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Brief article

Infants rapidly learn word-referent mappings via cross-situational statistics

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Abstract

First word learning should be difficult because any pairing of a word and scene presents the learner with an infinite number of possible referents. Accordingly, theorists of children's rapid word learning have sought constraints on word-referent mappings. These constraints are thought to work by enabling learners to resolve the ambiguity inherent in any labeled scene to determine the speaker's intended referent at that moment. The present study shows that 12- and 14-month-old infants can resolve the uncertainty problem in another way, not by unambiguously deciding the referent in a single word-scene pairing, but by rapidly evaluating the statistical evidence across many individually ambiguous words and scenes.

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1. Introduction

The pairing of a word and a scene is not enough to determine the meaning of the word. To illustrate this point, Quine (1960) famously imagined a stranger who hears a native say “gavagai” and points to a scene. To what does “gavagai” refer – a rab-

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26 bit, the grass, a tree, the rabbit's ears, or perhaps the beauty of the whole? Even if
27 one assumes a perceptual system that segments the scene into separate objects and
28 an attentional system biased towards objects, the intended referent is indeterminate
29 from this one experience. Infants are like strangers who do not know the native lan-
30 guage, yet they solve this indeterminacy problem. This paradox – the uncertainty of
31 the referent in word-scene associations and the fact that infants learn object names
32 nonetheless – is a core theoretical problem in the study of early word learning. For
33 the past 30 years most research on children's word learning has concentrated on how
34 the learner resolves the ambiguity *at the moment the novel word is first encountered*.
35 Experimental studies leave no doubt that by the time they are 2 years old children do
36 this at least for object names. That literature points to attentional (Smith, 2000),
37 social (Baldwin, 1993; Tomasello, 2000), linguistic (Gleitman, 1990) and representa-
38 tional (Markman, 1990) constraints as crucial to children's ability to resolve referen-
39 tial ambiguity and fastmap a word to its intended referent.

40 There are two reasons to suspect that this one-encounter solution to referential
41 uncertainty is not the only (or even the most important) mechanism of early word
42 learning. First, not all opportunities for word learning are as uncluttered as the
43 experimental settings in which fast-mapping has been demonstrated. In everyday
44 contexts, there are typically many words, many potential referents, limited cues as
45 to which words go with which referents, and rapid attentional shifts among the many
46 entities in the scene. It is possible that young learners just ignore the information in
47 such highly ambiguous learning contexts and wait for contexts in which the referents
48 of heard words are more certain (Brent & Siskind, 2001). However, a more optimal
49 learner might be expected to make use of all the available data.

50 Second, the evidence indicates that 9-, 10-, and certainly 12-month-old infants are
51 accumulating considerable receptive lexical knowledge (Fenson, Dale, Reznick, &
52 Bates, 1994; Swingley & Aslin, 2000). Yet many studies find that children even as
53 old as 18 months have difficulty in making the right inferences about the intended
54 referents of novel words (e.g., Katz, Baker, & Macnamara, 1974; Hirsh-Pasek,
55 Golinkoff, & Hollich, 1999; Moore, Angelopoulos, & Bennett, 1999; Pruden,
56 Hirsh-Pasek, Golinkoff, & Hennon, 2006). There are studies showing that infants
57 as young as 13 or 14 months (Woodward, Markman, & Fitzsimmons, 1994; Wood-
58 ward & Hoyne, 1999; Schafer & Plunkett, 1998; but perhaps not younger, Werker,
59 Cohen, Lloyd, Casasola, & Stager, 1998) can link a name to an object given repeated
60 unambiguous pairings in a single session. Overall, however, these effects are fragile
61 with small experimental variations often leading to no learning (see especially,
62 Woodward & Hoyne, 1999; Werker et al., 1998; also Oviatt, 1980, 1982 & Bloom,
63 2000 for a discussion). This raises the possibility that there might be some other
64 way that young children learn word-referent mappings.

65 The experiment reported here shows for the first time that infants *rapidly* learn
66 *multiple* word-referent pairs by accruing statistical evidence across multiple and indi-
67 vidualy *ambiguous* word-scene pairings. The indeterminacy problem is solved not in
68 a single trial but across trials, not for a single word and its referent but for a data set
69 of many words and referents. This learning is shown to be sufficiently rapid and
70 robust that it could play a significant role in early lexical learning.

71 Fig. 1 illustrates how cross-trial statistics might work. The learner hears the
 72 unknown words “bat” and “ball” in the context of seeing a BAT and BALL. With-
 73 out other information, the learner cannot know whether the word form “ball” refers
 74 to one or the other visual object. However, if subsequently, while viewing a scene
 75 with the potential referents of a BALL and a DOG, the learner hears the words
 76 “ball” and “dog” and if the learner can combine the co-occurrence frequencies from
 77 the two streams of data across trials, the learner could correctly map “ball” to
 78 BALL. This example represents the simplest case of cross-situational statistical
 79 learning – two words, two objects, two adjacently informative trials.

80 Several formal simulations of word-referent learning suggest the plausibility of
 81 cross-situational word learning in much more complex situations with many words,
 82 many possible referents, highly ambiguous individual learning trials, and the statisti-
 83 cal resolution of the ambiguities only through the accumulation and evaluation of
 84 information over many word-referent pairings and many trials (Siskind, 1996; Yu,
 85 Ballard, & Aslin, 2005). Consider the more complex case in Table 1. On trial 1, a
 86 learner could mistakenly link word A to referent b. On trial 4, the mistake could
 87 be corrected, *if* the system registers that word A occurred on trial 4 *without* possible
 88 referent b, *if* the cognitive system remembers the prior word-referent pairing, *if* it
 89 registers both co-occurrences and non co-occurrences, and *if* it calculates the right
 90 statistics. Can babies do this?

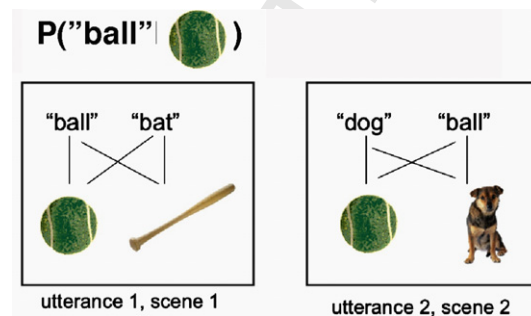


Fig. 1. Associations among words and referents across two individually ambiguous scenes. If a young learner calculates co-occurrences frequencies *across* these two trials, s/he can find the proper mapping of “Ball” to BALL.

Table 1
An example of cross-situational learning

Trial	Words	Potential referents in scene
1	AB	ba
2	CD	dc
3	EF	ef
4	GA	ga

91 There is evidence in such phenomena as the mutual-exclusivity effect and
92 contrast that 2- to 3-year-old children combine information across two adjacent
93 naming events, using, for example, knowledge of the just-heard name of one
94 thing to infer the object to which a subsequent name must apply (Akhtar,
95 2002; Akhtar & Montague, 1999; Markman, 1990; Namy & Gentner, 2002.)
96 However, there is no evidence as to whether young learners can combine
97 and evaluate information from highly ambiguous contexts over many trials.
98 Until recently, there was no evidence as to whether even adult learners were
99 capable of this, although Yu and Smith (2007a, 2007b) have now shown that
100 this form of learning is rapid and robust in adults even in situations of high
101 uncertainty.

102 In the following experiment, 12- and 14-month-old infants were taught 6 word-
103 referent pairs via a series of individually ambiguous trials. On each trial, two word
104 forms and two potential referents were presented with no information about which
105 word went with which referent. Although word-referent pairings were ambiguous
106 within individual trials, they were certain across trials. For example, for a particular
107 infant, whenever the form *tobi* occurred its assigned referent always occurred. After
108 training, infants were presented with a single word and two potential referents, the
109 cross-trial correct referent and a foil. Past research (e.g., Golinkoff, Hirsh-Pasek,
110 Cauley, & Gordon, 1997; Swingley & Aslin, 2000) shows that within this kind of
111 preferential looking task, infants look longer at the labeled test object. Thus if
112 infants have calculated the statistics appropriately, despite the uncertainty on indi-
113 vidual learning trials, they should look longer at the correct referent of the word
114 form.

115 2. Method

116 2.1. Participants

117 The participants, drawn from a working and middle-class population of a
118 midwestern college town, were 28 12-month-old infants (range – 11 mo 17 days
119 to 13 mo 0 days; mean – 12 mo 7 days; 13 males, 15 females) and 27 14-
120 month-old infants (range –14 mo 2 days to 15 mo 14 days; mean –14 mo 12
121 days; 14 males, 13 females). Two additional children began but did not finish
122 the experiment.

123 2.2. Stimuli

124 The 6 “words” – *bosa*, *gasser*, *manu*, *colat*, *kaki* and *regli* – followed the phono-
125 tactic probabilities of English and were recorded by a female speaker in isolation
126 and were presented to infants over loudspeakers. The 6 “objects” were drawings
127 of novel shapes, shown in Fig. 2; each was a unique bright color. On each trial,
128 two objects (12 by 14 in. in projected size and separated on the screen by 30 in.) were
129 simultaneously presented on a 47 by 60 in. white screen.

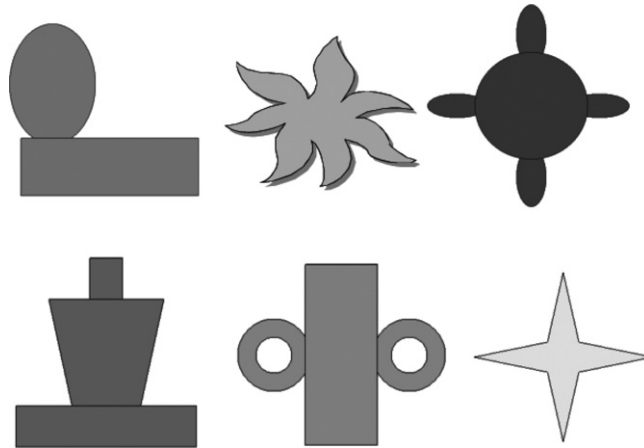


Fig. 2. The six stimulus shapes.

130 2.3. Procedure

131 Infants sat (on their mother's lap) 3.5 feet in front of screen with the mother's
132 chair set at the center of the screen. Infants' direction of eye gaze was recorded from
133 a camera centered at the base of the screen and pointed directly at the child's eyes.
134 Parents were instructed to keep their own eyes shut through the entire procedure so
135 as to not to influence their infant's behaviors. A camera directed on the parent
136 through out the procedure confirmed their adherence.

137 There were 30 training slides. Each presented two objects on the screen for 4 s; the
138 onset of the slide was followed 500 ms later by the two words – each said once with a
139 500 ms pause between. Across trials, the temporal order of the words and spatial
140 order of the objects were varied such that there was no relation between temporal
141 order of the words and the spatial position of the referents. Each correct word-object
142 pair occurred 10 times. The two words and two objects appearing together on a slide
143 (and creating the within trial ambiguities and possible spurious correlations) were
144 randomly determined such that each object and each word co-occurred with every
145 other word and every other object at least once across the 30 training trials. The first
146 four training trials each began with the centered presentation of a Sesame Street
147 character (3 s) to orient attention to the screen. After these first four trials, this atten-
148 tion grabbing slide was interspersed every 2–4 trials to maintain attention. The entire
149 training – an effort to teach six word-referent pairs – lasted less than 4 min (30 train-
150 ing slides and 19 interspersed Sesame Street character slides).

151 There were 12 test trials, each 8 seconds. This duration was chosen from pilot
152 studies to optimize the number of participants able to complete all 12 test compar-
153 isons (2 per target word). Each test trial presented one word, repeated 4 times with 2
154 objects – the target and a distracter – in view. The distracter was drawn from the
155 training set. Each of the 6 words was tested twice. The distracter for each trial

156 was randomly determined such that each object occurred twice as a distracter over
157 the 12 test trials.

158 There were 2 unique sets of training slides with different orderings of objects, dif-
159 ferent mappings of words to the objects, and different combinations of word-referent
160 pairs on the slides. For each set, the left–right locations of objects on the slides and
161 the order with which the names were presented were randomly generated with the
162 constraint that the object on the left was the target referent for the first presented
163 word on half the trials and the target referent for the word presented second on
164 the other half. There were also two unique test orders with unique randomly gener-
165 ated pairings of target and distracter, with the target appearing on the left on half the
166 slides on the right on the other half. Half the infants at each age level were randomly
167 assigned to each slide set.

168 Two coders naïve to condition and trial type coded direction of eye gaze from the
169 video recorded from the camera directed at the infant’s eyes. They coded, frame-by-
170 frame, all frames from the start to the end (indicated by light on the video) of each
171 training and test trial. The coder’s task for each frame was to categorize the direction
172 of look as right, left or away from the screen (hands, ceiling, mother’s face, floor,
173 etc.). For reliability, the two coders each coded the same random sample of 25%
174 of the frames. Agreement on these frames was 90.8%.

175 3. Results

176 3.1. Training trials

177 Infants were highly attentive to the training slides, looking (sum of right and left
178 looks) at each 4 s slide on average 3.27 s (12 month olds) and 3.04 s (14 month olds).
179 On average, infants looked at the left and right sides of each training slide for equal
180 durations ($t < 1.00$ for both 12- and 14-month olds). On 87% of all training slides,
181 the infants looked at both sides (both objects) for at least 1 s.

182 3.2. Test trials

183 On average, infants looked at each 8 s test slides for a total of 5.6 s for 12 month
184 olds and 6.1 s for 14 month olds. To examine whether infants *preferentially* looked in
185 the direction of the target object, the object that *across trials* was associated with the
186 auditorally presented label, each infant’s looking time to target and distracter on
187 each test trial was submitted to a 2(Age) by 2(Target/Distracter) \times 6 (Word) \times 2(-
188 Block – first or second test of each target word) analysis of variance for a mixed
189 design. The analysis revealed a highly reliable main effect of looking time to Tar-
190 get/Distracter, $F(1, 54) = 35.32$, $p < .001$ (partial eta squared = .37). As shown in
191 Fig. 3, 12- and 14-month-old children looked reliably longer to the Target than to
192 the Distracter. The analysis also revealed a reliable interaction between Word and
193 Target/Distracter, $F(5, 54) = 3.85$, $p < .05$ (partial eta squared = .19). This result,
194 that infants showed a greater difference in looking time to the target than distracter

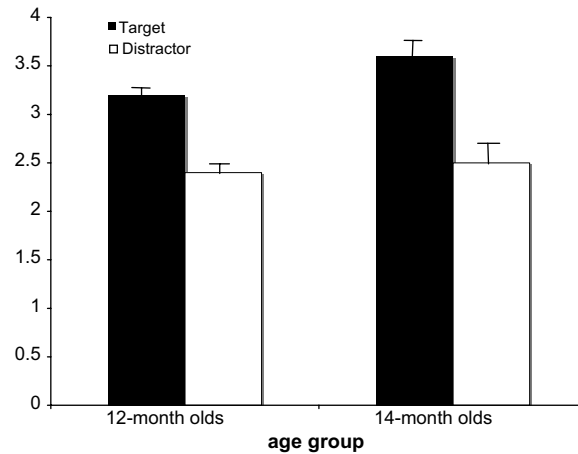


Fig. 3. Mean looking time to target and distracter per 8 s test trial (and standard error of the mean) for younger and older infants.

195 for some words than for others suggests that some word-picture correspondences
 196 were learned better than others. Finally, the analysis revealed an interaction between
 197 Age and Target/Distracter that approached significance, $F(5, 54) = 3.13$, $p < .08$
 198 (partial eta squared = .04). The older group of children, as can be seen in Fig. 3,
 199 showed a bigger preference for the target than did the younger children, although
 200 the difference in looking times to target and distracter is individually reliable for both
 201 age groups (Tukey's hsd, $p < .05$). No other main effects or interactions approached
 202 significance.

203 Post hoc analyses (Tukey's hsd, $p < .05$) conducted on the difference in looking
 204 time to target and distracter for the 6 individual words indicated reliably greater
 205 looking time to target than distracter for 4 of the 6 words for the 12-month-old
 206 group and for 4 of the 6 words for the 14-month-old age group. (Three of the indi-
 207 vidual words were the same at the two age levels, one was different; at neither age
 208 level were there reliable differences in the wrong direction for the remaining two
 209 words.) Since half the children at each age level had different word-object pairings
 210 as well as different training orders, and since analyses for effects of slide set yielded
 211 no effects or interactions that approached significance, the source of these differences
 212 is not obvious. However, the fact that looking times for 4 of the 6 words (67% of the
 213 training set) show reliable preferences for the target does indicate that infants can
 214 figure out *multiple* word-referent mappings from a *system* of experienced associa-
 215 tions. Finally, the group patterns appear to characterize the performance of individ-
 216 ual infants in that 46 of the 55 participants infants looked, on average, at the targets
 217 more than distracters.

218 In sum, these results tell us that cross-situational statistical learning is in the rep-
 219 ertoire of young word learners. Despite the ambiguity of word-referent mappings on
 220 any individual training trial, infants clearly accumulate information across trials and
 221 use that information to determine the underlying mappings. In less than four min-

222 utes, with six different word forms and six different objects, infants learned enough to
223 systematically look longer at the objects more strongly associated with the forms
224 than those more weakly associated.

225 4. General discussion

226 Parents, on average, direct between 300 and 400 utterances an hour to their
227 children (Hart & Risley, 1995). Even with social, linguistic and conceptual con-
228 straints in play, so many words in so little time seems likely to generate consid-
229 erable ambiguity about intended referents. These ambiguities are most likely
230 greater than those in this experiment. Nonetheless, the mechanisms responsible
231 for the present results may be relevant to *making use* of the complexity in natural
232 learning environments in that these mechanisms can keep track of multiple word-
233 referent co-occurrences, evaluate the regularities in the data set as a whole, and
234 determine the underlying mappings. Such mechanisms could even benefit from
235 increased complexity in the data set. Consistent with this idea, Yu and Smith
236 (2007a, 2007b), using a task much like the infant task used here, showed that
237 adults actually learned more word-referent pairs when the set contained 18 words
238 and referents than when it contained only 9. This is because more words and ref-
239 erents mean better evidence against spurious correlations. Although much remains
240 to be discovered about the relevant mechanisms, they clearly should help children
241 learn from the regularities that accrue *across* the many ambiguous word-scene
242 pairings that occur in everyday communication.

243 The present findings are thus reminiscent of evidence showing that infants' use
244 sequential probabilities to discover segmental units in speech (Gomez & Gerken,
245 1999; Kirkham, Slemmer, & Johnson, 2002; Saffran, Aslin, & Newport, 1996).
246 The statistical regularities to which infants must attend to learn word-referent pair-
247 ings are different from those underlying the segmentation of a sequential stream in
248 that word-referent pairings require computing co-occurrence frequencies across
249 two streams of events (words and referents) simultaneously for many words and ref-
250 erents. Nonetheless, the present findings, like the earlier ones showing statistical
251 learning of sequential probabilities, suggest that solutions to fundamental problems
252 in learning language may be found by studying the statistical patterns in the learning
253 environment and the statistical learning mechanisms in the learner (Newport &
254 Q1 Aslin, 2004; Saffran et al., 1996).

255 There are several possible learning mechanisms that could accomplish the cross-
256 situational learning of word-referent mappings. One is the formulation and evalua-
257 tion of hypotheses (e.g., Tenenbaum & Xu, 2000). Building on the “ball/bat” exam-
258 ple in Fig. 1, the learner could, for example, wrongly hypothesize on the initial trial
259 that “ball” refers to BAT but correct that hypothesis on trial 2, which presents dis-
260 confirming evidence. Given enough data across individually ambiguous trials, the co-
261 occurrence frequencies would support the “right” hypotheses for the language over
262 others. The outcome of this learning would seem to be a list of confirmed hypotheses,
263 each specifying a word and its referent.

264 However, statistical learning need not be the result of highly specialized statistical
265 learning mechanisms (e.g., Perruchet & Vinter, 1998). The learner could solve this
266 learning task via simple (or not so simple, see Kruschke, 2001; Kruschke, Kappen-
267 Q2 man, & Hetrick, 2005; Yu & Smith, 2007a, 2007b) associative learning mechanisms.
268 Across trials, the learner could accumulate associations between words and potential
269 referents by strengthening and weakening associative links with each co-occurrence
270 or nonco-occurrence (see, Plunkett, 1997). Building on the example in Fig. 1, the
271 learner could equally associate “ball” with BALL and BAT but after the experience
272 of “ball” in the context of BALL and DOG, the association between “ball” and
273 BALL would be stronger than that between “ball” and BAT. Over enough trials,
274 these association strengths would converge on the real world statistics. The outcome
275 of this learning, unlike the hypothesis testing account, might not be knowledge that
276 an individual word refers to one thing, but may only be stronger correct associations
277 than spurious ones.

278 The present results cannot distinguish these possibilities. Perhaps early asso-
279 ciative learning lays the ground work in infancy for more rapid (and perhaps
280 more hypothesis-testing like) processes in later word learning. A recent simula-
281 tion study by Yu (in press) makes this point. That study examined a probabi-
282 listic associative learning mechanism that learns a *system* of associations
283 (Yoshida & Smith, 2003). In such a system, a single word-referent pairing is
284 correlated with all the other pairings that share the same word and all the other
285 pairings that share the same referent, which are in turn correlated with more
286 word-referent pairs, yielding a system of correlations. Such large systems of
287 associations create system-wide accelerations of word-referent learning even when
288 the individually contributing associations are partially learned. This is because a
289 system of even partially learned associations yields latent structure that can be
290 used to guide subsequent learning. In this way, the lexical knowledge of 12-
291 to 14-month olds, even if based on associations, could contribute significantly
292 to the later, more rapid, and more seemingly sophisticated one-trial word learn-
293 ing of older children. Regardless of which kind of mechanism proves right, the
294 present results suggest that the relevant mechanisms may be best conceived as
295 not being about the learning of individual words and referents – not about
296 the testing of individual hypotheses or the learning of single associations –
297 but rather as being about processes that evaluate the regularities in data sets
298 of many words and referents (a point originally made by Billman & Knutson,
299 1996). The human learning environment is data rich. If human learners possess
300 the right learning mechanisms, they may mine this complexity and in so doing
301 solve the problem of referential uncertainty.

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