

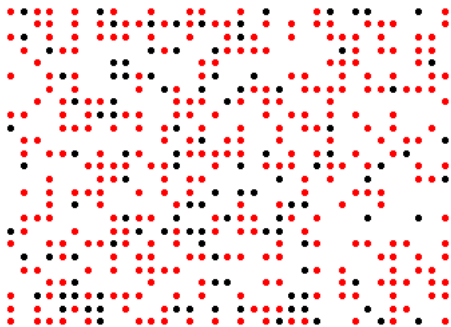
# Speaking of quantifiers

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— of the circles are red.

## Generalised quantifiers

Many quantity words place thresholds on the absolute or relative intersection set size.

- > 'some' means that  $|S \cap P| > 0$
- > 'fewer than half' means that  $|S \cap P|/|S| < 0.5$
- > 'most' means that  $|S \cap P|/|S| > 0.5$
- > 'exactly three' means that  $|S \cap P| = 3$
- > 'many' means that  $|S \cap P| > t$

## Monotonicity

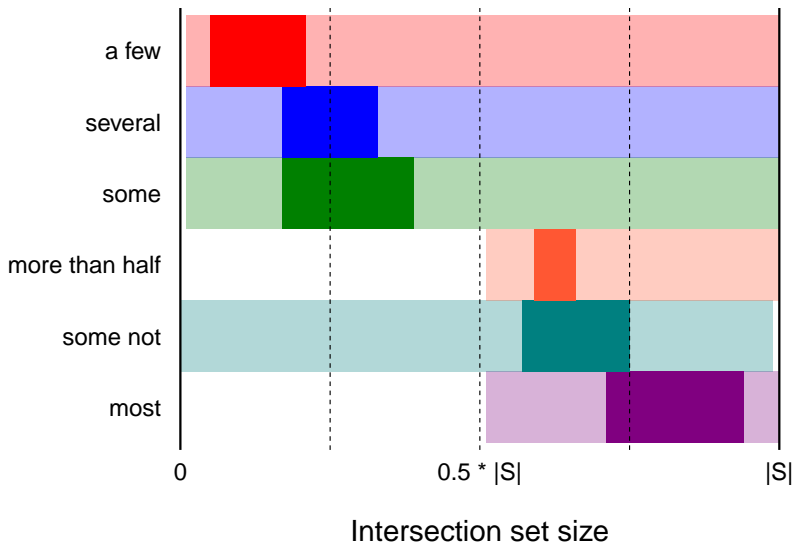
Quantity words can be...

- > monotone: licensing inferences from sets to supersets.
- > antitone: licensing inferences from sets to subsets.
- > neither.

Monotone quantity words are learned earlier (Katsos et al., 2016) and processed faster (Geurts et al., 2010). Antitone quantity words allow for negative polarity items (Zwarts, 1981).

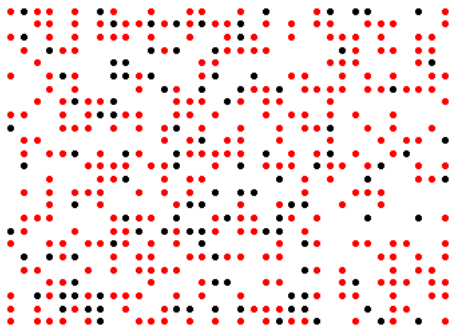
## **From meaning to use**

As it stands, GQ theory offers a poor account of the way speakers and hearers actually use quantity words (e.g, Hammerton, 1976; Newstead et al., 1987).



## **Production study**

- > 600 participants on Mechanical Turk.
- > Displays consisting of 432 black or red circles.
- > Participants saw 10 displays.
- > They had to describe the number of red circles by completing the sentence '— of the circles are red'.
- > Number words were not allowed.

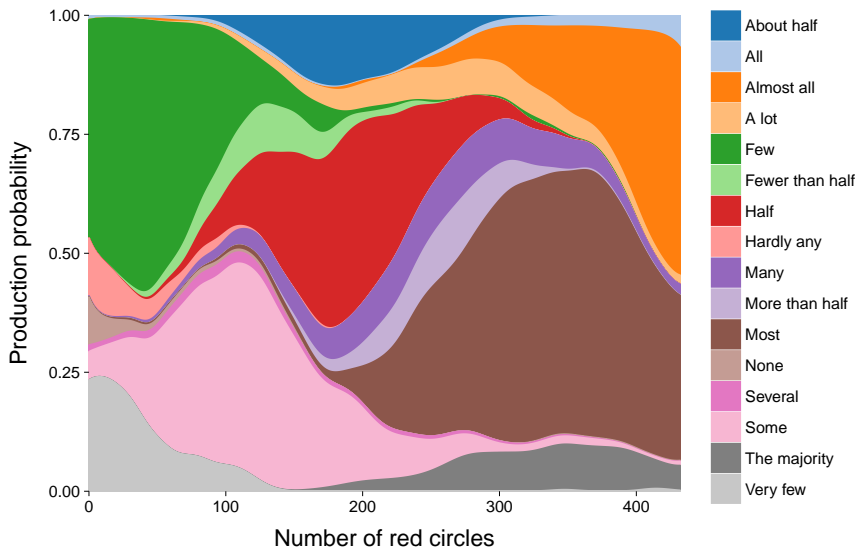


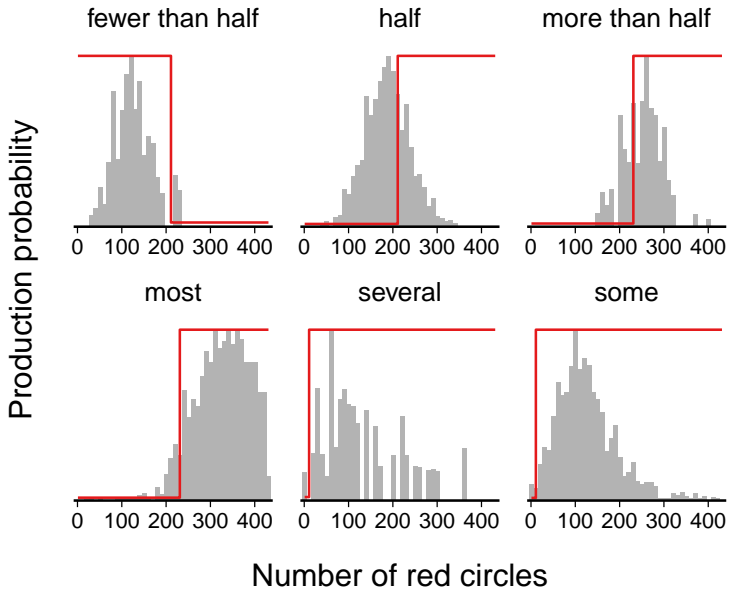
— of the circles are red.



## Results

- > 291 unique responses out of 5630 observations.
  - ‘all’
  - ‘less than a fraction of a third’
- > Reduced to 35 types of quantity words.
- > 16 types of quantity words produced more than 50 times.
- > ‘Most’, ‘few’, ‘some’, ‘half’, and ‘almost all’ together make up more than half of the responses.





## **Focal ranges**

The focal ranges of quantity words often diverge from their semantic ranges. Can we explain the presence and location of these focal ranges within GQ theory?

## Prolegomena

- > 433 possible intersection set sizes:  $T = \{t_0, \dots, t_{432}\}$ .
- > 16 quantity words:  $M = \{m_{\text{'a lot'}}, \dots, m_{\text{'very few'}}\}$ .
- > The probability that  $S$  produces  $m$  to refer to  $t$  depends on...
  - the salience of  $m$ .
  - the perceived intersection set size  $t'$ .
  - the meaning of  $m$ .

## Saliency

The saliency *Sal* of *m* is proportional to the frequency of *m* in the production study.

$$P_{Sal}(m) \propto n_m$$

## Confusability

The perceived intersection set size  $t'$  depends on...

- > the distance between  $t$  and the limits.
- > the accuracy of the estimates.

We formulate confusability by analogy to confusion probabilities of the approximate number system (Dehaene, 1997).

$$P_{Conf}(t' | t; w) \propto P_{ANS}(t' | t; w) P_{ANS}(432 - t' | 432 - t; w)$$

$$P_{ANS}(t' | t; w) = \int_{t'-0.5}^{t+0.5} \mathcal{N}(x, \mu = t, \sigma = w t) dx .$$

## Meaning

$B_\theta$  maps pairs of  $m$  and  $t$  to the truth value of  $m$  in  $t$ .  $B_\theta$  is parameterized, and we will use data-driven inference to determine best-fitting parameter values below.

The monotonicity of quantity words is gauged experimentally.



## **Monotonicity study**

- > 60 participants on Mechanical Turk.
- > Participants were presented with arguments containing inferences from sets to subsets and supersets.
- > They had to determine the validity of these arguments

### **Upward argument**

Premise: Q of the people P1.

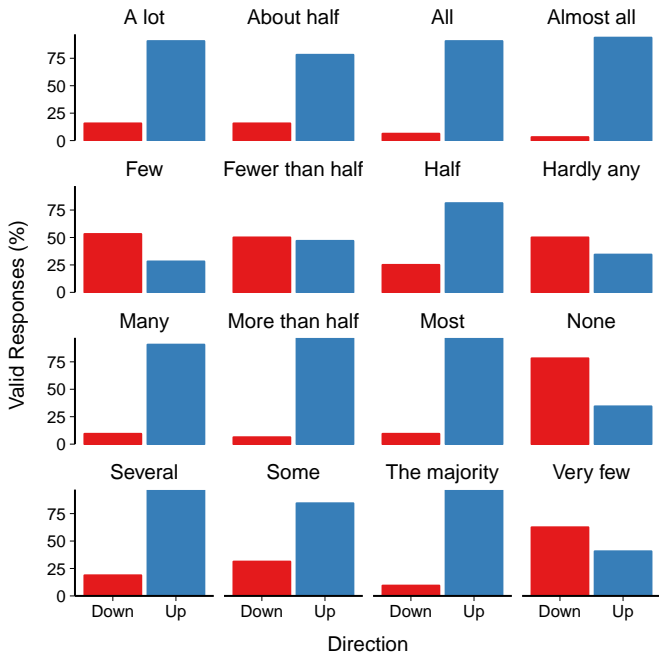
Conclusion: Q of the people P2.

### **Downward argument**

Premise: Q of the people P2.

Conclusion: Q of the people P1.

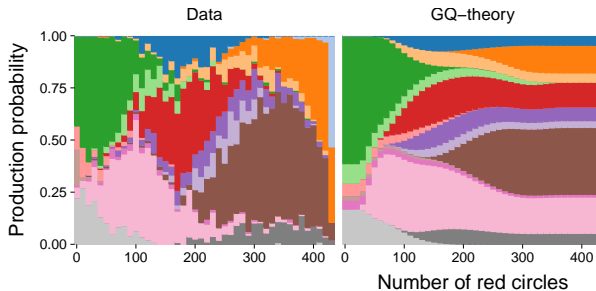
... where  $P1 \subseteq P2$



## A literal speaker

A literal speaker produces quantity words with a probability that is proportional to its semantic value.

$$P_{S_{Lit}}(m | t; \theta, \mathbf{w}) \propto \sum_{t' \in T} P_{Conf}(t' | t; \mathbf{w}) P_{Sal}(m) B_{\theta}(m, t')$$



## Model fit

The semantic model offers a reasonably good fit to the data ( $r = .809$ ). But it fails to account for the focal ranges.

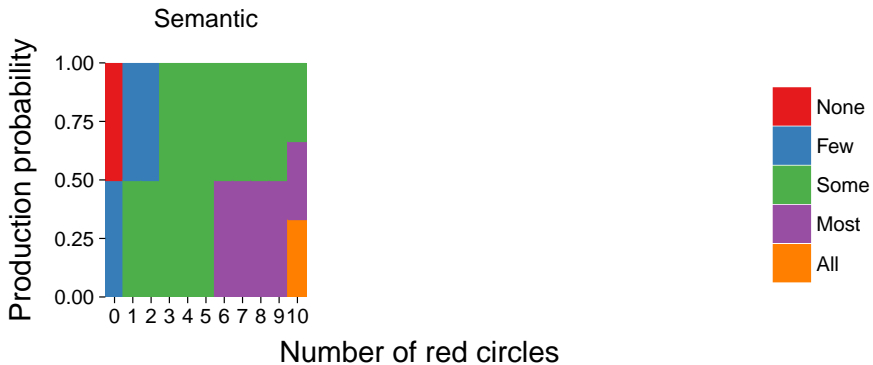
- ‘about half’ ( $r = .04$ )
- ‘fewer than half’ ( $r = .38$ )
- ‘none’ ( $r = .36$ )

Can we improve the model by taking into consideration the pragmatics of communication?

## Semantics

Suppose the speaker can produce 5 quantity words to describe displays with 10 circles.

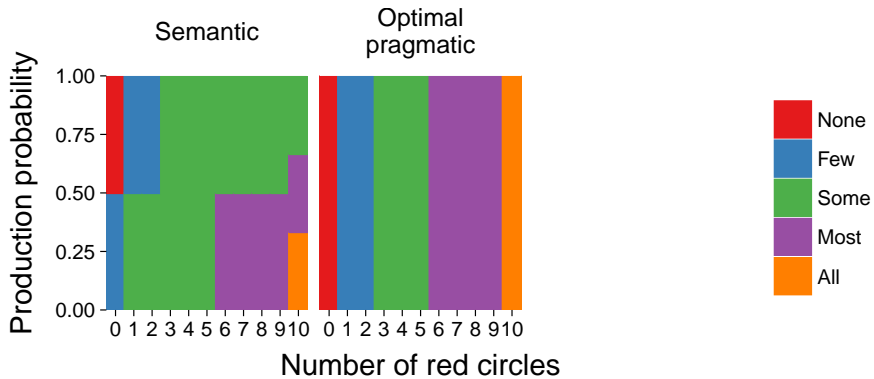
- > 'no' means that  $|S \cap P| = 0$
- > 'few' means that  $|S \cap P| < 2$
- > 'some' means that  $|S \cap P| > 0$
- > 'most' means that  $|S \cap P| > 5$
- > 'all' means that  $|S \cap P| = 10$





## **Pragmatics**

A completely rational pragmatic speaker chooses messages that are most likely to be interpreted correctly.



## **Probabilistic pragmatics**

A probabilistically pragmatic speaker would choose the optimal message with a probability proportional to her rationality.



## A pragmatic speaker

A pragmatic speaker optimises her behaviour based on how a naive hearer would behave.

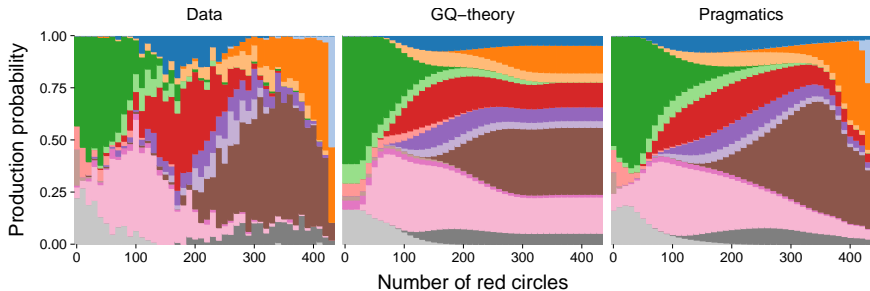
$$P_H(t | m) \propto B_\theta(m, t)$$

A pragmatic speaker optimises the probability of coordination with  $H$ .  $\lambda$  regulates how likely it is that the pragmatic speaker produces the optimal quantity word (Frank & Goodman, 2012).

$$P_S(m | t, \lambda) \propto P_H(t | m)^\lambda$$

Taken together, a pragmatic speaker is defined as follows.

$$P_{S_{Prg}}(m | t; \theta, \mathbf{w}, \lambda) \propto \sum_{t' \in T} P_{Conf}(t' | t; \mathbf{w}) P_{Sal}(m) P_S(m | t'; \theta, \lambda)$$



## Results

- >  $S_{Prq}$  offers a substantially better explanation of the production data than  $S_{Lit}$  ( $r = .809$  vs.  $.924$ ).
- > The difference is most pronounced for...
  - ‘about half’ ( $r = .04$  vs.  $.98$ ).
  - ‘fewer than half’ ( $r = .38$  vs.  $.82$ ).
  - ‘none’ ( $r = .36$  vs.  $1.00$ ).
- > GQ-theory offers a compelling account of the use of quantity words if it is supplemented with a pragmatic module.

## Conclusions

- > Gradient structure may emerge as a consequence of pragmatic reasoning and constraints on perception.
  - Vagueness.
  - Prototypes.
- > Further improvements to the model: individual differences in the location of the thresholds for vague quantity words.



**Thank you!**