

Wonky worlds: Listeners reconsider common ground when utterances are odd

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Abstract

World knowledge enters pragmatic utterance interpretation in complex ways. Sometimes, a speaker’s utterance suggests that listeners should disregard their world knowledge, yet current models of pragmatic interpretation either disregard the role of world knowledge or overestimate its role in interpretation. Here we provide an extension to the Rational Speech Act model of scalar implicature that captures whether listeners believe they are in an abnormal—or ‘wonky’—world after observing a speaker’s utterance, in which case they downweight their prior beliefs in the computation of speaker meaning. We show in four experiments that a) listeners have varying prior beliefs about the probability of various objects exhibiting an effect (e.g., marbles sinking), b) these beliefs influence listeners’ expectations about how many objects will show the effect after observing an utterance (like *Some of the marbles sank*), c) these beliefs influence scalar implicature strength, and d) listeners’ world wonkiness judgments are affected by the surprise of the observed utterance under their prior beliefs. The extended model is the first quantitative model that accounts for how rational listeners should integrate world knowledge in pragmatic utterance interpretation, and provides a close match to the empirically obtained data.

Keywords: scalar implicature; world knowledge; prior beliefs; experimental pragmatics; computational pragmatics

How often do you think marbles would sink in water? Probably extremely often, if not always. Now imagine reading *Max threw fifteen of his favorite marbles in the water. Some of them sank*. Have you begun to reconsider your assumptions? Perhaps you now suspect that these marbles are in fact made of plastic or the water is covered with thick algae? That is, that they are not just normal marbles in normal water. Here we explore how prior world knowledge enters into pragmatic utterance interpretation, and how this world knowledge is defeasible: some utterances lead listeners to conclude that the world under discussion is abnormal and has appropriately different prior probabilities. We refer to such an abnormal world as a *wonky* world.

The Rational Speech Acts framework (RSA) (Frank & Goodman, 2012; ?, ?), and related game-theoretic models (?, ?), treat communication as a signaling game (?, ?) between a speaker and a listener. The listener reasons by Bayesian inference about what the world is like given that a speaker who produced the utterance is trying to be informative (with respect to a naïve listener). Variants of these models have successfully captured listeners’ quantitative behavior on a number of pragmatic inference tasks, including ad hoc Quantity implicature [ref](#), M-implicature [ref](#), scalar implicature [ref](#), embedded scalar implicatures [ref](#), and non-literal language (?, ?). A defining feature of Bayesian reasoning is that prior beliefs affect inferences that will be drawn. Bayesian models of language interpretation, accordingly, predict that prior beliefs

about the world should affect the listener’s interpretation of an utterance. While this impact of prior knowledge has been noted, and included in models, it hasn’t been systematically studied.

Generalizing our opening example, take the case of *some of the X-s Y-ed* (for category X and event Y). When the prior probability, $\theta_{X,Y}$, of an X Y -ing is not extreme, RSA leads to the standard scalar implicature: the posterior probability that all 15 of the X s Y -ed, after hearing the utterance, is much lower than its prior probability. This is because a rational speaker would have been expected to say the more informative *all of the X-s Y-ed* if it had been true. As we will show below, RSA makes two strong predictions about the effect of the prior: (1) As $\theta_{X,Y}$ approaches 1, the interpretation probability that all X -s Y -ed approaches 1, that is, the scalar implicature disappears. This prediction follows because the extreme prior overwhelms the effect of the utterance. (2) The posterior expectation of the number of X -s that Y -ed **should be approximately the same as the prior expectation, when the number of tries is large (e.g. 15)**—UPDATE. . This prediction follows from the weak semantics of *some* and the isolated effect of the alternative *all*: because *some of the X-s Y-ed* only restricts the interpretation to be greater than zero and the scalar implicature resulting from alternative *all of the X-s Y-ed* can at the most rule out the state in which all X -s Y -ed, the prior will dominate the inference of exactly how many X -s Y -ed.

However, intuition is at odds with these predictions: for example, Geurts (2010) has observed that for events with very high prior probability of occurrence (e.g. marbles sinking), observing an utterance of *Some of the marbles sank* leads to very strong implicatures, that is, the subjective probability that all of the marbles sank is intuitively close to 0. In Exp. 1 we collect prior probabilities for a variety of events and categories. In experiment 2a and 2b we collect posterior interpretations after hearing utterances such as *some of the X-s Y-ed*. These experimental results confirm the intuition of relatively strong implicature—hence prediction (1) of RSA is incorrect—and show that the prior has a muted effect on posterior expectation—hence prediction (2) of RSA is incorrect. Given the previous success of RSA models, this constitutes a striking puzzle. To address this puzzle we pursue the intuition raised at the very beginning of this paper: that sometimes, the speaker’s utterance will lead the listener to infer that the world under discussion is wonky and she should therefore down-weight her prior beliefs in the computation of speaker meaning. In Experiment 3 we explore participants’ intuitions about whether the world is normal in the scenarios of Experiment 2. We then introduce a variant of RSA, wRSA,

in which the listener can revise her beliefs about the domain that the speaker intended. We show that this extension resolves the puzzle of the prior’s muted effects, and predicts listeners’ wonkiness judgements. **UPDATE IN LIGHT OF REORGANIZATION**

Experiment 1

Exp. 1 measured listeners’ prior beliefs about how many objects would exhibit a certain effect (e.g., how many marbles sink).

Method

Participants We recruited 60 participants over Amazon’s crowd-sourcing platform Mechanical Turk.

Procedure and materials On each trial,¹ participants read a one-sentence description of an event like *John threw 15 marbles into a pool*. They were then asked to provide a judgment of an effect, e.g. *How many of the marbles do you think sank?*, on a sliding scale from 0 to 15. Each item had a similar form: the first sentence introduced the objects at issue (e.g., marbles). The question always had the form *How many of the X Yed?*, where *X* was the object noun phrase introduced in the first sentence (e.g., *marbles*, *cups*, *balloons*) and *Yed* was a verb phrase denoting an effect that the objects underwent (e.g., *sank*, *broke*, *stuck to the wall*). Each verb phrase occurred with three different objects, e.g., *sank* occurred with *marbles*, *cups*, and *balloons*. Items were constructed to intuitively cover the range of probabilities as much as possible, while also somewhat oversampling the upper range of probabilities to have more fine-grained coverage of this region that is of most interest for testing the RSA model. Judgments were obtained for 90 items, of which each participant saw a random selection of 30 items.

Results

Data from one participant, who gave only one response throughout the experiment, were excluded. Each item received between 12 and 29 ratings. Distributions of ratings for each item were smoothed using **Laplace smoothing**. As intended, items covered a wide range of probabilities. **include measure or plot of this once you decide which smoothing to use ultimately.**

In the next section, we use these empirically obtained prior beliefs to derive RSA predictions for the interpretation of utterances like *Some of the marbles sank*, before empirically measuring participants’ interpretations.

Effect of the world prior in RSA

The basic Rational Speech Acts model defines a pragmatic listener $P_{\text{listener}}(s|u)$, who reasons about a speaker $P_{\text{speaker}}(u|s)$, who in turn reasons about a literal listener $P_{\text{literal}}(s|u)$. Each listener does Bayesian inference about the

¹This experiment can be viewed at https://www.stanford.edu/jdegen/12_sinking-marbles-prior15/sinking-marbles-prior.html

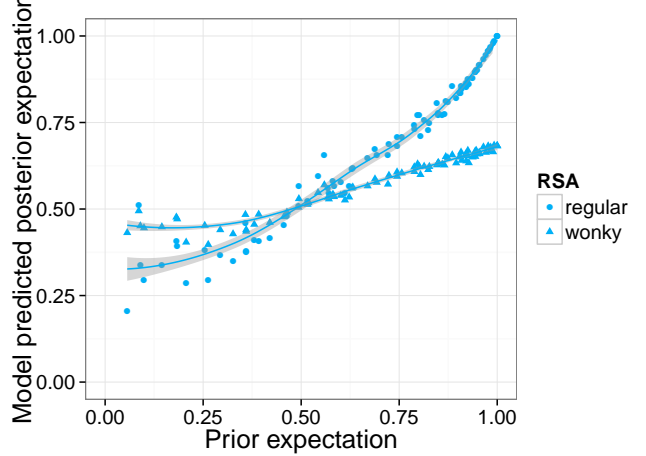


Figure 1: For each item, rRSA and wRSA model predicted mean empirical proportion of affected objects after observing *Some of the X Yed*, as a function of prior mean proportion of affected objects.

world state, given either the literal truth of utterance u or the speaker’s choice of u ; the speaker is a softmax-optimal decision maker, with the goal of being informative about the state s . RSA is defined by:

$$P_{\text{literal}}(s|u) \propto F_u(s) \cdot P(s) \quad (1)$$

$$P_{\text{speaker}}(u|s) \propto \exp(\lambda \ln P_{\text{literal}}(s|u)) \quad (2)$$

$$P_{\text{listener}}(s|u) \propto P_{\text{speaker}}(u|s) \cdot P(s) \quad (3)$$

Here $F_u : s \mapsto \{0, 1\}$ is a truth-function specifying the meaning of each utterance.

For concreteness, assume that the set of states of the world is $S = \{s_0, s_1, s_2, \dots, s_{15}\}$, where the subscript indicates the number of objects (e.g., marbles) that exhibit a certain effect (e.g., sinking). Assume also that the set of utterances *All/None/Some of the marbles sank* is denoted $U = \{u_{\text{all}}, u_{\text{none}}, u_{\text{some}}\}$ and each has its usual literal meaning: $F_{u_{\text{none}}}(s) = \delta_{s=0}$, $F_{u_{\text{some}}}(s) = \delta_{s>0}$, $F_{u_{\text{all}}}(s) = \delta_{s=15}$. Let us imagine that the prior is binomial with single event probability θ : $P(s) = \text{Binomial}(s|\theta)$.

In Figure ?? we show the predictions of RSA as θ varies. **put in figure showing the basic RSA predictions for allstate $P(s_{15}|u_{\text{some}})$ and expectation. this figure could also have wRSA predictions for later reference.** We see that the prior has a strong effect, which can be summarized by the two predictions described above. We next turn to an empirical test of these predictions, or rather, of the intuition that they may be incorrect.

Experiment 2a

Exp. 2a measured participants’ posterior beliefs $p(s|u)$ in how many objects exhibited a certain effect (e.g., marbles sinking), after observing an utterance.

Participants We recruited 120 participants over Amazon's crowd-sourcing platform Mechanical Turk.

Procedure and materials ²

Participants read the same descriptions as in Exp. 1. They additionally saw an utterance produced by a knowledgeable speaker about the event, e.g. *John, who observed what happened, said: "Some of the marbles sank"*, and, as in Exp. 1, were asked to provide a judgment of an effect, e.g. *How many of the marbles do you think sank?*, on a sliding scale from 0 to 15.

Each participant saw 10 *some* trials and 20 fillers, of which 10 contained the quantifiers *all* or "none", and the rest were utterances that did not address the number of objects that displayed the effect, e.g. *What a stupid thing to do*. The utterances were randomly paired with 30 random items for each participant.

Results and discussion XXX question

The mean number of objects judged to exhibit the effect increased with increasing expectation of the prior distribution ($\beta=.18$, $SE=.02$, $t=7.4$, $p<.0001$, see also Figure 2).

, replicating the effect observed in Exp. 2a. Again, the effect of the prior was much smaller than predicted by rRSA and resulted in mean proportions of affected objects between 30% and 65%, where rRSA predicts a range from XXX to XXX for these items.

Results XXX question

$p(s_{\forall}|u_{\text{some}})$ increased with increasing talk about both $p(s_{\forall})$ and the prior expectation of the distribution? $p(s_{\forall})$ ($\beta=.1$, $SE=.01$, $t=6.9$, $p<.0001$); however, mean $p(s_{\forall}|u_{\text{some}})$ was never higher than .26, suggesting that a) participants drew strong implicatures in this paradigm and b) the effect of $P(s)$ is much smaller than predicted by rRSA.

Experiment 2b

Exp. 2b replicates the effect of the prior on participants' posterior estimates of different objects exhibiting a certain effect (e.g., marbles sinking) using a different dependent measure.

Participants We recruited 120 participants over Amazon's crowd-sourcing platform Mechanical Turk.

Procedure and materials ³

The procedure and materials were identical to those of Exp. 2a with the exception of the dependent measure. Rather than providing estimates of how many objects exhibited an effect, they rated on sliding scales with endpoints labeled "very unlikely" and "very likely", how likely they thought 0%, 1-50%, 51-99%, or 100% of the objects exhibited the effect.

need to discuss fillers, and that this means the muted effect

²This experiment can be viewed at <https://web.stanford.edu/jdegen/sinking-marbles-nullutterance.html>

³This experiment can be viewed at https://www.stanford.edu/jdegen/13_sinking-marbles-prior0v-15/sinking-marbles.html

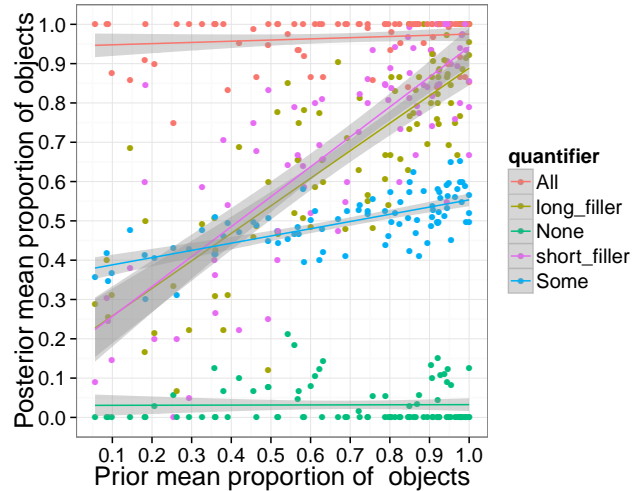


Figure 2: For each item, mean empirical proportion of affected objects after observing an utterance, as a function of prior mean proportion, for different quantifiers and filler conditions. two panels, one for 2a and one for 2b?

of prior is not because Ss are insensitive to it.

Exps. 2a and 2b demonstrate that there is an effect of listeners' prior beliefs on the interpretation of utterances with *some*. However, this effect is quantitatively much smaller than predicted by rRSA, and qualitatively does not show the critical limit effect (converging to the upper-right corner as seen in Fig. 1). earlier be clear about the prediction that as prior goes to one allprob and expectation should go to one. reference that qualitative prediction of RSA here and in 2a results.

Discussion of why; wonky world intuition

Experiment 3

Exp. 3 measured participants' beliefs in world wonkiness after observing the scenarios and utterances from Exps. 2a and 2b.

Participants We recruited 60 participants over Amazon's crowd-sourcing platform Mechanical Turk.

Procedure and materials ⁴

The procedure and materials were identical to those of Exps. 2a and 2b, with the exception of the dependent measure. Rather than providing estimates of what they believed the world was like, participants were asked to indicate how likely it was that the objects (e.g., the marbles) involved in the scenario were normal objects, by adjusting a slider that ranged from *definitely not normal* to *definitely normal*.

Results The extreme ends of the sliders were coded as 1 (*definitely not normal*, i.e., wonky) and 0 (*definitely normal*,

⁴This experiment can be viewed at https://web.stanford.edu/jdegen/17_sinking-marbles-normal-sliders/sinking-marbles-normal.html

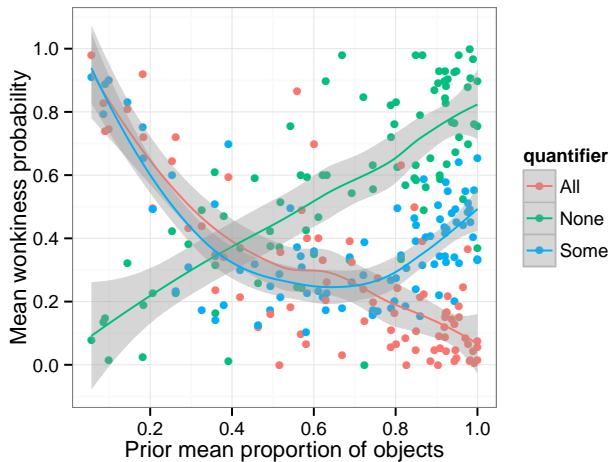


Figure 3: For each item, mean wonkiness probability after observing an utterance, as a function of expected prior proportion of affected objects, for different quantifiers.

i.e., not wonky). We interpret the slider values as probability of world wonkiness. Mean wonkiness probability ratings are shown in Figure 3. For *all* and *none*, increasing prior expectation of objects exhibiting the effect resulted in a fairly linear decrease and increase in the probability of wonkiness, respectively. For *some*, the pattern is somewhat more intricate: probability of wonkiness initially decreases sharply, but rises again in the upper range of the prior expected value.

yay!! XXX

Model

The wonky RSA model we propose for capturing the defeasibility of listeners' world knowledge is an extension of regular RSA: in wRSA, the listener infers the value of θ_{wonky} jointly with s . θ_{wonky} captures for each utterance and item, how likely the objects involved in the event (e.g., marbles) are in fact "wonky" (in which case the computation draws on a uniform prior, i.e. disregards prior beliefs) or not (in which case the model draws on the smoothed empirical prior distribution for that item, obtained in Exp. 1). The resulting $p(s|u)$ is a mixture of computations based on the uniform and empirical prior, with mixture parameter θ_{wonky} . The inferred value of θ_{wonky} itself depends on $p(u|s)$: the more surprising a particular utterance is given prior beliefs, the higher the probability of θ_{wonky} .

NDG: note to self – difference between revising own beliefs and revising common ground....

Model evaluation

From the abstract: Comparing the fit of rRSA and wRSA model predictions to the posterior state estimates from Exp. 2 yields a much better fit for wRSA. The better fit of wRSA suggests that listeners use speakers' utterances as cues to how strongly to incorporate world knowledge. wRSA also pro-

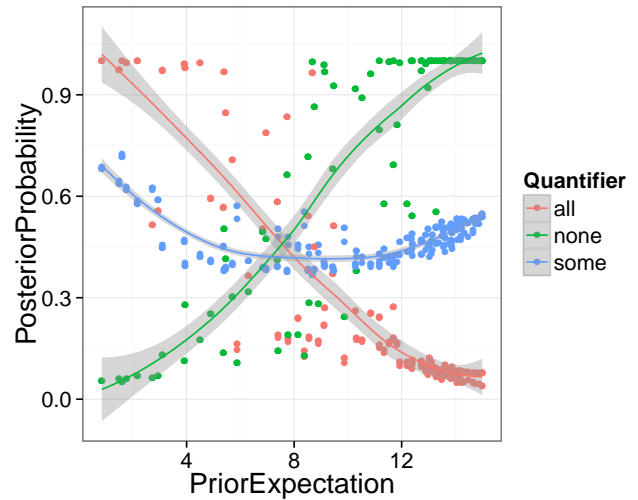


Figure 4: For each item, predicted proportion of 'wonky' ratings after observing an utterance, as a function of prior mean proportion, for different quantifiers.

vided a better fit than a model which used only a uniform prior, confirming that listeners do make use of world knowledge in a systematic way in the computation of scalar implicature.

it's possible we'd get less noise from some more stable estimator of prior. consider trying the plots with prior mode and median as x-axis.... or inferred binomial prob fit to each prior, if the fits are at all decent.

Discussion and conclusion

- what is wonky?
- other ways of asking about wonkiness
- what's the right prior to back off to?
- revising private beliefs vs revising common ground.
- connection to presupposition (cf stalnaker), and other phenomena
- implication for experiments on language understanding

References

- Frank, M. C., & Goodman, N. D. (2012). Predicting pragmatic reasoning in language games. *Science*, 336, 998.
- Geurts, B. (2010). *Quantity implicatures*. Cambridge: Cambridge Univ Press.