Rational Speech Act models are utterance-independent updates of world priors

Jean-Philippe Bernardy, Julian Grove, and Christine Howes Semdial 2022 - DubDial, August 23, 2022

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Introduction to RSA



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- A pragmatic listener L_1 : $u \mapsto P_{L_1}(w|u)$

The literal listener L_0

$$P_{L_0}(w|u) \propto \mathbb{1}(w \geq n) \times P(w)$$

W	5	6	7
'JP ate 5 cookies'	1/3	1/3	1/3
'JP ate 6 cookies'	0	1/2	1/2
'JP ate 7 cookies'	0	0	1

The pragmatic speaker S_1

$$P_{S_1}(u \mid w) \propto \frac{P_{L_0}(w \mid u)^{\alpha}}{e^{\alpha \times C(u)}}$$

(Assume $\alpha = 4$, and that C(u) is constant.)

W	5	6	7
'JP ate 5 cookies'	1	0.16	0.01
'JP ate 6 cookies'	0	0.84	0.06
'JP ate 7 cookies'	0	0	0.93

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W	5	6	7
'JP ate 5 cookies'	0.85	0.14	0.01
'JP ate 6 cookies'	0	0.93	0.07
'JP ate 7 cookies'	0	0	1

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 - · for computing RSA models;
 - for the algorithmic plausibility of RSA (Marr, 1982).

RSA via information gain

Information gain as K-L divergence

Kullback-Leibler (K-L) divergence (between distributions P and Q over X):

$$D_{\mathsf{KL}}(Q \parallel P) = -\sum_{x \in \mathcal{X}} Q(x) \log \left(\frac{P(x)}{Q(x)} \right)$$

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Higher values: more information is gained by going from P to Q.

A theorem

Theorem

If $Q(x) \propto f(x) \times P(x)$, and the range of f is $\{0, 1\}$, then:

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In words: if f updates P merely by *filtering* it (resulting in Q), then the information gain of Q is the negative log of the expected value of f.

• Important: the expected value of *f* is *also* the normalizing constant for *Q*:

$$Q(x) = \frac{f(x) \times P(x)}{\sum_{x'} f(x') \times P(x')}$$

$$P_{L_0}(w|u) = \frac{l(u,w) \times P(w)}{\sum_{w' \in \mathcal{W}} l(u,w') \times P(w')}$$

where l(u, w) = 1 or 0, depending on whether u is true at w

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Let us define the *literal information gain* of *u*:

$$G_{L_0}(u) = D_{\mathsf{KL}}(P_{L_0}(w|u) \parallel P(w))$$

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Upshot:

- l(u, w) is our filter
- $P_{S_1}(u) \propto e^{\alpha \times (G_{L_0}(u) C(u))}$ is our *new* "prior"!

Reformulating the pragmatic speaker: example

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JP ran u km' $] = w \ge u$

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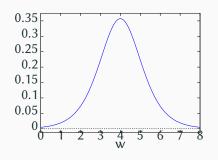
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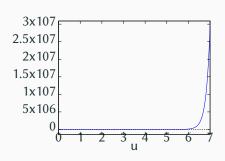
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Prior over JP's running distance

$$P_{S_1}(u|w=7)$$
 with $\alpha=4$

Making the normalizing constant explicit

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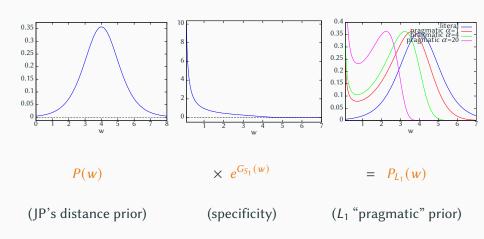
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Upshot:

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Reformulating the pragmatic listener: example



Some consequences

Computing RSA models

Main result:

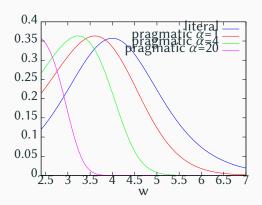
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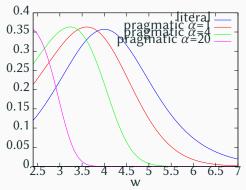


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(Side note: the expected implicature is not always generated!)

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(RSA-style) pragmatic interpretations can be *learned*, just as probability distributions over events can; pragmatic distributions are additionally sensitive to an event's specificity.

References

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