Using contrastive inferences to learn about new words and categories

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9 Abstract

In the face of unfamiliar language or objects, description is one cue people can use to learn 10 about both. Beyond narrowing potential referents to those that match a descriptor (e.g., 11 "tall"), people could infer that a described object is one that contrasts with other relevant 12 objects of the same type (e.g., "the tall cup" contrasts with another, shorter cup). This 13 contrast may be in relation to other objects present in the environment (this cup is tall 14 among present cups) or to the referent's category (this cup is tall for a cup in general). In 15 three experiments, we investigate whether people use such contrastive inferences from 16 description to learn new word-referent mappings and learn about new categories' feature 17 distributions. People use contrastive inferences to guide their referent choice, though 18 size—and not color—adjectives prompt them to consistently choose the contrastive target 19 over alternatives (Experiment 1). People also use color and size description to infer that a 20 novel object is atypical of its category (Experiments 2 and 3). However, these two inferences do not trade off substantially: people infer a described referent is atypical even when the descriptor was necessary to establish reference. We model these experiments in the Rational Speech Act (RSA) framework and find that it predicts both of these inferences, and a very small trade-off between them—consistent with the non-significant trade-off we observe in 25 people's inferences. Overall, people are able to use contrastive inferences from description to resolve reference and make inferences about a novel object's category, allowing them to learn 27 more about new things than literal meaning alone allows. 28

Keywords: concept learning; contrastive inference; word learning; pragmatics; communication; computational modeling

Word count: 11021

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An utterance can say much more about the world than its literal interpretation might 33 suggest. For instance, if you hear a colleague say "We should hire a female professor," you 34 might infer something about the speaker's goals, the makeup of a department, or even the 35 biases of a field—none of which is literally stated. These inferences depend on recognition that a speaker's intended meaning can differ from the literal meaning of their utterance, and 37 the process of deriving this intended meaning is called pragmatics. Frameworks for understanding pragmatic inference posit that speakers tend to follow general principles of conversation—for instance, that they tend to be relevant, brief, and otherwise helpfully informative (Clark, 1990; Grice, 1975; Sperber & Wilson, 1986). When a speaker deviates from these principles, a listener can reason about the alternative utterances the speaker might have said and infer some intended meaning that goes beyond the literal meaning of their utterance.

Pragmatic inference is also a potentially powerful mechanism for learning language.

People can learn the meanings of words by tracking statistical properties of their literal

meaning alone (Yu & Smith, 2007), but reasoning about a speaker's intended meaning—and

not just the words they say—may support more rapid and accurate learning (Frank,

Goodman, & Tenenbaum, 2009). For example, Akhtar, Carpenter, and Tomasello (1996)

showed that young children can infer the meaning of a new word by using the principle that

people tend to remark on things that are new and interesting to them. In this study, an

experimenter leaves the room and a new toy emerges in her absence; once she comes back,

the toy is familiar to the child but not to the experimenter. When she uses a novel name,

"gazzer," the child can infer that the word refers to the toy that is novel to the experimenter,

and not other toys the experimenter had already seen. Experiments with adults show that

they too can use general principles of informativeness to infer a novel referent's name (Frank

& Goodman, 2014).

One potential pragmatic tool for learning about referents is contrastive inference from 58 description. To the extent that communicators strive to be minimal and informative, 59 description should discriminate between the referent and some relevant contrasting set. This 60 contrastive inference is fairly obvious from some types of description, such as some 61 postnominal modifiers: "The door with the lock" clearly implies a contrasting door without one (Ni, 1996). The degree of contrast implied by more common descriptive forms, such as 63 prenominal adjectives in English, is less clear: speakers do not always use prenominal adjectives minimally, often describing more than is needed to establish reference (Engelhardt, Barış Demiral, & Ferreira, 2011; Mangold & Pobel, 1988; Pechmann, 1989). Nevertheless, Sedivy, Tanenhaus, Chambers, and Carlson (1999) showed that people can use these inferences to resolve referential ambiguity in familiar contexts. When asked to "Pick up the tall cup," people directed their attention more quickly to the target when a short cup was present, and did so in the period before they heard the word "cup." Because the speaker would not have needed to specify "tall" unless it was informative, listeners were able to use the adjective to direct their attention to a tall object with a shorter counterpart. Subsequent work using similar tasks has corroborated that people can use contrastive inferences to direct 73 their attention among familiar referents (Aparicio, Xiang, & Kennedy, 2016; Ryskin, Kurumada, & Brown-Schmidt, 2019; Sedivy, 2003).

But what if you didn't know the meaning of the key words in someone's

utterance—could you use the same kind of contrastive inferences to learn about new words

and categories? Suppose a friend asks you to "Pass the tall dax." Intuitively, your friend

must have said the word "tall" for a reason. One possibility is that your friend wants to

distinguish the dax they want from another dax they do not. In this case, you might look

around the room for two similar things that vary in height, and hand the taller one to them.

If, alternatively, you only see one object around whose name you don't know, you might

draw a different inference: this dax might be a particularly tall dax. In this case, you might

think your friend used the word "tall" for a different reason—not to distinguish the dax they

want from other daxes around you, but to distinguish the dax they want from other daxes in
the world. This would be consistent with data from production studies, in which people tend
to describe atypical features more than they describe typical ones (Mitchell, Reiter, &
Deemter, 2013; Rubio-Fernández, 2016; Westerbeek, Koolen, & Maes, 2015). For instance,
people almost always say "blue banana" to refer to a blue banana, but almost never say
"yellow banana" to refer to a yellow one.

In each of these cases, you would have used a pragmatic inference to learn something 91 new. In the second case, you would have learned the name for a novel category "dax," and also something about the typical of size of daxes: most of them are shorter than the one you 93 saw. In the first case, you would have resolved the referential ambiguity in the speaker's utterance. But would have you learned something about the typical size of daxes as well, 95 beyond the daxes you observed? One possibility is that you would not: You can explain your friend's use of "tall" as being motivated by the need to distinguish between the two daxes in the room, and thus you should infer nothing about the other daxes in the world. If reference is the primary motivator of speakers' word choice, as implicitly assumed in much research (e.g., Pechmann, 1989; Arts, Maes, Noordman, & Jansen, 2011; Engelhardt et al., 2011), 100 then people should draw no further inferences once the need for referential disambiguation 101 can explain away a descriptor like "tall." On this reference-first view, establishing reference 102 has priority in understanding the utterance, and any further inferences are blocked if the 103 utterance is minimally informative with respect to reference. If, on the other hand, 104 pragmatic reasoning weighs multiple goals simultaneously—here, reference and conveying 105 typicality-people may integrate typicality as just one factor the speaker considers in using description. On this probabilistic weighing view, people can use description to make graded inferences about the referent's identity and about its category's features, and the fact that 108 an adjective would have helped identify the referent does not completely block an inference 109 about atypicality. 110

In this paper, we present a series of experiments that test two ways in which people 111 could use pragmatic inference to learn about novel categories. First, we examine whether 112 listeners use contrastive inference to resolve referential ambiguity. In a reference game, 113 participants saw groups of novel objects and were asked to pick one with a referring 114 expression, e.g., "Find the small toma." If people interpret description contrastively, they 115 should infer that the description was necessary to identify the referent—that the small toma 116 contrasts with some different-sized toma on the screen. We show that people can use 117 contrastive inference—even with unfamiliar objects—to resolve reference and thus to learn the 118 meaning of the new word "toma." 119

Second, we test whether people use contrastive inference to learn about a novel 120 category's feature distribution. Participants were presented with two interlocutors who 121 exchange objects using referring expressions, such as "Pass me the blue toma." If people 122 interpret description as contrasting with an object's category, they should infer that in 123 general, few tomas are blue. Crucially, we vary the object contexts such that in some 124 contexts, the adjective is necessary to establish reference, and in others, it is superfluous. 125 Overall, we show that people can use contrastive inferences both to establish reference and 126 to make inferences about novel categories' feature distributions, and that they do not trade 127 off strongly between these two inferences. We extend a version of the Rational Speech Act 128 model (Frank & Goodman, 2014) that captures how listeners' reasoning about speakers 120 reflects a graded integration of informativity with respect to both reference and typicality. 130

In order to determine whether people can use contrastive inferences to disambiguate referents and learn about categories' feature distributions, we use reference games with novel objects. Novel objects provide both a useful experimental tool and an especially interesting testing ground for contrastive inferences. These objects have unknown names and feature distributions, creating the ambiguity that is necessary to test referential disambiguation and category learning. Testing pragmatic inference in novel, ambiguous situations lays the

groundwork to determine the role of pragmatic inference in learning language. Much work
has focused on how pragmatic inference enriches literal meaning when the literal meaning is
known—when the words and referents in play are familiar. Here, we ask: can people use
pragmatic inferences from description to learn about unfamiliar things in the world?

Experiment 1

In Experiment 1, we ask whether people use descriptive contrast to identify the target 142 of an ambiguous referring expression. Our experiment was inspired by work from Sedivy et 143 al. (1999) showing that people can use contrastive inferences to guide their attention to 144 referents as utterances progress. In their task, participants saw displays of four objects: a 145 target (e.g., a tall cup), a contrastive pair (e.g., a short cup), a competitor that shares the 146 target's feature but not category (e.g., a tall pitcher), and an irrelevant distractor (e.g., a 147 key). Participants then heard a referring expression: "Pick up the tall cup." Participants 148 looked more quickly to the correct object when the utterance referred to an object with a 149 same-category contrastive pair (tall cup vs. short cup) than when it referred to an object 150 without a contrastive pair (e.g., when there was no short cup in the display). 151

Their results suggest that listeners expect speakers to use prenominal description when 152 they are distinguishing between potential referents of the same type, and listeners use this 153 inference to rapidly allocate their attention to the target as an utterance progresses. This 154 principle does not apply equally across adjective types, however: color adjectives seem to 155 hold less contrastive weight (Sedivy, 2003), perhaps because color adjectives are often used 156 redundantly in English-that is, people describe objects' colors even when this description is not necessary to establish reference (Pechmann, 1989). Kreiss and Degen (2020) demonstrate 158 that listeners' familiar referent choices closely conform to speakers' production norms, such 159 that over-specified modifiers hold less contrastive weight. If this generalizes to novel object 160 choice, we should find that size adjectives prompt stronger contrastive inferences than color 161 adjectives. 162

In a pre-registered referential disambiguation task, we presented participants with 163 arrays of novel fruit objects. On critical trials, participants saw a target object, a lure object 164 that shared the target's critical feature but not its shape, and a contrastive pair that shared 165 the target's shape but not its critical feature (Fig. 1). Participants heard an utterance, 166 sometimes mentioning the critical feature: "Find the [blue/big] toma." In all trials, 167 utterances used the definite determiner "the," which conveys that there is a specific referent 168 to be identified. For the target object, which had a same-shaped counterpart, use of the 169 adjective was necessary to establish reference. For the lure, which was unique in shape, the 170 adjective was relatively superfluous description. If participants use contrastive inference to 171 choose novel referents, they should choose the target object more often than the lure. To 172 examine whether contrast occurs across adjective types, we tested participants in two 173 conditions: color contrast and size contrast. Though we expected participants to shift toward choosing the item with a contrastive pair in both conditions, we did not expect them 175 to treat color and size equally. Because color is often used redundantly in English while size is not, we expected size to hold more contrastive weight, encouraging a more consistent 177 contrastive inference (Pechmann, 1989). The pre-registration of our method, recruitment 178 plan, exclusion criteria, and analyses can be found on the Open Science Framework here: 179 https://osf.io/pqkfy. 180

$_{^{181}}$ \mathbf{Method}

Participants. We recruited a pre-registered sample of 300 participants through
Amazon Mechanical Turk. Half of the participants were assigned to a condition in which the
critical feature was color (stimuli contrasted on color), and the other half were assigned to a
condition in which the critical feature was size. Each participant gave informed consent and
was paid \$0.30 in exchange for their participation.

Stimuli. Stimulus displays were arrays of three novel fruit objects. Fruits were chosen randomly at each trial from 25 fruit kinds. Ten of the 25 fruit drawings were adapted

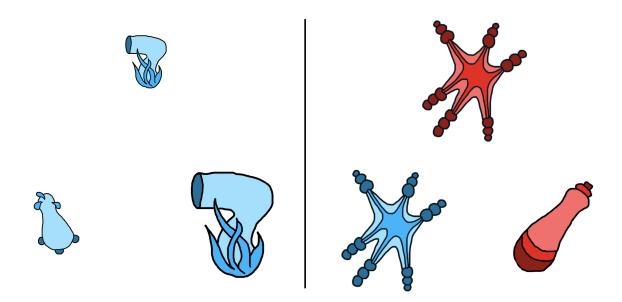


Figure 1. On the left: an example of a contrastive trial in which the critical feature is size. Here, the participant would hear the instruction "Find the [small] toma." On the right: an example of a contrastive trial in which the critical feature is color. Here, the participant would hear the instruction "Find the [red] toma." In both cases, the target is the top object.

and redrawn from Kanwisher, Woods, Iacoboni, and Mazziotta (1997); we designed the 189 remaining 15 fruit kinds. Each fruit kind had an instance in each of four colors (red, blue, 190 green, or purple) and two sizes (big or small). Particular target colors were assigned 191 randomly at each trial and particular target sizes were counterbalanced across display types. 192 There were two display types: unique target displays and contrastive displays. Unique target 193 displays contained a target object that had a unique shape and was unique on the trial's 194 critical feature (color or size), and two distractor objects that matched each other's (but not 195 the target's) shape and critical feature. These unique target displays were included as a check that participants were making reasonable referent choices and to space out contrastive 197 displays to prevent participants from dialing in on the contrastive object setup during the 198 experiment. Contrastive displays contained a target, its contrastive pair (matched the 199 target's shape but not its critical feature), and a lure (matched the target's critical feature 200 but not its shape; Fig. 1). The on-screen positions of the target and distractor items were 201

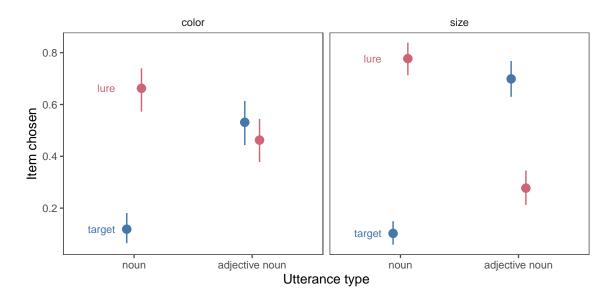


Figure 2. Proportion of times that participants chose the target and lure items as a function of condition and whether an adjective was provided. Points indicate group means; error bars indicate 95% confidence intervals computed by non-parametric bootstrapping.

randomized within a triad configuration.

Design and Procedure. Participants were told they would play a game in which they would search for strange alien fruits. Each participant saw eight trials. Half of the trials were unique target displays and half were contrastive displays. Crossed with display type, half of trials had audio instructions that described the critical feature of the target (e.g., "Find the [blue/big] toma"), and half of trials had audio instructions with no adjective description (e.g., "Find the toma"). A name was randomly chosen at each trial from a list of eight nonce names: blicket, wug, toma, gade, sprock, koba, zorp, and lomet.

After completing the study, participants were asked to select which of a set of alien words they had heard previously during the study. Four were words they had heard, and four were novel lure words. Participants were dropped from further analysis if they did not meet our pre-registered exclusion criteria of responding to at least 6 of these 8 memory check questions correctly (above chance performance as indicated by a one-tailed binomial test at the p = .05 level) and answering all four color perception check trials correctly (resulting n = .05 level)

 $163)^1$.

217 Results

We first confirmed that participants understood the task by analyzing performance on 218 unique target trials, the filler trials in which there was a target unique on both shape and the 219 relevant adjective. We asked whether participants chose the target more often than expected 220 by chance (33%) by fitting a mixed effects logistic regression with an intercept term, a 221 random effect of subject, and an offset of logit(1/3) to set chance probability to the correct 222 level. The intercept term was reliably different from zero for both color ($\beta = 6.64$, t = 4.10, 223 p < .001) and size ($\beta = 2.25$, t = 6.91, p < .001), indicating that participants consistently chose the unique object on the screen when given an instruction like "Find the (blue) toma." 225 In addition, participants were more likely to select the target when an adjective was provided in the audio instruction in both conditions. We confirmed this effect statistically by fitting a 227 mixed effects logistic regression predicting target selection from condition, adjective use, and 228 their interaction with random effects of participants. Use of description in the audio 229 increased target choice ($\beta = 3.85$, t = 3.52, p < .001), and adjective type (color vs. size) was 230 not statistically related to target choice ($\beta = -0.48$, t = -1.10, p = .269). The two effects had 231 a marginal interaction ($\beta = -2.24$, t = -1.95, p = .051). Participants had a general tendency 232 to choose the target in unique target trials, which was strengthened if the audio instruction 233 contained the relevant adjective. These effects did not significantly differ between color and 234 size adjectives, which suggests that participants did not treat color and size differently in 235 these baseline trials, though the marginal interaction suggests that use of an adjective may 236 strengthen their tendency to choose the unique object more powerfully in the size condition. 237

¹ Experiments 1 and 3 were run in 2020, during the COVID-19 pandemic, when high exclusion rates on Amazon Mechanical Turk were being reported by many experimenters. Though our pre-registered criteria led to many exclusions, the check given to participants tested memory for a few novel words heard in the experiment, which we do not believe was an overly stringent requirement.

Our key pre-registered analysis was whether participants would choose the target 238 object on contrastive trials—when they heard an adjective in the referring expression. To 239 perform this test, we compared participants' rate of choosing the target to their rate of 240 choosing the lure, which shares the relevant critical feature with the target, when they heard 241 the adjective. Overall, participants chose the target with a contrasting pair more often than 242 the unique lure, indicating that they used contrastive inferences to resolve reference (β 243 0.53, t = 3.83, p = < .001). To test whether the strength of the contrastive inference differed 244 between color and size conditions, we pre-registered a version of this regression with a term 245 for adjective type, and found that people were more likely to choose the target over the lure 246 in the size condition than the color condition ($\beta = 0.87$, t = 3.12, p = .002).

Given this result, we tested whether people consistently chose the target over the lure 248 on the color and size data separately, as a stricter check of whether the effect was present in 249 both conditions (not pre-registered). Considering color and size separately, participants 250 chose the target significantly more often than the lure in the size condition ($\beta = 0.86$, t =251 4.41, p = < .001), but not in the color condition ($\beta = 0.15$, t = 0.75, p = .455). On 252 contrastive trials in which a descriptor was not given, participants dispreferred the target, 253 instead choosing the lure object, which matched the target on the descriptor but had a 254 unique shape ($\beta = -2.65$, t = -5.44, p = < .001). Participants' choice of the target in the size 255 condition was therefore not due to a prior preference for the target in contrastive displays, 256 but relied on contrastive interpretation of the adjective. In the Supplemental Materials, we 257 report an additional pre-registered analysis of all Experiment 1 data with maximal terms 258 and random effects; those results are consistent with the more focused tests reported here. 259

Discussion

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When faced with unfamiliar objects referred to by unfamiliar words, people can use pragmatic inference to resolve referential ambiguity and learn the meanings of these new words. In Experiment 1, we found that people have a general tendency to choose objects

that are unique in shape when reference is ambiguous. However, when they hear an
utterance with description (e.g., "blue toma", "small toma"), they shift away from choosing
unique objects and toward choosing objects that have a similar contrasting counterpart.
Furthermore, use of size adjectives—but not color adjectives—prompts people to choose the
target object with a contrasting counterpart more often than the unique lure object. We
found that people are able to use contrastive inferences about size to successfully resolve
which unfamiliar object an unfamiliar word refers to.

271 Model

To formalize the inference that participants were asked to make, we developed a model 272 in the Rational Speech Act Framework (RSA, Frank & Goodman, 2012). In this framework, 273 pragmatic listeners (L) are modeled as drawing inferences about speakers' (S)274 communicative intentions in talking to a hypothetical literal listener (L_0) . This literal 275 listener makes no pragmatic inferences at all, evaluating the literal truth of a statement (e.g., 276 it is true that a red toma can be called "toma" and "red toma" but not "blue toma"), and 277 chooses randomly among all referents consistent with that statement. In planning their 278 referring expressions, speakers choose utterances that are successful at accomplishing two 279 goals: (1) making the listener as likely as possible to select the correct object, and (2) 280 minimizing their communicative cost (i.e., producing as few words as possible). Note that 281 though determiners are not given in the model's utterances, the assumption that the 282 utterance refers to a specific reference is built into the model structure, consistent with the definite determiners used in the task. Pragmatic listeners use Bayes' rule to invert the 284 speaker's utility function, essentially inferring what the speaker's intention was likely to be 285 given the utterance they produced. 286

 $Literal: P_{Lit} = \delta(u, r) P(r)$

 $Speaker: P_S(u|r) \propto \alpha \left(P_{Lit}(r|u) - C\right)$

$$Listener: P_{Learn}(r|u) \propto P_s(u|r) P(r)$$

For this experiment, we build on a Rational Speech Act model developed by Frank and Goodman (2014) to jointly resolve reference and learn new words. The primary modification of RSA is use of a pragmatic learner: a pragmatic listener who has uncertainty about the meanings of words in their language, and thus cannot directly compute the speaker's utility as written. Instead, the speaker's utility is conditioned on the set of mappings, and the learner must also infer which set of mappings is correct:

Learner:
$$P_L(r|u) \propto P_s(u|r;m) P(r) P(m)$$

In these experiments, we assume that the prior probability to refer to each object (P(r)) is equal, and similarly that all mappings (P(m)) are equally likely, so they cancel out in computations. We further assume that the cost of producing any word is identical, and so the cost of an utterance is equal to its length. All that remains is to specify the possible mappings, and literal meanings, and alternative utterances possible on each trial of the experiment. We describe the size condition here, but the computation for the color condition is analogous.

On the trial shown in the left panel of Figure 1 people see two objects that look something like a hair dryer and one that looks like a pear and they are asked to "Find the toma." Here, in the experiment design and the model, we take advantage of the fact that English speakers tend to assume that nouns generally correspond to differences in shape rather than other features (Landau, Smith, & Jones, 1992). Given this, the two possible mappings are $\{m_1 : hairdryer - "toma", pear - "?"\}$ and $\{m_2 : hairdryer - "?", pear - "toma"\}$. The literal semantics of each object allow them to be referred to by their shape label (e.g. "toma"), or by a descriptor that is true of them (e.g. "small"), but not names for other shapes or untrue descriptors.

Having heard "Find the toma," the model must now choose a referent. If the true 310 mapping for "toma" is the hair dryer (m_1) , this utterance is ambiguous to the literal listener, 311 as there are two referents consistent with the literal meaning toma. Consequently, whichever 312 of the two referents the speaker intends to point out to the learner, the speaker's utility will 313 be relatively low. Alternatively, if the true mapping for "toma" is the pear (m_2) , then the 314 utterance will be unambiguous to the literal listener, and thus the speaker's utterance will 315 have higher utility. As a result, the model can infer that the more likely mapping is m_2 and 316 choose the pear, simultaneously resolving reference and learning the meaning of "toma." 317

If instead the speaker produced "Find the small toma," the model will make a different inference. If the true mapping for "toma" is hair dryer (m_2) , this utterance now uniquely identifies one referent for the literal listener and thus has high utility. It also uniquely identifies the target if "toma" means pear (m_1) . However, if "toma" means pear, the speaker's utterance was inefficient because the single word utterance "toma" would have identified the target to the literal listener and incurred less cost. Thus, the model can infer that "toma" is more likely to mean hair dryer and choose the small hair dryer appropriately.

While these descriptions use deterministic language for clarity, the model's
computation is probabilistic and thus reflects tendencies to choose those objects rather than
fixed rules. Figure 3 shows model predictions alongside people's behavior for the size and
color contrast conditions in Experiment 1. In line with the intuition above, the model
predicts that hearing a bare noun (e.g. "toma") should lead people to infer that the intended
referent is the unique object (lure), whereas hearing a modified noun (e.g. "small toma")
should lead people to infer that the speaker's intended referent has a same-shaped
counterpart without the described feature (i.e., is the target object).

Our empirical data suggest that people treat color and size adjectives differently,
making a stronger contrastive inference with size than with color. One potential explanation
for this difference is that people are aware of production asymmetries between color and size.

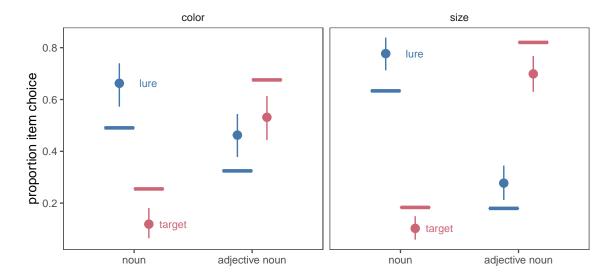


Figure 3. Proportion of times that people (and our model) chose the target and lure items as a function of adjective type and whether an adjective was provided. Points indicate empirical means; error bars indicate 95% confidence intervals computed by non-parametric bootstrapping. Solid horizontal lines indicate model predictions.

As mentioned, speakers tend to over-describe color, providing more color adjectives than
necessary to establish reference, while describing size more minimally (Nadig & Sedivy, 2002;
Pechmann, 1989). Listeners may be aware of this production asymmetry and discount the
contrastive weight of color adjectives with respect to reference.

In the Rational Speech Act model, this kind of difference is captured neatly by a
difference in the listener's beliefs about the speaker's rationality (i.e. how sensitive the
speaker is to differences in utility of different utterances). We estimated the rationality
parameter separately for color and size, reflecting that listeners may believe speakers are
more attentive to differences in utility for some feature descriptions than others. (Note that
the rationality parameter is sometimes used to explain individual differences in speaker
rationality, and estimated on a person level; that is not how we are using it here.) To
determine the value of the rationality parameter that best describes participants' behavior in
each condition, we used Bayesian data analysis, estimating the posterior probability of the

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observed data under each possible value of α multiplied by the prior probability of each of those values. In each condition, α was drawn from a Gamma distribution with shape and scale parameters set to 2 (Gamma(2,2)). This prior encodes a weak preference for small values of α , but the estimates below were not sensitive to other choices of hyper-parameters.

Posterior mean estimates of rationality varied substantially across conditions. In the color condition, the rationality parameter was estimated to be 2.00 with a 95% credible interval of [1.37, 2.63]. In the size condition, rationality was estimated to be 3.98 [3.22, 4.74].

Figure 3 shows the model predictions along with the empirical data from Experiment 1. 356 The model broadly captures the contrastive inference—when speakers produce an adjective 357 noun combination like "red toma," the model selects the target object more often than the 358 lure object. The extent to which the model makes this inference varies as predicted between 350 the color and size adjective conditions in line with the different estimated rationality values. 360 In both conditions, despite estimating the value of rationality that makes the observed data 361 most probable, the model overpredicts the extent of the contrastive inference that people 362 make. Intuitively, it appears that over and above the strength of their contrastive inferences, 363 people have an especially strong tendency to choose a unique object when they hear an unmodified noun (e.g. "toma"). In an attempt to capture this uniqueness tendency, the model overpredicts the extent of the contrastive inference.

The model captures the difference between color and size in a difference in the
rationality parameter, but leaves open the ultimate source of this difference in rationality.
Why do people make stronger pragmatic inferences about size than color when determining
reference? Our model implements this difference in a relatively agnostic way (in a
feature-specific rationality parameter), and our results cannot arbitrate between particular
explanations, but we spell out a few possibilities and modeling alternatives here.

One way to capture this asymmetry would be to locate it in a different part of the

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model: in the literal semantics of color and size. A recent model from Degen, Hawkins, Graf, Kreiss, and Goodman (2020) does predict a color-size asymmetry based on different semantic 375 exactness. In this model, literal semantics are treated as continuous rather than discrete, so 376 "blue" is neither 100% true nor 100% false of a particular object, but can instead be 90%377 true. They successfully model a number of color-size asymmetries in production data by 378 treating color as having stronger literal semantics (e.g. "blue toma" is a better description of 379 a small blue toma than "small toma" is). However, implementing semantic inexactness alone 380 in our model predicts the opposite asymmetry of what we found. Because color has stronger 381 semantics than size, the listener in this model shows a stronger contrast effect for color than 382 size (see demonstration in the Supplemental Materials). Thus, though a continuous 383 semantics can explain our asymmetry, this explanation is unlikely given that the continuous 384 semantics that predicts other empirical color-size asymmetries does not predict our findings.

Another possibility is that people attend to the production probabilities of different adjective types and attenuate their inferences accordingly. As discussed, speakers mention color more often than size, and often mention color when it is not necessary for reference. If listeners keep track of these probabilities, they can discount the weight of color description in identifying referents. Kreiss and Degen (2020) show just that in a task with familiar objects: people discount the weight of often-produced adjectives and

[talk about production norms predicted by inexactness] [then talk about gradableness separate from production norms]

Another difference between size and color adjectives is that size adjectives are gradable:
their meaning is often judged relative to a comparison class (e.g., "He is a tall basketball
player" may have a meaning akin to "He is tall for a basketball player") (Kennedy, 2007).
Because this comparison class is sensitive to context (it can even change within a sentence,
e.g., "He is tall, but not tall for a basketball player"), there is active disagreement about
whether this aspect of gradable adjective meaning is properly considered semantics or

pragmatics (Xiang, Kennedy, Xu, & Leffel, 2022). Thus, a possible explanation is that the 400 presence of a comparison class is necessary to judge size, and this comparison class may be 401 attributed to either pragmatics or literal semantics. That is, in a trial such as the one on the 402 left in 1, a participant sees two hairdryer-shaped objects (one big and one small) and one 403 small pear-shaped object. When they hear "Find the small toma," they choose the only 404 object that is small and has a potential comparison class: the small hairdryer, which has a 405 larger hairdryer counterpart. On the other hand, color adjectives are not relative gradable 406 adjectives, and so a comparison class is not necessary to interpret them: they have more 407 absolute meaning. Thus, it is possible to explain the color-size asymmetry by the necessity 408 of a comparison class for judging size, which may be considered either an element of 409 semantics or pragmatics. 410

Yet another way to explain the difference between size and color adjectives is to 411 attribute size adjectives' contrastive strength with respect to reference to the fact that size 412 adjectives are gradable and relative. There are multiple ways to implement this possibility in 413 the model. One way would be to specify that speakers tend to remark on relative, gradable 414 features when making distinctions among present objects because direct comparisons for the 415 meaning of 'small' and 'big' are at hand, whereas color adjectives are more often mentioned 416 superfluously because they have more absolute meaning and do not need immediate 417 comparisons. This possibility is consistent with the model we have specified, and is just one 418 possible reason for a production asymmetry which listeners are responding to rationally in 419 their inferences. 420

Another possibility is that the gradable, relative nature of size adjectives should be encoded in the pragmatic learner part of the model: a learner might need a comparison point to tell whether a novel object is small or big, but not red or purple, and thus avoid choosing a unique (shaped) object when size is specified but be willing to choose a unique object when color is specified. This possibility would require more fundamental changes to the model.

Here, we make the conservative choice to encode the color-size asymmetry in the broad rationality parameter, though changing the pragmatic learner's decision process is an intriguing possibility for future work.

Overall, we found that people can use contrastive inferences from description to map 429 an unknown word to an unknown object. This inference is captured by an extension of the 430 Rational Speech Act model using a pragmatic learner, who is simultaneously making 431 inferences over possible referents and possible lexicons. This model can also capture people's 432 tendency to make stronger contrastive inferences from color description than size description 433 through differences in the rationality parameter, though the origin of these differences cannot 434 be pinned down with this experiment alone. Our experiment and model results suggest that 435 people can resolve a request like "Give me the small dax" by reasoning that the speaker must 436 have been making a useful distinction by mentioning size, and therefore looking for multiple 437 similar objects that differ in size and choosing the smaller one. Immediately available objects 438 are not the only ones worth making a distinction from, however. Next, we turn to another salient set of objects a speaker might want to set a referent apart from: the referent's category.

Experiment 2

When referring to a big red dog or a hot-air balloon, we often take care to describe
them—even when there are no other dogs or balloons around. Speakers use more description
when referring to objects with atypical features (e.g., a yellow tomato) than typical ones
(e.g., a red tomato; Mitchell et al., 2013; Bergey, Morris, & Yurovsky, 2020; Rubio-Fernández,
2016; Westerbeek et al., 2015). This selective marking of atypical objects potentially supplies
useful information to listeners: they have the opportunity to not only learn about the object
at hand, but also about its broader category. Horowitz and Frank (2016) demonstrated that,
combined with other contrastive cues (e.g., "Wow, this one is a zib. This one is a TALL zib"),
prenominal adjectives prompted adults and children to infer that the described referent was

less typical than one that differed on the mentioned feature (e.g., a shorter zib). This work 452 provided a useful demonstration that adjective use can contribute to inferences about feature 453 typicality, though it did not isolate the effect of adjectives specifically. Their experiments 454 used several contrastive cues, such as prosody (contrastive stress on the adjective: "TALL 455 zib"), demonstrative phrases that may have marked the object as unique ("this one") and 456 expressions of surprise at the object ("wow"), and participants may have inferred the object 457 was atypical primarily from these cues and not from the adjective. Thus, in this experiment. 458 we first set out to ask whether adjective use alone prompts an inference of atypicality: when 459 you hear "purple toma," do you infer that fewer tomas in general are purple? 460

We will also test how this inference differs (or does not) between size and color 461 adjectives. The fact that people use adjectives to draw a contrast between an object and its 462 category may help make sense of the asymmetry between color and size adjectives we found 463 in Experiment 1. Color adjectives that are redundant with respect to reference are not 464 necessarily redundant in general. Rubio-Fernández (2016) demonstrates that speakers often 465 use 'redundant' color adjectives to describe colors when they are variable and central to the 466 category's meaning (e.g., colorful t-shirts) or when they are atypical (e.g., a purple banana). 467 Comprehenders, in turn, expect color adjectives to be used informatively with respect to 468 typicality, and upon hearing color adjectives tend to look to referents for which the adjective 460 describes a less-typical feature (e.g., "Choose the yellow..." prompts people to look to a 470 vellow shirt over a vellow banana; Rohde & Rubio-Fernandez, 2021; Kreiss & Degen, 2020). 471 Therefore, while size may hold more contrastive weight with respect to reference, color and 472 size may hold similar contrastive weight with respect to the category's feature distribution. 473 In Experiment 2, we test whether listeners use descriptive contrast with a novel object's 474 category to learn about the category's feature distribution.

If listeners do make contrastive inferences about typicality, it may not be as simple as judging that an over-described referent is atypical. Description can serve many purposes: in

Experiment 1, we investigated its use in contrasting between present objects. If a descriptor was needed to distinguish between two present objects, it may not have been used to mark atypicality. For instance, in the context of a bin of heirloom tomatoes, a speaker who wanted a red one in particular might specify that they want a "red tomato" rather than just asking for a "tomato." In this case, the adjective "red" is being used contrastively with respect to reference (as in Experiment 1), and not to mark atypicality. Thus, a listener who does not know much about tomatoes may attribute the use of "red" to referential disambiguation given the context and not infer that red is an unusual color for tomatoes.

In Experiment 2, we used an artificial language task to set up just this kind of learning 486 situation. We manipulated the contexts in which listeners hear adjectives modifying novel 487 names of novel referents. These contexts varied in how useful the adjective was to identify 488 the referent: in one context the adjective was necessary, in another it was helpful, and in a 489 third it was entirely redundant. On a reference-first view, use of an adjective that was 490 necessary for reference can be explained away and should not prompt further inferences 491 about typicality—an atypicality inference would be blocked. If, on the other hand, people 492 take into account speakers' multiple reasons for using adjectives without giving priority to 493 reference, they may alter their inferences about typicality across these contexts in a graded 494 way: if an adjective was necessary for reference, it may prompt slightly weaker inferences of 495 atypicality; if an adjective was redundant with respect to reference, it may be inferred to 496 mark atypicality more strongly. Further, these contexts may also prompt distinct inferences 497 when no adjective is used: for instance, when an adjective is necessary to identify the referent but elided, people may infer that the elided feature is particularly typical. To account for the multiple ways context effects might emerge, we analyze both of these 500 possibilities. Overall, we asked whether listeners infer that these adjectives identify atypical 501 features of the named objects, and whether the strength of this inference depends on the 502 referential ambiguity of the context in which adjectives are used. 503

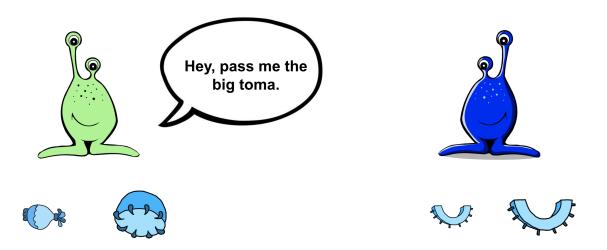


Figure 4. Experiment 2 stimuli. In the above example, the critical feature is size and the object context is a within-category contrast: the alien on the right has two same-shaped objects that differ in size.

$_{504}$ Method

Participants. 240 participants were recruited from Amazon Mechanical Turk. Half of the participants were assigned to a condition in which the critical feature was color (red, blue, purple, or green), and the other half of participants were assigned to a condition in which the critical feature was size (small or big).

Stimuli & Procedure. Stimulus displays showed two alien interlocutors, one on the left side (Alien A) and one on the right side (Alien B) of the screen, each with two novel fruit objects beneath them (Figure 4). Alien A, in a speech bubble, asked Alien B for one of its fruits (e.g., "Hey, pass me the big toma"). Alien B replied, "Here you go!" and the referent disappeared from Alien B's side and reappeared on Alien A's side. Note that the participants do not make a referent choice in this experiment; the measure of interest is their typicality judgments of the objects' features, described below.

We manipulated the critical feature type (color or size) between subjects. Two factors (presence of the critical adjective in the referring expression and object context) were fully

crossed within subjects. Object context had three levels: within-category contrast, 518 between-category contrast, and same feature (Figure 5). In the within-category contrast 519 condition, Alien B possessed the target object and another object of the same shape, but 520 with a different value of the critical feature (e.g., a big toma and a small toma). In the 521 between-category contrast condition, Alien B possessed the target object and another object 522 of a different shape, and with a different value of the critical feature (e.g., a big toma and a 523 small blicket). In the same feature condition, Alien B possessed the target object and 524 another object of a different shape but with the same value of the critical feature as the 525 target (e.g., a big toma and a big dax). Thus, in the within-category contrast condition, the 526 descriptor was necessary to distinguish the referent; in the between-category contrast 527 condition it was unnecessary but potentially helpful; and in the same feature condition it 528 was unnecessary and unhelpful.

Note that in all context conditions, the set of objects on screen was the same in terms 530 of the experiment design: there was a target (e.g., big toma), an object with the same shape 531 as the target and a different critical feature (e.g., small toma), an object with a different 532 shape from the target and the same critical feature (e.g., big dax), and an object with a 533 different shape from the target and a different critical feature (e.g., small blicket). Context 534 was manipulated by rearranging these objects such that the relevant referents (the objects 535 under Alien B) differed and the remaining objects were under Alien A. Thus, in each case, 536 participants saw the target object and one other object that shared the target object's shape 537 but not its critical feature—they observed the same kind of feature distribution of the target 538 object's category in each trial type. The particular values of the features were chosen 539 randomly for each trial. 540

Participants completed six trials. After each exchange between the alien interlocutors, they made a judgment about the prevalence of the target's critical feature in the target object's category. For instance, after seeing a red blicket being exchanged, participants would be asked, "On this planet, what percentage of blickets do you think are red?" They
would answer on a sliding scale between zero and 100. In the size condition, participants
were asked, "On this planet, what percentage of blickets do you think are the size shown
below?" with an image of the target object they just saw available on the screen.

After completing the study, participants were asked to select which of a set of alien words they had seen previously during the study. Four were words they had seen, and four were novel lure words. Participants were dropped from further analysis if they did not respond to at least 6 of these 8 correctly (above chance performance as indicated by a one-tailed binomial test at the p = .05 level). This resulted in excluding 47 participants, leaving 193 for further analysis.

Results

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Our key test is whether participants infer that a mentioned feature is less typical than one that is not mentioned. In addition, we tested whether inferences of atypicality are modulated by context. One way to test this is to analyze the interaction between utterance type and context, seeing if the difference between adjective and no adjective utterances is larger when the adjective was highly redundant or smaller when the adjective was necessary for reference.

We analyzed participants' judgments of the prevalence of the target object's critical feature in its category. We began by fitting a maximum mixed-effects linear model with effects of utterance type (adjective or no adjective), context type (within category, between category, or same feature, with between category as the reference level), and critical feature (color or size) as well as all interactions and random slopes of utterance type and context type nested within subject. Random effects were removed until the model converged. The final model included the effects of utterance type, context type, and critical feature and their interactions, and a random slope of utterance type by subject. This model revealed a

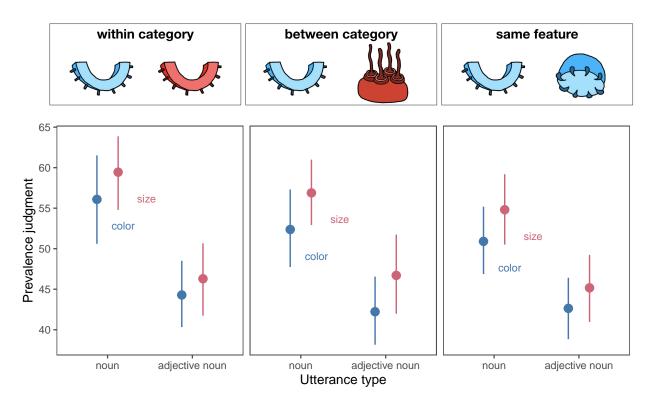


Figure 5. Prevalence judgments from Experiment 2. Participants consistently judged the target object as less typical of its category when the referent was described with an adjective (e.g., "Pass me the blue toma") than when it was not (e.g., "Pass me the toma"). This inference was not significantly modulated by object context (examples shown above each figure panel).

significant effect of utterance type ($\beta_{adjective} = -10.22$, t = -3.37, p = .001), such that 569 prevalence judgments were lower when an adjective was used than when it was not. 570 Participants' inferences did not significantly differ between color and size adjective conditions 571 $(\beta_{size} = 4.73, t = 1.46, p = .146)$. Participants' inferences did not significantly vary by context type ($\beta_{within} = 3.92, t = 1.63, p = .104; \beta_{same} = -1.48, t = -0.62, p = .537$). There was not a significant interaction between context and presence of an adjective in the 574 utterance ($\beta_{within*adjective} = -1.58$, t = -0.46, p = .644; $\beta_{same*adjective} = 2.13$, t = 0.63, p = .646575 .532). That is, participants did not significantly adjust their inferences based on object 576 context, nor did they make differential inferences based on the combination of context and 577

adjective use. However, they robustly inferred that mentioned features were less prevalent in
the target's category than unmentioned features.

This lack of a context effect may be because people do not take context into account. 580 or because they make distinct inferences when an adjective is not used: for instance, when 581 an adjective is necessary for reference but elided, people may infer that the unmentioned 582 feature is very typical. This inference would lead to a difference between the adjective and 583 no adjective utterances in the within-category context, but not because people are failing to 584 attribute the adjective to reference. To account for this possibility, we separately tested 585 whether there are effects of context among just the trials with adjectives and just the trials 586 without adjectives. In each case, we fit a model with effects of context type and critical 587 feature as well as their interaction and random slopes by subject. Participants did not significantly adjust their inferences by context among only the noun utterances (β_{within} = $3.94, t = 1.47, p = .143; \beta_{same} = -1.46, t = -0.54, p = .587$). That is, we did not find 590 evidence here that people were inferring a feature to be highly typical because it went 591 unmentioned when it was necessary for reference. Participants also did not significantly 592 adjust their inferences by context among only the adjective noun utterances ($\beta_{within} = 2.43$, 593 $t = 1.16, p = .247; \beta_{same} = 0.67, t = 0.32, p = .750$). That is, we did not find evidence that 594 people modulated their typicality inferences based on the referential context among trials 595 where this inference could not have been driven by omission either. Overall, we did not find 596 evidence that participants significantly adjusted their inferences based on context. 597

598 Discussion

Description is often used not to distinguish among present objects, but to pick out an object's feature as atypical of its category. In Experiment 2, we asked whether people would infer that a described feature is atypical of a novel category after hearing it mentioned in an exchange. We found that people robustly inferred that a mentioned feature was atypical of its category, across both size and color description. Further, participants did not use object

context to substantially explain away description. That is, even when description was
necessary to distinguish among present objects (e.g., there were two same-shaped objects
that differed only in the mentioned feature), participants still inferred that the feature was
atypical of its category. This suggests that, in the case of hearing someone ask for a "red
tomato" from a bin of many-colored heirloom tomatoes, a person naive about tomatoes
would infer that tomatoes are relatively unlikely to be red.

Unlike Experiment 1, in which people made stronger contrastive inferences for size 610 than color, there were not substantial differences between people's inferences about color and 611 size in Experiment 2. If an account based on production norms is correct, this suggests that 612 people track both how often people use color compared to size description and also for what 613 purpose-contrasting with present objects or with the referent's category. That is, color description may be more likely to be used superfluously with respect to present objects but informatively with respect to the category. Indeed, color description that seems 616 overdescriptive with respect to object context often occurs when the category has many-colored members (e.g., t-shirts) or when the object's color is atypical 618 (Rubio-Fernández, 2016). However, our results are consistent with several potential 619 explanations of the color-size asymmetry (or lack thereof). Future work addressing the 620 source of the color-size asymmetry will need to explain differences in its extent when 621 distinguishing among present objects compared to the referent's category. 622

623 Model

To allow the Rational Speech Act Framework to capture inferences about typicality, we modified the Speaker's utility function to have an additional term: the listener's expected processing difficulty. Speakers may be motivated to help listeners to select the correct referent not just eventually but as quickly as possible. People are both slower and less accurate at identifying atypical members of a category as members of that category (Dale, Kehoe, & Spivey, 2007; Rosch, Simpson, & Miller, 1976). If speakers account for listeners'

processing difficulties, they should be unlikely to produce bare nouns to refer to low typicality exemplars (e.g. unlikely to call a purple carrot "carrot"). This is roughly the kind of inference encoded in Degen et al. (2020)'s continuous semantics Rational Speech Act model.

We model the speaker as reasoning about the listener's label verification process. 633 Because the speed of verification scales with the typicality of a referent, a natural way of 634 modeling it is as a process of searching for that particular referent in the set of all exemplars 635 of the named category, or alternatively of sampling that particular referent from the set of 636 all exemplars in that category, P(r|Cat). On this account, speakers want to provide a 637 modifying adjective for atypical referents because the probability of sampling them from 638 their category is low, but the probability of sampling them from the modified category is 639 much higher (a generalization of the size principle (Xu & Tenenbaum, 2007)). Typicality is 640 just one term in the speaker's utility, and thus is directly weighed with the literal listener's 641 judgment and against cost.

If speakers use this utility function, a listener who does not know the feature
distribution for a category can use a speaker's utterance to infer it. Intuitively, a speaker
should prefer not to modify nouns with adjectives because they incur a cost for producing an
extra word. If they did use an adjective, it must be because they thought the learner would
have a difficult time finding the referent from a bare noun alone because of typicality,
competing referents, or both. To infer the true prevalence of the target feature in the
category, learners combine the speaker's utterance with their prior beliefs about the feature
distribution.

We model the learner's prior about the prevalence of features in any category as a Beta distribution with two parameters α and β that encode the number of hypothesized prior psuedo-exemplars with the feature and without feature that the learner has previously observed (e.g., one red dax and one blue dax or one big dax and one small dax). We assume that the learner believes they have previously observed one hypothetical psuedo-examplar of

each type, which is a weak symmetric prior indicating that the learner expects features to 656 occur in half of all members of a category on average, but would find many levels of 657 prevalence unsurprising. To model the learner's direct experience with the category, we add 658 the observed instances in the experiment to these hypothesized prior instances. After 659 observing one member of the target category with the relevant feature and one without, the 660 listener's prior is thus updated to be Beta (2, 2). Thus, we model learners as believing the 661 feature prevalence is roughly 50% based on their initial priors and direct observation in the 662 trial; they then combine this knowledge of the feature distribution with their pragmatic inference about the utterance to arrive at a final prevalence judgment.

As in Experiment 1, we encoded potential differences between people's inferences about 665 color and size in feature rationality parameters, which we estimated separately for 666 Experiment 2. In contrast to Experiment 1, the absolute values of these parameters are 667 driven largely by the number of pseudo-exemplars assumed by the listener prior to exposure. 668 Thus, the feature rationality parameters inferred in the two experiments are not directly 669 comparable. However, differences between color and size within each model are interpretable. 670 As in Experiment 1, we found that listeners inferred speakers to be more rational when using 671 size adjectives (0.89 [0.63, 1.13]) than color adjectives (0.60 [0.37, 0.83]), but the two inferred 672 confidence intervals were overlapping, suggesting that people treated size and color adjectives 673 similarly when making inferences about typicality. 674

Figure 6 shows the predictions of our Rational Speech Act model compared to
empirical data from participants. The model captures the trends in the data correctly,
inferring that the critical feature was less prevalent in the category when it was mentioned
(e.g., "red dax") than when it was not mentioned (e.g., "dax"). The model also infers the
prevalence of the critical feature to be numerically higher in the within-category condition,
like people do. That is, in the within-category condition when an adjective is used to
distinguish between referents, the model thinks that the target color is slightly less atypical.

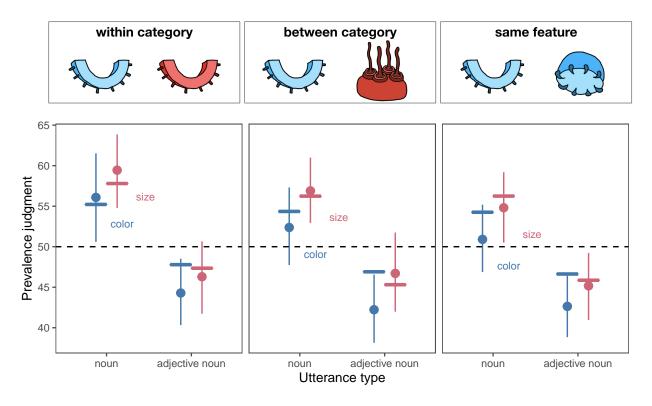


Figure 6. Participants' prevalence judgments from Experiment 2, compared to model predictions (horizontal lines).

When an adjective would be useful to distinguish between two objects of the same shape but one is not used, the model infers that the color of the target object is slightly more typical.

Overall, our model captures the inference people make: when the speaker mentions a feature (e.g., "the blue dax"), that feature is inferred to be less typical of the category (daxes are less likely to be blue in general). It further captures that when the object context requires an adjective for successful reference, people weaken this atypicality inference only slightly, if at all. In contrast to a reference-first view, which predicts that these two kinds of inferences would trade off strongly—that is, using an adjective that is necessary for reference blocks the inference that it is marking atypicality—the model captures the graded way in which people consider these two communicative goals.

Experiment 3

In Experiments 1 and 2, we established that people can use contrastive inferences to 693 resolve referential ambiguity and to make inferences about the feature distribution of a novel 694 category. Additionally, in Experiment 2, we found that these two inferences do not seem to 695 trade off substantially: even if an adjective is necessary to establish reference, people infer 696 that it also marks atypicality. We also found that inferences of atypicality about color and 697 size adjectives pattern very similarly, though their baseline typicality is shifted, while color 698 and size are not equally contrastive with respect to referential disambiguation (Experiment 699 1). 700

To strengthen our findings in a way that would allow us to better detect potential 701 trade-offs between these two types of inference, in Experiment 3 we conducted a 702 pre-registered replication of Experiment 2 with a larger sample of participants. In addition, 703 we tested how people's prevalence judgments from utterances with and without an adjective 704 compare to their null inference about feature prevalence by adding a control utterance 705 condition: an alien utterance, which the participants could not understand. This also tests 706 the model assumption we made in Experiment 2: that after seeing two exemplars of the 707 target object with two values of the feature (e.g., one green and one blue), people's 708 prevalence judgments would be around 50%. In addition to validating this model 700 assumption, we more strongly tested the model here by comparing predictions from same 710 model, with parameters inferred from the Experiment 2 data, to data from Experiment 3. 711 Our pre-registration of the method, recruitment plan, exclusion criteria, and analyses can be 712 found on the Open Science Framework: https://osf.io/s8gre (note that this experiment is 713 labeled Experiment 2 in the OSF repository but is Experiment 3 in the paper). 714

Method

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Participants. A pre-registered sample of 400 participants was recruited from
Amazon Mechanical Turk. Half of the participants were assigned to a condition in which the

critical feature was color (red, blue, purple, or green), and half of the participants were assigned to a condition in which the critical feature was size (small or big).

Stimuli & Procedure. The stimuli and procedure were identical to those of 720 Experiment 2, with the following modifications. Two factors, utterance type and object 721 context, were fully crossed within subjects. Object context had two levels: within-category 722 contrast and between-category contrast. In the within-category context condition, Alien B 723 possessed the target object and another object of the same shape, but with a different value of the critical feature (color or size). In the between-category contrast condition, Alien B possessed the target object and another object of a different shape, and with a different value 726 of the critical feature. Thus, in the within-category contrast condition, an adjective is necessary to distinguish the referent; in the between-category contrast condition it is 728 unnecessary but potentially helpful. There were three utterance types: adjective, no 729 adjective, and alien utterance. In the two alien utterance trials, the aliens spoke using 730 completely unfamiliar utterances (e.g., "Zem, noba bi yix blicket"). Participants were told in 731 the task instructions that sometimes the aliens would talk in a completely alien language, 732 and sometimes their language will be partly translated into English. To keep participants 733 from making inferences about the content of the alien utterances using the utterance content 734 of other trials, both alien language trials were first; other than this constraint, trial order was 735 random. We manipulated the critical feature type (color or size) between subjects. 736

After completing the study, participants were asked to select which of a set of alien words they had seen previously during the study. Four were words they had seen, and four were novel lure words. Participants were dropped from further analysis if they did not meet our pre-registered criteria of responding to at least 6 of these 8 correctly (above chance performance as indicated by a one-tailed binomial test at the p = .05 level) and answering all four color perception check questions correctly. Additionally, six participants were excluded because their trial conditions were not balanced due to an error in the run of the experiment. This resulted in excluding 203 participants, leaving 197 for further analysis. In our

pre-registration, we noted that we anticipated high exclusion rates, estimating that approximately 150 people per condition would be sufficient to test our hypotheses.

747 Results

We began by fitting a pre-registered maximum mixed-effects linear model with effects 748 of utterance type (alien utterance, adjective, or no adjective; alien utterance as reference level), context type (within category or between category), and critical feature (color or size) 750 as well as all interactions and random slopes of utterance type and context type nested 751 within subject. Random effects were removed until the model converged, which resulted in a 752 model with all fixed effects, all interactions and a random slope of utterance type by subject. 753 The final model revealed a significant effect of the no adjective utterance type compared to 754 the alien utterance type ($\beta = 7.48$, t = 2.80, p = .005) and no significant effect of the 755 adjective utterance type compared to the alien utterance type ($\beta = -0.64$, t = -0.24, p =756 .808). The effects of context type (within-category or between-category) and adjective type 757 (color or size) were not significant ($\beta_{within} = -2.70$, $t_{within} = -1.23$, $p_{within} = .220$; $\beta_{size} = 4.44$, 758 $t_{size} = 1.33, p_{size} = .185$). There were marginal interactions between the adjective utterance 759 type and the size condition ($\beta = -6.56$, t = -1.72, p = .086), the adjective utterance type and 760 the within-category context ($\beta = 5.77$, t = 1.86, p = .064), and the no adjective utterance 761 type and the within-category context ($\beta = 5.57$, t = 1.79, p = .073). No other effects were 762 significant or marginally significant. Thus, participants inferred that an object referred to in 763 an intelligible utterance with no description was more typical of its category on the target 764 feature than an object referred to with an alien utterance. Participants did not substantially adjust their inferences based on the object context. The marginal interactions between the within-category context and both the adjective and no adjective utterance types suggest that 767 people might have judged the target feature as slightly more prevalent in the within-category 768 context when intelligible utterances (with a bare noun or with an adjective) were used 769 compared to the alien utterance. If people are discounting their atypicality inferences when 770

the adjective is necessary for reference, we should expect them to have slightly higher
typicality judgments in the within-category context when an adjective is used, and this
marginal interaction suggests that this may be the case. However, since typicality judgments
in the no adjective utterance type are also marginally greater in the within-category context,
and because judgments in the alien utterance conditions (the reference category) also
directionally move between the two context conditions, it is hard to interpret whether this
interaction supports the idea that people are discounting their typicality judgments based on
context.

Given that interpretation of these results with respect to the alien utterance condition can be difficult, we pre-registered a version of the same full model excluding alien utterance 780 trials with the no adjective utterance type as the reference level. This model revealed a 781 significant effect of utterance type: participants' prevalence judgments were lower when an 782 adjective was used than when it was not ($\beta = -8.12$, t = -3.46, p = .001). No other effects 783 were significant. This replicates the main effect of interest in Experiment 2: when an 784 adjective is used in referring to the object, participants infer that the described feature is less 785 typical of that object's category than when the feature goes unmentioned. It also shows that 786 the possibility that people may discount their typicality judgments based on context 787 (suggested by the marginal interaction described above) is not supported when we compare 788 the adjective and no adjective utterance types directly. In the Supplemental Materials, we 780 report two more pre-registered tests of the effect of utterance type alone on prevalence 790 judgments whose results are consistent with the fuller models reported here. 791

As in Experiment 2, our test of whether participants' inferences are modulated by
context is potentially complicated by people making distinct inferences when an adjective is
necessary but *not* used. Thus, we additionally tested whether participants' inferences varied
by context among only utterances with an adjective by fitting a model with effects of context
and adjective type and their interaction, as well as random slopes by subject (not

pre-registered). Participants' inferences did not significantly differ by context ($\beta_{within} = 3.07$, $t_{within} = 1.70$, $p_{within} = .091$). Thus, participants' inferences did not significantly differ between contexts, whether tested by the interaction between utterance type and contexts or by the effect of context among only utterances with an adjective.

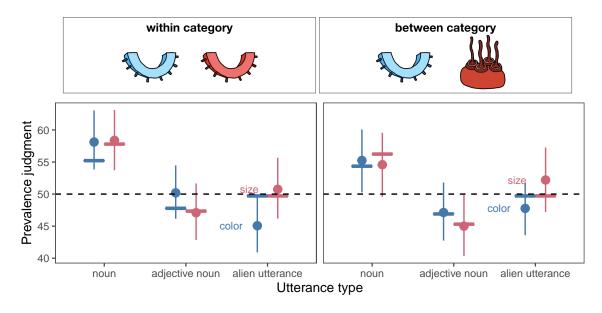


Figure 7. Participants' prevalence judgments in Experiment 3, with model predictions using the parameters estimated in Experiment 2 (horizontal lines).

Model

To validate the model we developed for Experiment 2, we compared its estimates using 802 the previously fit parameters to the new data for Experiment 3. As shown in Figure 7, the 803 model predictions were well aligned with people's prevalence judgments. In addition, in 804 Experiment 2, we fixed the model's prior beliefs about the prevalence of the target object's 805 color or size to be centered at 50% because the model had seen one pseudo-exemplar of the target color/size, and one psuedo-exemplar of the non-target color/size. In Experiment 3, we aimed to estimate this prior empirically in the alien utterance condition, reasoning that 808 people could only use their prior to make a prevalence judgment (as we asked the model to 809 do). In both the color and size conditions, people's judgments indeed varied around 50%, 810 although in the color condition they were directionally lower. This small effect may arise 811

from the fact that size varies on a scale with fewer nameable points (e.g., objects can be big,
medium-sized or small) whereas color has many nameable alternatives (e.g., red, blue, green,
etc.). Thus, the results of Experiment 3 confirm the modeling assumptions we made in
estimating people's prior beliefs, and further validate the model we developed as a good
candidate model for how people simultaneously draw inferences about speakers' intended
referents and the typicality of these referents. That is, when people think about why a
speaker chose their referring expression, they consider the context of not only present objects,
but also the broader category to which the referent belongs.

820 Discussion

In Experiment 3, we replicated the main finding of interest in Experiment 2: when a 821 novel object's feature is described, people infer that the feature is rarer of its category than 822 when it goes unmentioned. Again, this effect was consistent across both size and color 823 adjectives, and people did not substantially adjust this inference based on how necessary the 824 description was to distinguish among potential referents. We also added an alien language 825 condition, in which the entire referring expression was unintelligible to participants, to probe 826 people's priors on feature typicality. We found that in the alien language condition, people 827 judged features to be roughly between the adjective utterance and no adjective utterance 828 conditions, and significantly different from the no adjective utterance condition. In the alien language condition, people's prevalence judgments were roughly around our model's 830 prevalence judgments (50%) after observing the objects on each trial and before any 83: inferences about the utterance.

The similarity of people's prevalence judgments in the alien language condition and the adjective condition raises the question: is this effect driven by an atypicality inference in the adjective conditions, or a *typicality* inference when the feature is unmentioned? Our results suggest that it is a bit of both. When someone mentions an object without extra description, the listener can infer that its features are likely more typical than their prior; when they use

description, they can infer that its features are likely less typical. Because using an extra
word—an adjective—is generally not thought of as the default way to refer to something, this
effect is still best described as a contrastive inference of atypicality when people use
description. However, the fact that people infer high typicality when an object is referred to
without description suggests that, in some sense, there is no neutral way to refer: people will
make broader inferences about a category from even simple mentions of an object.

General Discussion

When we think about what someone is trying to communicate to us, we go far beyond 845 the literal meanings of the words they say: we make pragmatic inferences about why they 846 chose those particular words rather than other words they could have used instead. In most 847 work on pragmatic reasoning, speakers and listeners share the same knowledge of language, 848 and the question of interest is whether listeners can use their knowledge of language to learn 849 something about the unknown state of the world. Here we focus on an even more challenging 850 problem: Can pragmatic inference be used to learn about language and the world 851 simultaneously? 852

In three studies we showed that people can use pragmatic inference to (1) learn the 853 meaning of a novel word, (2) learn the typical features of the category described by this 854 novel word, and (3) rationally integrate these two kinds of reasoning processes. In 855 Experiment 1, we show that people can use descriptive contrast implied by adjectives like 856 "big" or "blue" to resolve referential ambiguity to learn a new word; in the case of color, they 857 shift substantially in the direction of the correct mapping, and in the case of size, they choose the correct mapping significantly more often than the incorrect one. In Experiments 2 and 3, we show that people infer that a noted feature is atypical of the object being referred to. Critically, people infer that the described feature is atypical even when the descriptor is 861 helpful for referential disambiguation—although the size of the atypicality inference is 862 numerically reduced. 863

Why do people think that the mentioned feature is atypical even when its mention is 864 helpful for referential disambiguation? If people use language for multiple goals—for example, 865 both for reference and for description—then listeners should reason jointly about all of the 866 possible reasons why speakers could have used a word. To determine what rational listeners 867 would do in this circumstance, we developed an extension of the Rational Speech Act 868 Framework that reasons both about reference and about the typical features of categories to 860 which objects belong. The behavior of this model was closely aligned to the behavior we 870 observed in people. Because rational inference is probabilistic rather than deterministic, 871 descriptors still lead to atypicality inferences even when they are helpful for referential 872 disambiguation. This work thus adds to the growing body of work extending the Rational 873 Speech Act framework from reasoning about just reference to reasoning about other goals as 874 well, such as inferring that speech is hyperbolic, inferring when speakers are being polite rather than truthful, and learning new words in ambiguous contexts (Frank & Goodman, 876 2014; Goodman & Frank, 2016; Kao, Wu, Bergen, & Goodman, 2014; Yoon, Tessler, Goodman, & Frank, 2020).

In considering how people may integrate inferences about typicality and about reference, we raised two broad possibilities: (1) a reference-first view, whereby if an adjective was necessary for reference people would [XXXXXXXX]

Though the participants in our experiments were adults, the ability to disambiguate
novel referents using contrast most obviously serves budding language learners—children.
Contrastive use of adjectives is a pragmatic regularity in language that children could
potentially exploit to establish word—referent mappings. Use of adjectives has been shown to
allow children to make contrastive inferences among familiar present objects (Davies,
Lingwood, Ivanova, & Arunachalam, 2021; Huang & Snedeker, 2008). When paired with
other contrastive cues such as prosody, preschoolers can make inferences about novel object
typicality (Horowitz & Frank, 2016), and can use novel adjectives and nouns to restrict

reference (Diesendruck, Hall, & Graham, 2006; Gelman & Markman, 1985). Future work
should explore whether adjective contrast that is less scaffolded by other cues is a viable way
for children to learn about novel concepts.

The core computation in pragmatic inference is reasoning about alternatives—things the 893 speaker could have said and did not. Given that others are reasoning about these 894 alternatives, no choice is neutral. In the studies in this paper, for instance, using an adjective 895 in referring to an object led people to infer that the feature described by that adjective was 896 less typical than if it had not been mentioned. But, conversely, not using an adjective led 897 them to think that the feature was more typical than if they could not understand the 898 meaning of the utterance at all-all communicative choices leak one's beliefs about the world. 899 This has implications not only for learning about novel concrete objects, as people did here, 900 but for learning about less directly accessible entities such as abstract concepts and social 901 groups. These inferences can be framed positively, as ways for learners to extract additional 902 knowledge that was not directly conveyed, but can also spread beliefs that the speaker does 903 not intend. A core challenge will be to understand how people reason about the many 904 potential meanings a speaker might convey in naturalistic contexts to learn about others' 905 words for and beliefs about the world. 906

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