

# Communicative pressure can lead to input that supports language learning

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## Abstract

Infants prefer to listen to and learn better from child-directed speech. This speech might support learning in part due to communicative pressure: parents must use language that their children understand. We designed a Mechanical Turk study to experimentally validate this idea, putting Turkers in the role of parents talking with children who knew a novel language less well. Participants could communicate in 3 ways: pointing—expensive but unambiguous, labelling—cheap but knowledge-dependent, or both. They won points only for communicating successfully; using pointing and labelling together was costly, but could teach, allowing cheaper communication on later trials. Participants were modulated their communicate behavior in response to their own and their partner’s knowledge and teaching emerged when the speaker had substantially more knowledge of the lexicon. While language is more than reference games, this work validates the hypothesis that communicative pressure alone can lead to supportive language input.

**Keywords:** Language; Child-Directed Speech; POMDP.

## Introduction

One of the most striking aspects of children’s language learning is just how quickly they master the complex system of their natural language (Bloom, 2000). In just a few short years, children go from complete ignorance to conversational fluency that is the envy of second-language learners attempting the same feat later in life (Newport, 1990). What accounts for this remarkable transition?

One possibility is that children’s caregivers deserve most of the credit; that the language parents produce to their children is optimized for teaching. Although there is some evidence that aspects of child-directed support learning, other aspects—even in the same subproblem, e.g. phoneme discrimination—appear to make learning more difficult (Eaves Jr, Feldman, Griffiths, & Shafto, 2016; McMurray, Kovack-Lesh, Goodwin, & McEchron, 2013). In general, parents rarely explicitly correct their children, and children are resistant to the rare explicit language correction they do get (Newport, Gleitman, & Gleitman, 1977). Thus while parents may occasionally offer a supervisory signal, the bulk of the evidence suggests that parental supervision is unlikely to explain rapid early language acquisition.

Alternatively, even the youngest infants may already come to language acquisition with a precocious ability to learn the latent structure of language from the statistical properties of the language in their ambient environment (Saffran & 2003, 2003; L. B. Smith & Yu, 2008). While a number of experiments clearly demonstrate the early availability of such precocisms, there is reason to be suspicious about just how precocious they are early in development. For example, infants’ ability to track the co-occurrence information connecting words to their referents appears to be highly constrained

by both their developing memory and attention systems (L. B. Smith & Yu, 2013; Vlach & Johnson, 2013). Further, computational models of these processes show that the rate of acquisition is highly sensitive to variation in environmental statistics (Blythe, Smith, & Smith, 2010; Vogt, 2012). Thus precocious unsupervised statistical learning also appears to fall short of an explanation for rapid early language learning.

In this paper we explore the consequences of a third possibility: The language that children hear is neither designed for pedagogy, nor is it random: it is designed for communication (Brown, 1977). We take as the caregiver’s goal the desire to communicate with the child, not about language itself, but instead about the world in front of them. To succeed, the caregiver must produce the kinds of communicative signals that the child can understand, and thus might tune the complexity of their speech not for the sake of learning itself, but as a byproduct of in-the-moment pressure to communicate successfully (Yurovsky, 2017).

We take as our model system a simple iterated reference game in which two players earn points for communicating successfully with each-other about the objects in front of them. First in a computational model, and then in a set of experiments with adults on Mechanical Turk, we show that pedagogically-supportive input can arise from purely selfish motives to maximize the cost of communicating successfully while minimizing the cost of communication. We take these results as a proof of concept that both the features of child-directed speech that support learning as well as those that inhibit it may arise from a single unifying goal: The desire to communicate efficiently.

## Model: Communication as Planning

We implemented a two-agent model of speaker and listener behavior in this game based on partially observable Markov decision processes. The speaker is given a target referent each trial and must signal to the listener which object to select. Speakers send messages to the listener by speaking, a low cost cue that relies on the listener’s knowledge, or by pointing, a higher cost cue that is unambiguous.

The speaker estimates the listener’s knowledge. First, the speaker uses Markov chain Monte Carlo to infer their own learning rate based on how well they were able to learn a novel lexicon after  $N$  exposures. Then, given the listener’s degree of exposure, the speaker can use their own learning rate to infer the probability that the listener would know any given object-label mapping. Across trials, the speaker gains further information about the listener through their selections, allowing them to update their beliefs about the listener’s knowledge

state.

The model specifies the relative costs for each communicative modality. On each trial, the speaker estimates the expected utility of each modality by accounting for these costs, their own knowledge of the object's label, and the probability that the listener knows the object's label. The speaker then uses Luce's choice axiom to select a communicative modality based on the expected utilities.

When estimating expected utility, the model sums the expected utility of a given trial and any remaining trials for that particular object. This allows the speaker to engage in planning by accounting for the way a given message may affect knowledge changes and thus affect subsequent expected utilities. Utilities were scaled using an exponential discounter as a function of delay (Do we want to cite a precedent for this?) to give greater weight to immediate rewards than subsequent rewards.

Crucially, speakers can also combine both communication cues, paying the upfront cost of both, to produce a message that is both unambiguous and informative. In this way, speakers are able to teach their partners object-label mappings. A speaker that plans may thus infer that teaching the listener, especially if there is an asymmetry in their knowledge states, may have a high expected utility after accounting for remaining trials where the speaker could use a less costly cue (i.e. speech). After producing both cues, the speaker also updates their own beliefs about the partner's knowledge state to reflect this exposure.

Over the course of three experiments, [TOTAL N] participants were recruited via Amazon Mechanical Turk, an online platform that allows workers to complete surveys and short tasks for payment. Turkers played partnered, iterated reference games. In these studies, all participants were placed in the role of speaker and listener responses were programmed.

## Experiment 1a

Our model's speaker chooses a communicative modality by optimizing the expected utilities of the possible modalities. By manipulating the utility of each strategy, we can evaluate our model's predictions about the extent to which speakers sensibly adapt to these imposed utilities. In this experiment, we explicitly implemented strategy utility by assigning point values to each strategy.

## Participants, Materials, Methods

[n] participants were recruited through Amazon Mechanical Turk and received a small payment for their participation.

Participants were told they would be introduced to novel object-label pairs and then asked to play a communication game with a partner wherein they would have to refer to a particular target object. Participants were exposed to nine novel objects, each with a randomly assigned pseudoword label. We manipulated the degree of exposure within subjects, such that, during training, participants saw three of the nine object-label mappings four times, two times, or one time, resulting in a total of 27 training trials. Participants were then given a

simple recall task to establish their baseline knowledge of the novel lexicon.

After being introduced to the rules of the game, participants were told they would be asked to communicate about each object three times. During gameplay, speakers saw the target object in addition to an array of all nine objects. Speakers had the option of either directly selecting the target object from the array (gesture)- a higher cost cue but without ambiguity- or typing a label for the object (speech)- a lower cost cue but contingent on the listener's knowledge. After sending the message, speakers are shown which object the listener selected.

If the speaker clicked on an object (gesture message), the listener was coded to simply make the same selection. If the speaker typed an object label, the listener evaluated the Levenshtein distance (LD) between the typed label and each of the nine possible labels and selected the candidate with the smallest edit distance. If none of the nine object labels had an LD less than or equal to two, the listener always selected an incorrect object.

Speakers could win up to 100 points per trial if the listener correctly selected the target referent based on their message. If the listener failed to identify the target object, the speaker received no points. We manipulated the relative utilities of each of the strategies between-subjects. In the 'Talk is Cheap' condition, speakers received 30 points for gesturing and 100 points for labeling, while in the 'Talk is Less Cheap' condition speakers received 50 points for gesturing and 80 points for labeling.

## Results

As expected, participants were sensitive to both the exposure rate and the relative utilities of the communication strategies. Speakers produced significantly more labels for object-label mappings that were shown more frequently during training than for the mappings that were less frequent (...). The reverse was true for gesture (...). The more exposures to a given object-label pair during training, the more likely participants were to recall that label at pretest (...). The effect of exposure rate on modality choice held even after controlling for pretest knowledge, indicating the effect was more than the result of being constrained by knowledge of the label (...). ... [I'm not sure which statistics to include here, since most of what I'd been looking at are these logistic regression models with all our predictors. Should I subset to just one partner's exposure, and then do the logistic regression with exposure rate, utility condition, appearance, and the random effect? Does that make sense?]. ...

Participants also modulated their communicative behavior on the basis of the utility manipulation. Speakers in the Talk is Cheap condition produced significantly more labels than participants in the Talk is Less Cheap condition (...). Again, this effect was reversed for gesture (...). Thus, participants are sensitive to our manipulations- altering their choices about how to communicate with their partner on the basis of their

own knowledge, the degree of training, and the imposed utilities.

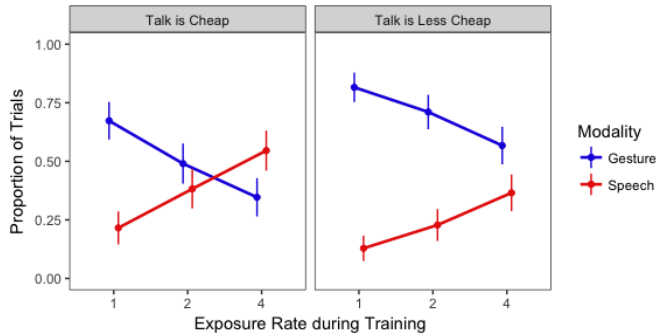


Figure 1: Plot of modality choice as a function of exposure, split by utility condition.

## Experiment 1b

As a replication and extension of Experiment 1a, we sought to embed the relative utilities of the communicative strategies in the game more implicitly. In experiment 1b, we used the same procedure described above; however, a timer controlled the number of points that were earned on each trial of gameplay. Rather than impose an explicit point system as in Experiment 1a, we set the relative utilities of communication strategies by manipulating the speed with which participants could respond.

### Participants, Materials, Methods

[n] participants were recruited through Amazon Mechanical Turk and received a small payment for their participation.

Experiment 1b utilized the same training and pretest procedures, but gameplay differed in a number of ways. During the game, the nine novel objects were arranged in a circle around a cartoon pointer (see Figure 2). To manipulate the relative speeds of gesturing and labeling, we required speakers to move the small pointer with the arrow keys on their keyboards in order to select an object with gesture. With this method we could directly manipulate the speed of the pointer while holding the arrow keys to effectively set maximum possible utility of using gesture.

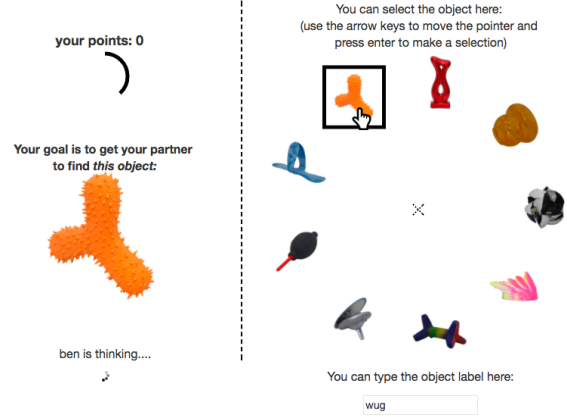


Figure 2: Screenshot of speaker view during gameplay.

If the listener correctly identified the target object, speakers received points determined by the amount of time remaining on the timer. For successful communications, speakers always received a minimum of 10 points, regardless of the time elapsed. Based on pilot response times, the timer was set to 6.5 seconds and points were scaled to match Experiment 1a. Participants were randomly assigned into one of three conditions, where the pointer was set to either a fast, slow, or slowest setting.

This framework allowed us to implement the point system as an implicit extension of gameplay and resulted in each participant having unique gesture-speech utilities, based on their speed of responses across the two methods.

### Results

As expected, the same patterns emerge as experiment 1a. Speakers again labeled significantly more (TBD) and gestured significantly less (TBD) on high exposure targets, relative to low exposure targets. Participants were significantly more likely to recall the high exposure labels than the low exposure labels at pretest (TBD), and this effect did not account the effect of exposure rate on modality choice (TBD).

Critically, there was a main effect of our utility manipulation, such that speakers in the fast pointer condition produced significantly more gestures than participants in the slowest pointer condition (TBD). The pattern is reversed for labeling (TBD). In the absence of an explicit framework for assigning utilities based on method, speakers continue to adapt their communicative choices taking into account the expected utility of possible strategies.

## Experiment 2a

[I don't know how to divvy up these ns]

[n] participants were recruited through Amazon Mechanical Turk and received a small payment for their participation.

Thus far, we have focused on relatively straightforward scenarios to demonstrate that a pressure to communicate successfully in the moment can lead speakers to trade-off be-

tween gesture and speech sensibly. However, critical to these repeated interactions is the ability to learn about an interlocuter and potentially influence their learning. In Experiment 2a, participants could send a third type of message by utilizing both gesture and speech within a single trial to effectively teach the listener an object-label mapping.

### Participants, Materials, Methods

In order to produce teaching behavior, speakers had to pay the cost of producing both cues, yielding 30 points in either condition of our explicit points framework. Our communicative game was designed to reward in-the-moment communication, and teaching required the speaker paying a cost upfront. However, rational communicators may understand that if one is accounting for future trials, paying the cost upfront to teach the listener allows a speaker to use a less costly message strategy on subsequent trials.

Incorporating teaching means that our speaker must now reason about their interlocuter's knowledge state more explicitly, in order to make rational decisions about what to teach and when. To address this added dimension, we also manipulated participants' expectations about their partner's knowledge. Prior to beginning the game, participants were told that their partner had no experience with the lexicon, had half the experience of the speaker, or had twice the experience of the speaker.

Listener starting knowledge state was also initialized accordingly. Listeners with no exposure were given no knowledge of the lexicon to start. Listeners with half the exposure of the speaker began with knowledge of three object-label pairs (2 high frequency, 1 mid frequency), based on the typical retention rates found previously. Lastly, the listener with twice as much exposure as the speaker began with perfect knowledge of the lexicon. Through gesture-speech co-occurrence, the speaker could effectively teach the listener a novel mapping. Listeners could learn a novel mapping from one teaching event in this game and integrate it into their set of candidate words when evaluating the LD of subsequent labels.

[I'm not sure how to describe the methods here]. [Experiment 2a directly mirrored Experiment 1a, except that participants were introduced to an additional strategy, i.e. teaching, during the introduction to the game (feels a little wonky).]

### Results

As predicted, mixed effects logistic regression predicting whether or not teaching occurred on a given trial revealed that teaching rates across conditions depend on all of the same factors that predict speech and gesture. There was a significant effect of initial exposure to the mapping on the rates of teaching, such that more exposures to a word predicted higher rates of teaching behavior ( $B = 0.09$ ,  $p = 0.045$ ). There was also a significant effect of the utility manipulation such that being in the Talk is Less Cheap condition predicted lower rates of teaching than being in the Talk is Cheap condition ( $B = -0.98$ ,  $p < 0.01$ ), a rational response considering the eventual benefit for teaching a novel word is less strong in this former

condition.

We also included whether this was an object's first, second, or third appearance as a predictor in our model. The expected utility of teaching on a given trial should decrease as there are fewer subsequent trials for that object, thus we predicted that teaching rates would drop dramatically. Indeed, this is consistent with the results from our model; compared with the first appearance of an object, speakers were significantly less likely to teach on the second appearance ( $B = -1.11$ ,  $p < 0.001$ ) or the third appearance ( $B = -2.01$ ,  $p < 0.001$ ).

Interestingly, as we expected, there was a significant effect of our manipulation of listener knowledge. Compared with listeners with no experience with the lexicon, speakers were significantly less likely to teach when they were told the listener had half as much exposure ( $B = -1.19$ ,  $p < 0.01$ ) or twice as much exposure as they themselves did ( $B = -4.64$ ,  $p < 0.001$ ).

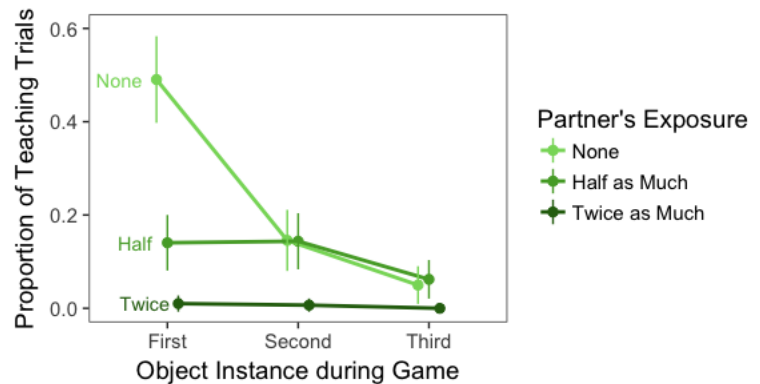


Figure 3: Proportion of teaching behavior as a function of whether it was the first, second, or third appearance of a given target, split by partner's level of exposure.

### Experiment 2b

We investigated the emergence of teaching in both a paradigm with an explicit points system (Experiment 2a) and an implicit points system (Experiment 2b).

### Participants, Materials, Methods

### Results

### Experiment 3

Planning is critical to our model. A pressure to communicate successfully in-the-moment can lead to pedagogical-like input without pedagogical goals, merely by wishing to communicate more efficiently in the future. As a final check to ensure that accounting for the future is indeed crucial to the emergence of pedagogical behaviors in our framework, we ran a version of our partnered game where speakers are told they will only send messages about each object 1 time. As expected, teaching behavior dropped off completely. . . (TBD).

### Model Comparison

[Is this where the actual model fit stuff should live?]

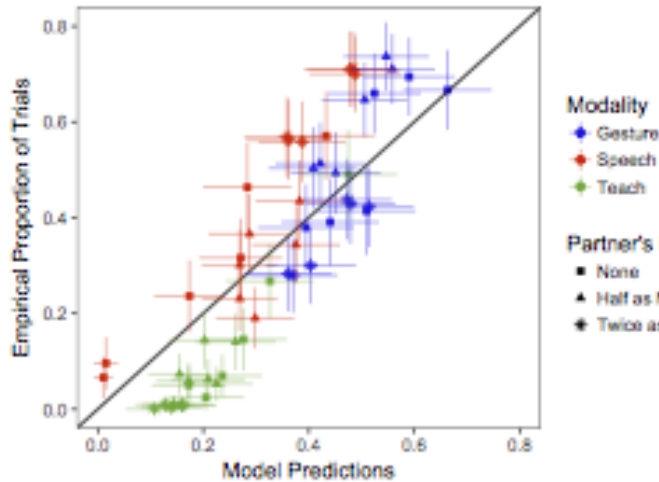


Figure 4: Plot of the fit between model predictions and empirical data.

## Discussion

## Acknowledgements

Place acknowledgments (including funding information) in a section at the end of the paper.

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