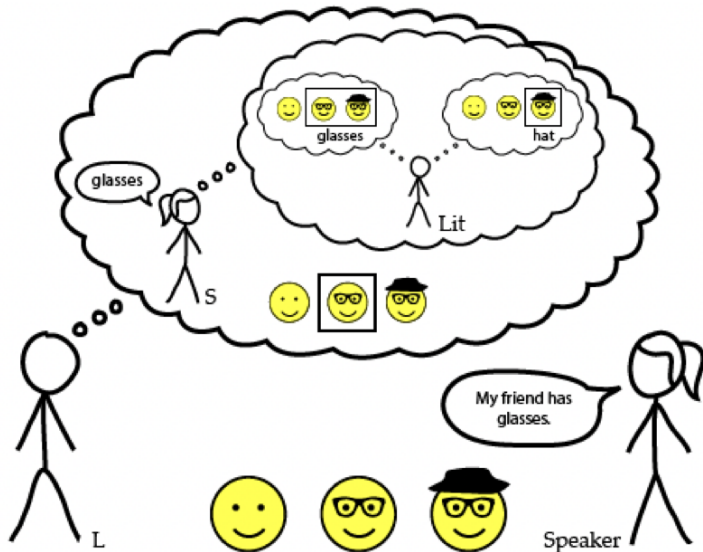
# Modeling pragmatic inference: the RSA framework



(Lewis '69; Franke '09;  
Frank & Goodman '12)

Goodman and Frank (2016)

# Modeling pragmatic inference: the RSA framework



Goodman and Frank (2016)

# The basic 'vanilla' RSA model

Language use as signaling game between speaker and listener

Defined over utterances  $U$  and meanings  $S$

Semantic foundation: literal meaning  $[[\cdot]] : U \rightarrow S \rightarrow \{0, 1\}$

Recursive probabilistic production and interpretation rules:

Literal listener  $Pr_{lit}$ : interprets according to literal semantics

Pragmatic speaker  $Pr_{speaker}$ : reasons about  $Pr_{lit}$ , balances informativeness and cost

Pragmatic listener  $Pr_L$ : infers  $Pr_{speaker}$ 's meaning w/ Bayes' rule

**Bayes' rule:** for a hypothesis  $H$  and data  $d$ :

$$Pr(H|d) \propto Pr(d|H) * Pr(H)$$

For  $Pr_L$ , Hypotheses are elements of  $S$  (intended meanings), data are elements of  $U$  (observed utterances).

# Pragmatic interpretation in 'vanilla' RSA

$$S = \left\{ \text{😊} \text{ 😁} \text{ 😎} \right\} \quad U = \{ \text{"hat"}, \text{"glasses"}, \text{"smiling"} \}$$

'Literal' listener:  $Pr_{lit}(s|u) = \frac{[u]^s}{\sum_{s' \in S} [u]^{s'}}$

'Literal' listener	😊	😁	😎
"hat"	0	0	1
"glasses"	0	1	1
"smiling"	1	1	1

# Pragmatic interpretation in 'vanilla' RSA

$$S = \left\{ \begin{array}{c} \text{😊} \\ \text{👓😊} \\ \text{👓😊👒} \end{array} \right\} \quad U = \{ \text{"hat"}, \text{"glasses"}, \text{"smiling"} \}$$

'Literal' listener:  $Pr_{lit}(s|u) = \frac{[u]^s}{\sum_{s' \in S} [u]^{s'}}$

'Literal' listener	😊	👓😊	👓😊👒
"hat"	0	0	1
"glasses"	0	0.5	0.5
"smiling"	0.33	0.33	0.33

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$$S = \left\{ \text{😊} \text{ 😁} \text{ 😎} \right\} \quad U = \{ \text{"hat"}, \text{"glasses"}, \text{"smiling"} \}$$




'Literal' listener:  $Pr_{lit}(s|u) = \frac{[u]^s}{\sum_{s' \in S} [u]^{s'}}$

'Literal' listener	😊	😁	😎
"hat"		0	
"glasses"		0.5	
"smiling"		0.33	

# Pragmatic interpretation in 'vanilla' RSA

$$S = \left\{ \begin{array}{c} \text{😊} \\ \text{👓😊} \\ \text{👓😊👒} \end{array} \right\} \quad U = \{ \text{"hat"}, \text{"glasses"}, \text{"smiling"} \}$$

'Pragmatic' speaker:  $Pr_{speaker}(u|s) = \frac{Pr_{lit}(s|u)}{\sum_{u' \in U} Pr_{lit}(s|u')}$




'Pragmatic' speaker	"hat"	"glasses"	"smiling"
	0	0	1
	0	0.5	0.33
	1	0.5	0.33



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$$s = \left\{ \begin{array}{c} \text{😊} \\ \text{👓😊} \\ \text{👓😊👒} \end{array} \right\} \quad U = \{ \text{"hat"}, \text{"glasses"}, \text{"smiling"} \}$$

'Pragmatic' speaker:  $Pr_{speaker}(u|s) = \frac{Pr_{lit}(s|u)}{\sum_{u' \in U} Pr_{lit}(s|u')}$

'Pragmatic' speaker	"hat"	"glasses"	"smiling"
	0	0	1
	0	0.60	0.40
	0.55	0.27	0.18

# Pragmatic interpretation in 'vanilla' RSA

$$S = \left\{ \begin{array}{c} \text{😊} \\ \text{👓😊} \\ \text{👓😊👒} \end{array} \right\} \quad U = \{ \text{"hat"}, \text{"glasses"}, \text{"smiling"} \}$$

'Pragmatic' listener:  $Pr_L(s|u) = \frac{Pr_{speaker}(u|s)}{\sum_{s' \in S} Pr_{speaker}(u|s')}$

'Pragmatic' listener	😊	👓😊	👓😊👒
"hat"	0	0	0.55
"glasses"	0	0.60	0.27
"smiling"	1	0.40	0.18

# Pragmatic interpretation in 'vanilla' RSA

$$S = \left\{ \begin{array}{c} \text{😊} \\ \text{👓} \\ \text{👓🎩} \end{array} \right\} \quad U = \{ \text{"hat"}, \text{"glasses"}, \text{"smiling"} \}$$

'Pragmatic' listener:  $Pr_L(s|u) = \frac{Pr_{speaker}(u|s)}{\sum_{s' \in S} Pr_{speaker}(u|s')}$

'Pragmatic' listener	😊	👓	👓🎩
"hat"	0	0	1
"glasses"	0	0.69	0.31
"smiling"	0.64	0.25	0.11

# Enriching the model: additional parameters

- Listener prior over meanings: 'what do I think my interlocutor is trying to say, before observing an utterance'?

$$\text{'Literal' listener: } Pr_{lit}(s|u) = \frac{[[u]]^s * P_{prior}(s)}{\sum_{s' \in S} [[u]]^{s'} * P_{prior}(s')}$$

- 'Cost' term on utterances:
  - o Grice's maxim of manner: be brief (avoid unnecessary prolixity).
  - o Speaker utility formalized as a tradeoff of informativity and cost.
- Rationality term  $\alpha$  'penalizes' low-utility utterances.

$$\text{'Pragmatic' speaker: } Pr_{speaker}(u|s) \propto e^{\alpha[\ln(Pr_{lit}(s|u)) - cost(u)]}$$

- <http://webppl.org/>: online interpreter for WebPPL, a probabilistic programming language (Goodman & Stuhlmüller, 2014)  
Javascript syntax; functional programming properties (e.g., no for-loops)
- <http://forestdb.org/>: a repository with several WebPPL implementations of published RSA models
- <https://www.problang.org/> (Scontras, Tessler, & Franke, online): an intro to RSA taught in WebPPL.

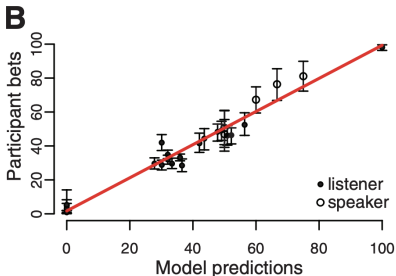
Implementation of the 'smiley' model:

- Model code: <https://tinyurl.com/rsa-model>
- Copy and paste into WebPPL interpreter <http://webppl.org/>

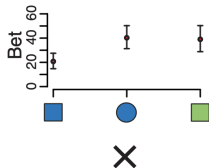
**A** *Speaker:* Imagine you are talking to someone and you want to refer to the middle object. Which word would you use, “blue” or “circle”?



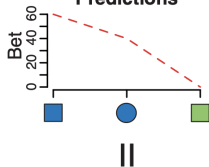
*Listener/Salience:* Imagine someone is talking to you and uses [the word “blue”/a word you don’t know] to refer to one of these objects. Which object are they talking about?



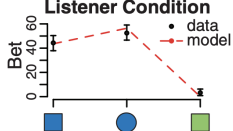
**C** *Prior: Salience Condition*



**Likelihood: Model Predictions**




**Posterior: Model vs. Listener Condition**



Frank and Goodman (2012)

# Enriching the model: multiple speaker 'goals'

In 'vanilla' model, speakers have one communicative goal (e.g.,  
listener identifies intended referent from  )

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In 'vanilla' model, speakers have one communicative goal (e.g., listener identifies intended referent from 😊 😎 🧐 )

In practice, goals change across contexts, and utterances are themselves indicative of speaker goals.

S: "The kettle cost me a million dollars"

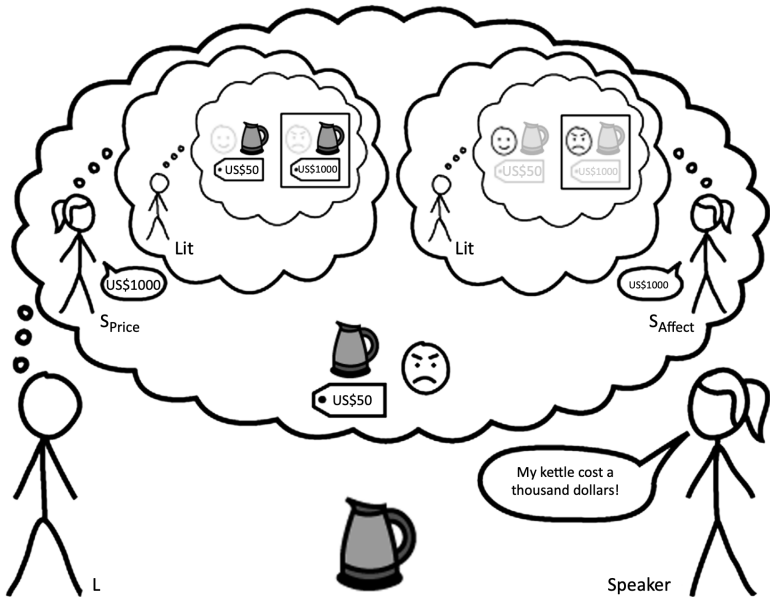


'Price' goal: *How much did S pay?*

'Affect' goal: *How does S feel about the price?*

Kao et al. (2014)





Goodman and Frank (2016)

# Jointly inferring multiple variables

The kettle cost me a thousand dollars!

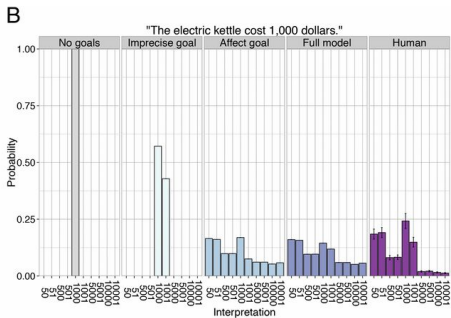
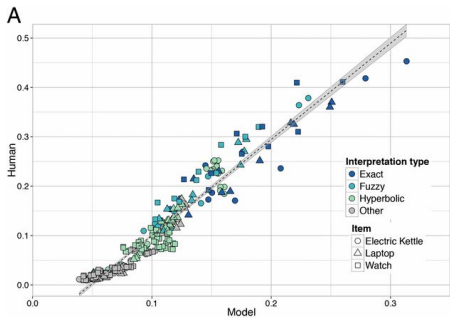


$$S = \{50, 51, \dots, 10001\}$$

$$G = \{\text{price}, \text{affect}\}$$

$$P_L(s, g|u) \propto P_{\text{speaker}}(u|s, g) * P_{\text{state-prior}}(s) * P_{\text{goal-prior}}(g)$$

- If goal were price, "\$1000!" would be communicatively optimal [ $P_{\text{speaker}}(\text{"$1000!"} | 1000, \text{price})$  is high].
- But kettles are unlikely to cost that much [ $P(1000)$  is low].
- $\rightarrow$  Listener explains utterance via affect goal.



# Assessing RSA model fit

Recall: models frequently include free parameters ( $\alpha$  optimality parameter, cost terms on utterances)

Free parameter values often estimated using Bayesian Data Analysis techniques

**Bayes' rule:** for a hypothesis  $H$  and data  $d$ :

$$Pr(H|d) \propto Pr(d|H) * Pr(H)$$

In BDA, *Hypotheses* are model parameterizations, *data* are empirical (often experimental) observations.

- Step 1: 'condition' on experimental data to compute posterior distributions over model parameter values.
- Step 2: sample from posterior to 'run the model forward'
- Step 3: assess predictions against observations

## Some other domains of application

- Scalar inference (the 'drosophila' of pragmatics):  
*John ate some of the cookies* → 'John didn't eat all of the cookies'.  
(Goodman & Frank, 2016; Potts, Lassiter, Levy, & Frank, 2015; Waldon & Degen, 2020, inter alia)
- Linguistic vagueness & imprecision (e.g., *John is tall*, *The townspeople are asleep*):  
Listeners jointly infer speaker meanings and contextual variables  
(thresholds of predication, precision standards)  
(Lassiter & Goodman, 2013; Schuster & Degen, 2020; Waldon, 2022, inter alia)
- Social meaning: listeners jointly infer speaker meanings and social goals (e.g., appear 'relaxed'/'professional')  
(Burnett, 2019)

# NLP application 1: contrastive image captioning



“[H]igh-quality captions are not merely *true*, but also *pragmatically informative* in the sense that they highlight salient properties and help distinguish their inputs from similar images.”

(Cohn-Gordon, Goodman, & Potts, 2018, “Pragmatically Informative Image Captioning with Character-Level Inference”)

- $S_0(\textit{caption}|\textit{image}) \propto$  sequence probability from a neural image captioning model (trained on individual images).
- $L_{lit}(\textit{image}|\textit{caption}) \propto S_0(\textit{caption}|\textit{image})$
- Pragmatic  $S_1(\textit{caption}|\textit{image}) \propto L_{lit}(\textit{image}|\textit{caption})$

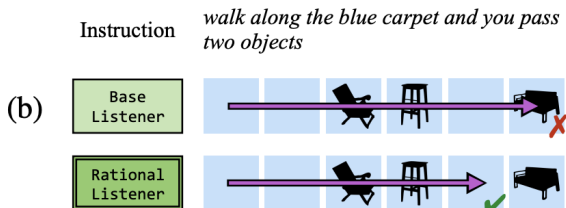
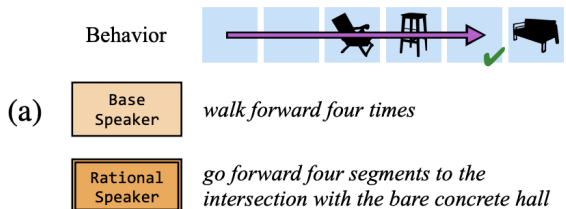


$S_0$  caption: a double decker bus

$S_1$  caption: a red double decker bus

- Implements an incremental, character-level  $S_1$  (more tractable utterance space!)

# NLP application 2: directing agents in navigable space



(Fried, Andreas, & Klein, 2018, “Unified Pragmatic Models for Generating and Following Instructions”)



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