



# Modeling Communication to Coordinate Perspectives in Cooperation

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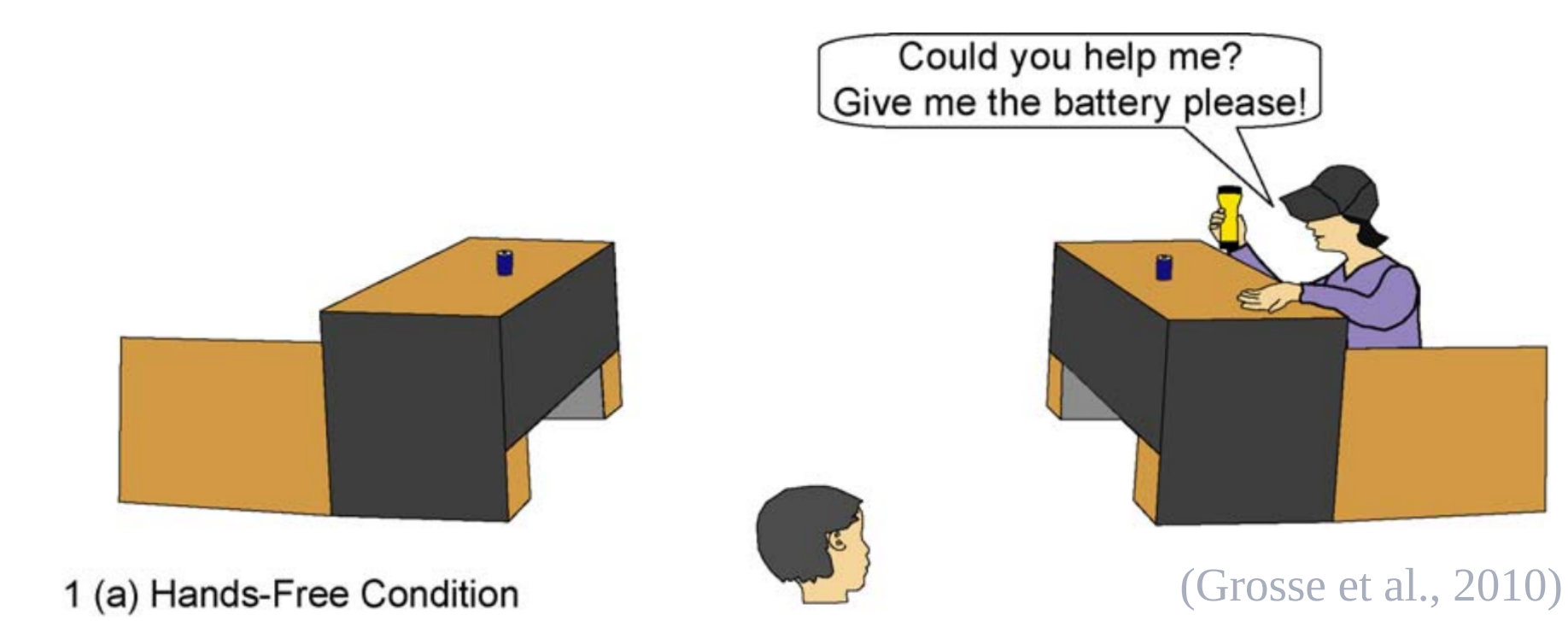
## Motivation

Emergence of communication — an adaptation for human cooperation?

Still, communication is highly overloaded

“Point to a piece of paper. And now point to its shape — now to its color — now to its number ... How did you do it?”  
-Wittgenstein, 1953

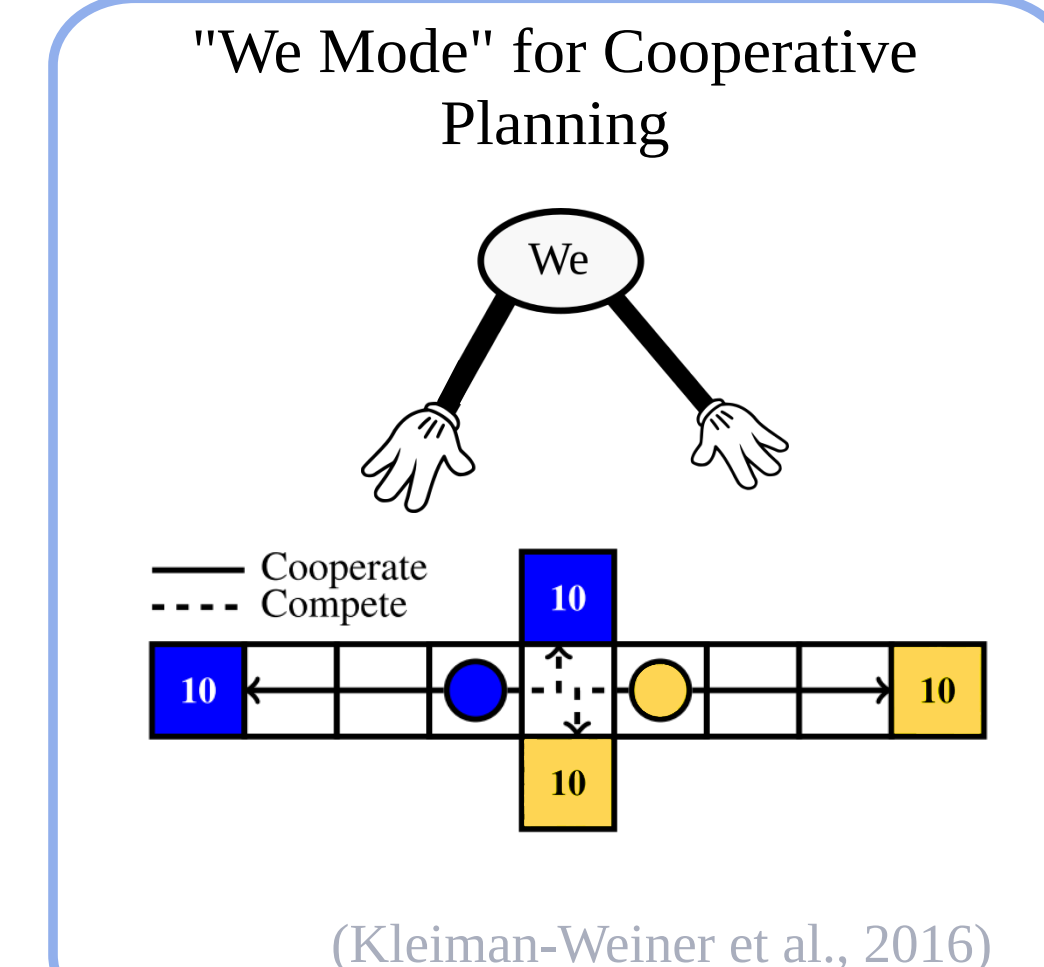
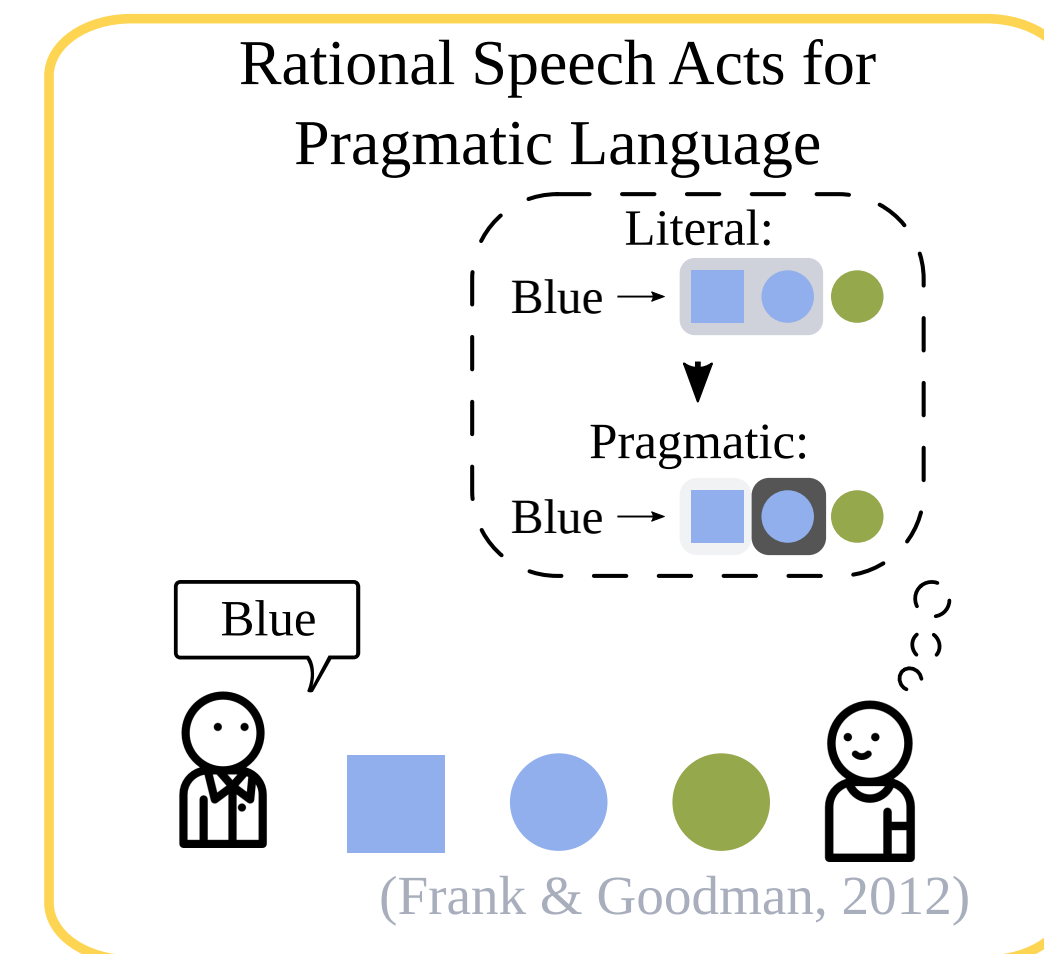
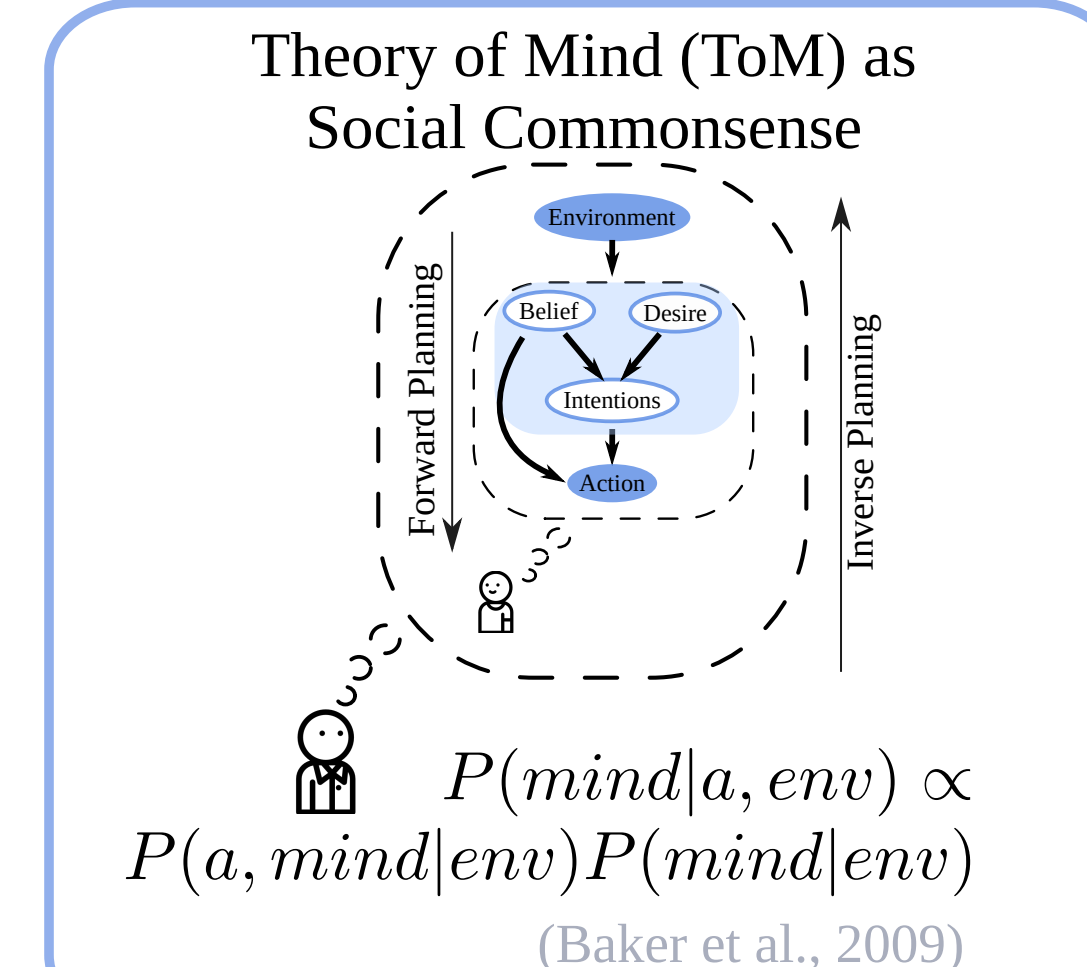
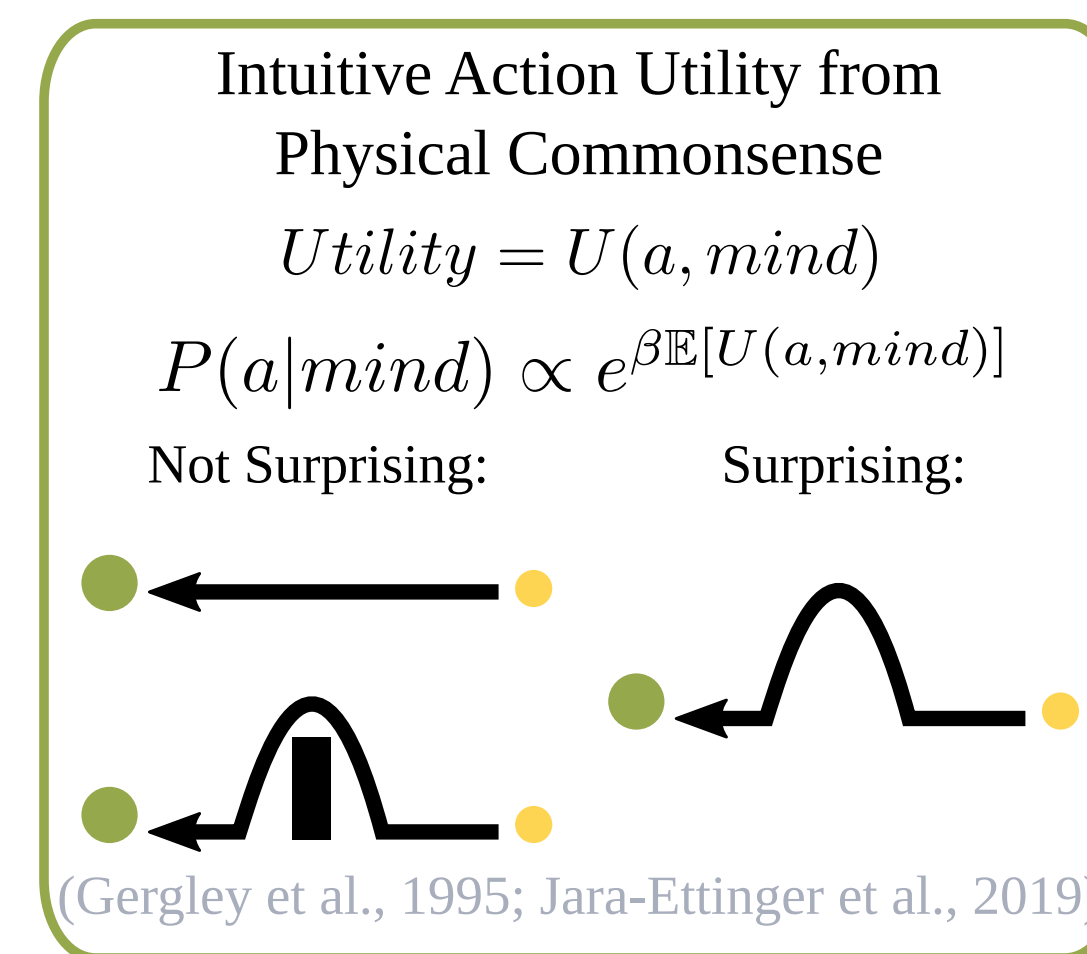
Despite this, even young children are good at interpreting cooperative requests



How can we model cooperative communication under this ambiguity?

## Model Components

### Building Blocks



### Challenges

1. ToM is designed to handle actions not communication

2. The utility of a signal is not straightforward to define

3. Rationality is defined in terms of self interests

### Solutions

Treat signals as rational actions. Subject them to the same modeling framework as actions

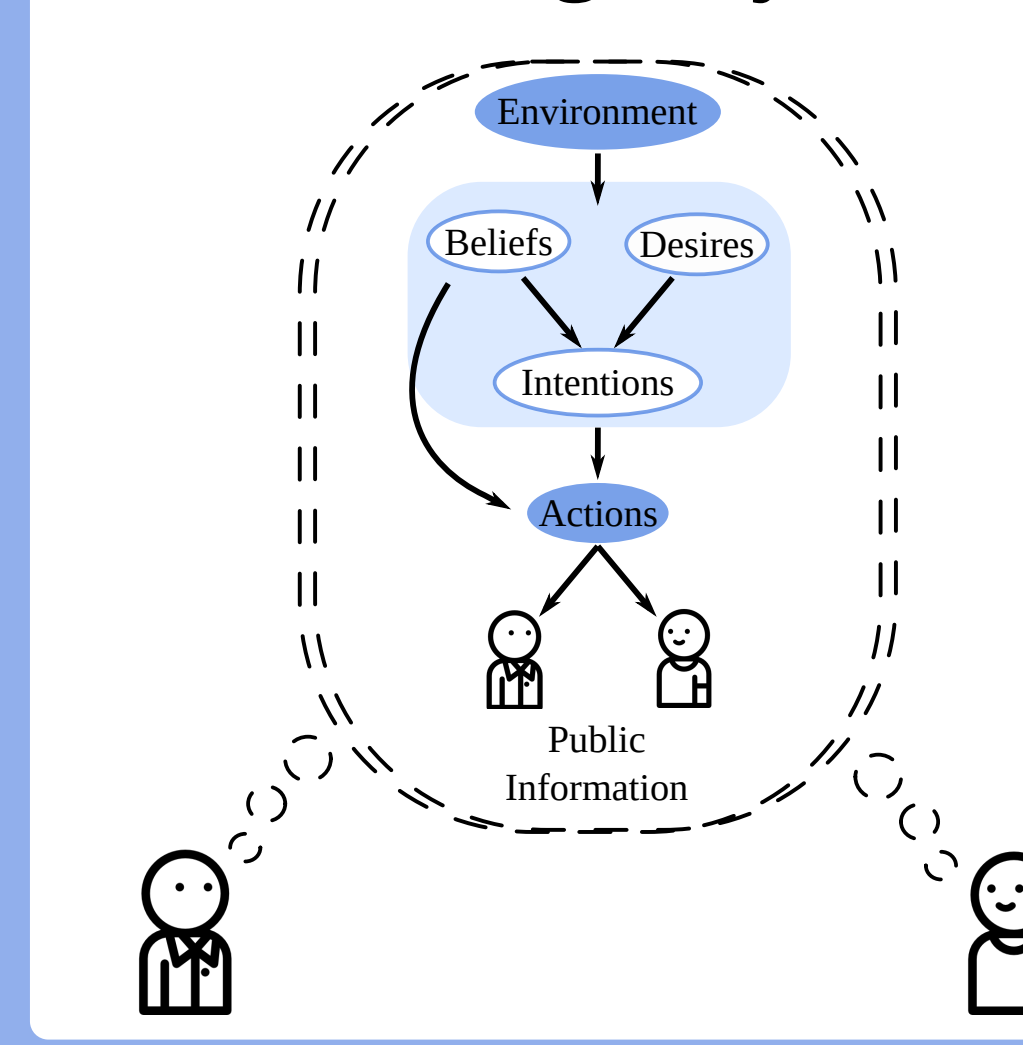
Signals change the mind. The mind determines rational actions. Actions have well defined utilities

Selfishness creates ambiguity. Cooperative logic as a joint utility can resolve the ambiguity

### Imagined We for Communication

Rational Speech Acts Style Pragmatics

Shared Agency ToM



Joint Utility Calculus

Signal

Joint Mind

Joint Action

## Computational Model

Given private knowledge, a signaler **chooses a signal**

$$P(signal|mind_t) \propto e^{\beta \mathbb{E}[U(signal, mind_t)]}$$

According to **how good that signal is** expected to be

$$\mathbb{E}[U(signal, mind_t)] = \mathbb{E}_{P(a|signal)}[U(a, mind_t)]$$

Which is measured by how likely cooperators **take an action** upon **hearing that signal** (through updating the joint public mind)

$$P(a|signal) = \sum_{mind_j} P(mind_j|signal)P(a|mind_j)$$

Bayesian inference formalizes how signals affect the joint public mind

$$P(mind_j|signal) \propto P(signal|mind_j)P(mind_j)$$

And rational planning tells us which actions are rational given a particular mind

$$P(a|mind_j) \propto e^{\beta U(a, mind_j)}$$

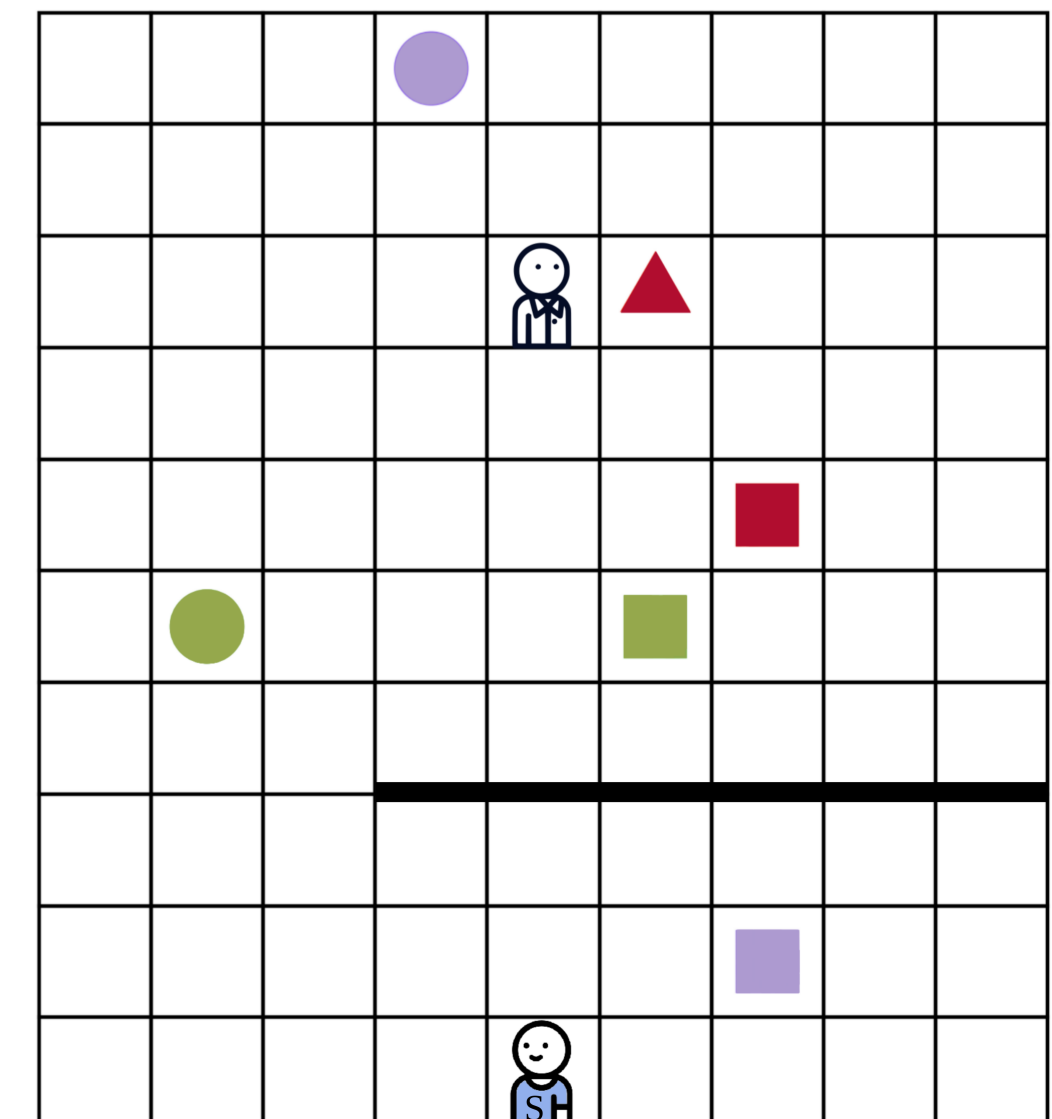
And **how good those actions are** according to the private knowledge in the signaler's mind

$$P(a|mind_t) \propto e^{\beta U(a, mind_t)}$$

## Signaling Task

When and how should I ask for help?

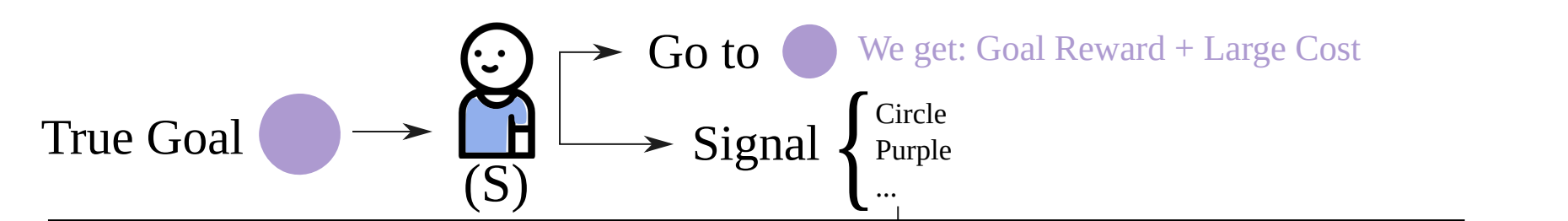
Signals: Green, Red, Purple, Square, Circle, Triangle



**Sampled:**  
- Item #, location, and identity  
- Target item

**Fixed:**  
- Grid size  
- Agent locations  
- Item feature set (3 colors; 3 shapes)

### Trial Walk-through



### Models for Comparison

#### Optimal Central Control (CC)

Best action by joint utility if goal were known to all

#### Joint Utility (JU)

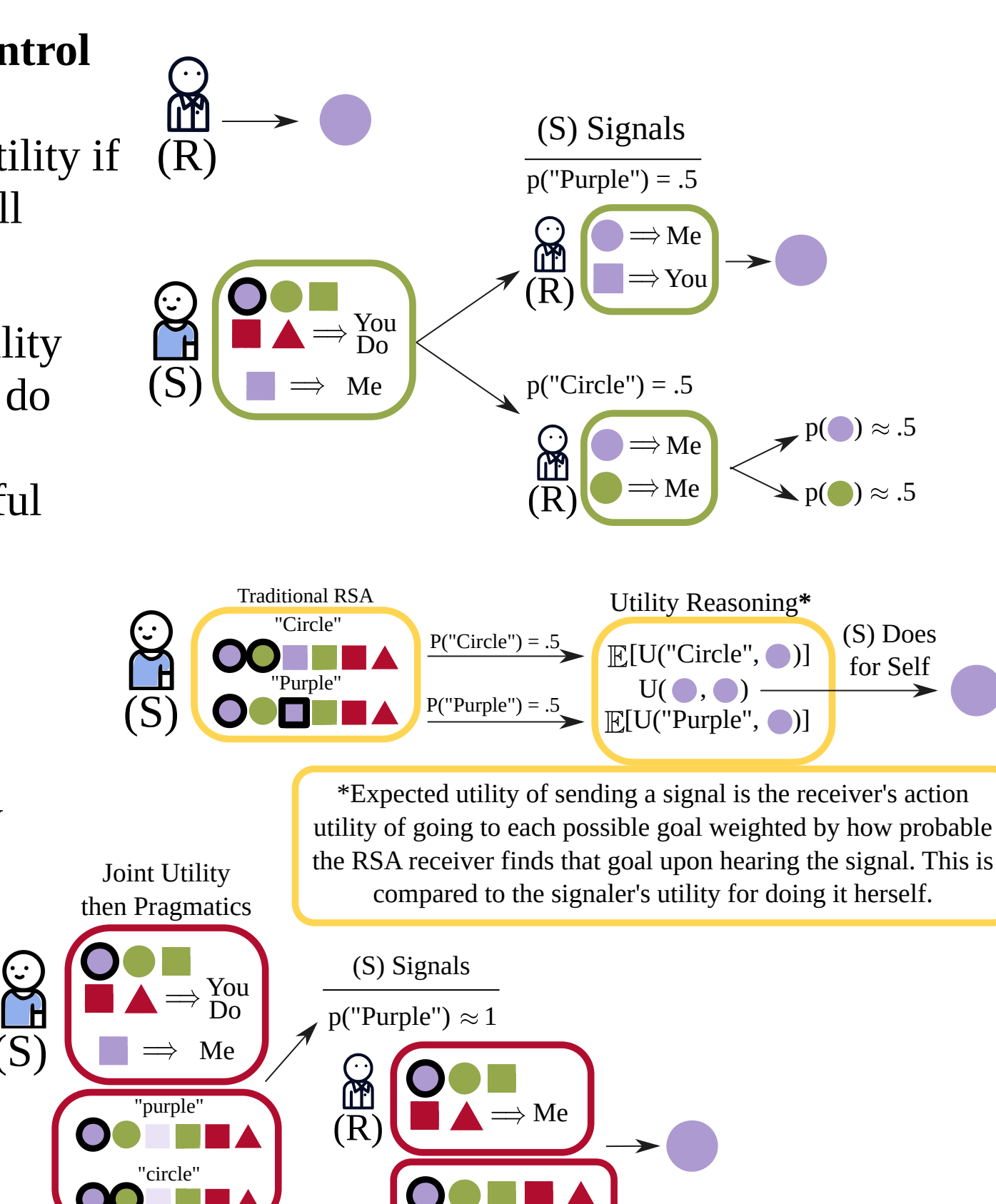
Signaler uses joint utility to decide who should do it; if receiver should, signaler sends a truthful signal uniformly

#### Pragmatics (aRSA)

Traditional RSA for language augmented with individual utility reasoning

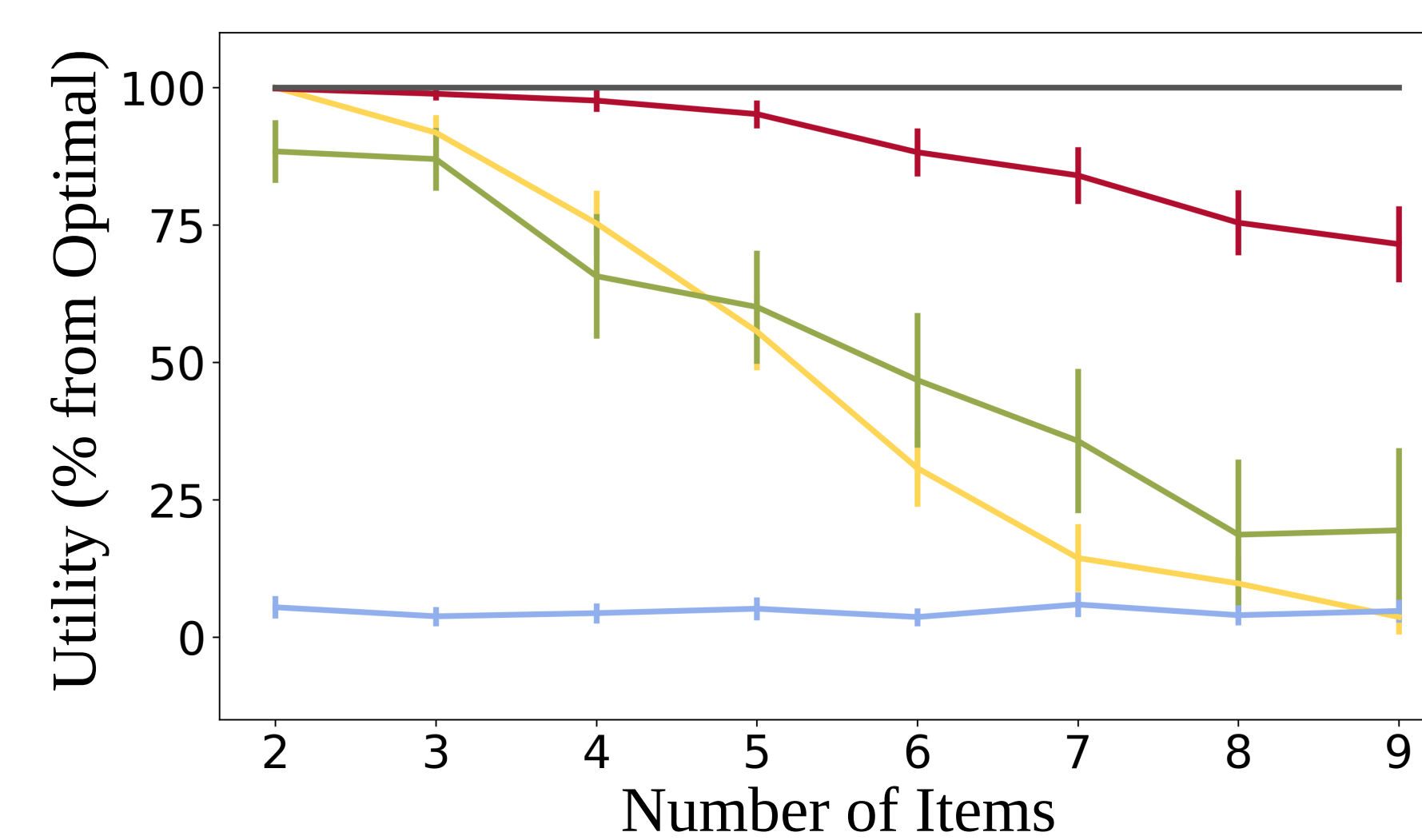
#### Imagined We (IW)

First joint utility reasoning then joint pragmatics



## Signaling Joint Intentions

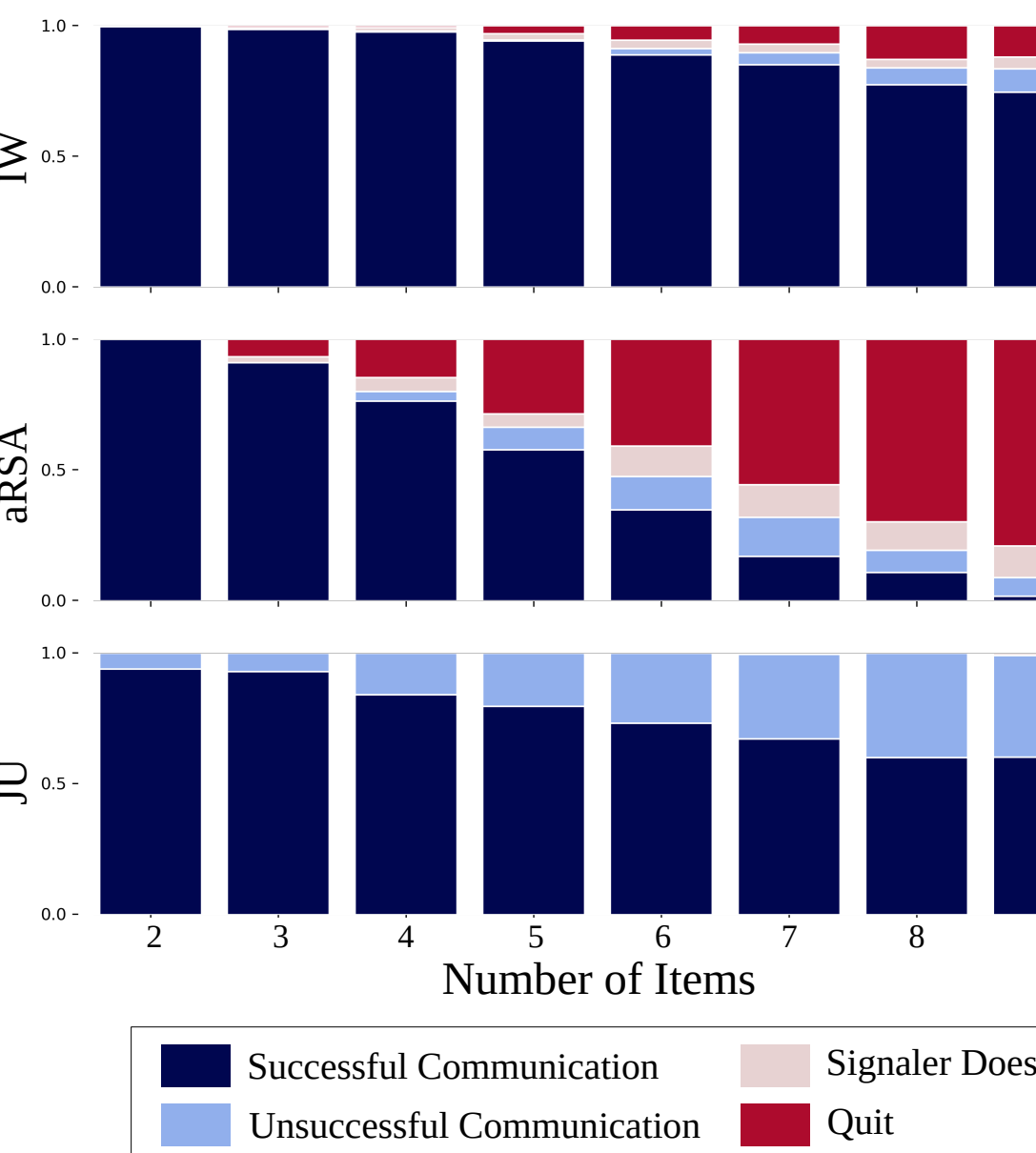
### Task Performance (Utility) When Communication is Optimal



We examine how different models perform in the signaling task under increasing ambiguity (# items).

1. Communication improves performance, except when overloading is at its maximum
2. IW cooperators are able to successfully communicate under high ambiguity; whereas performance in JU and aRSA decay substantially

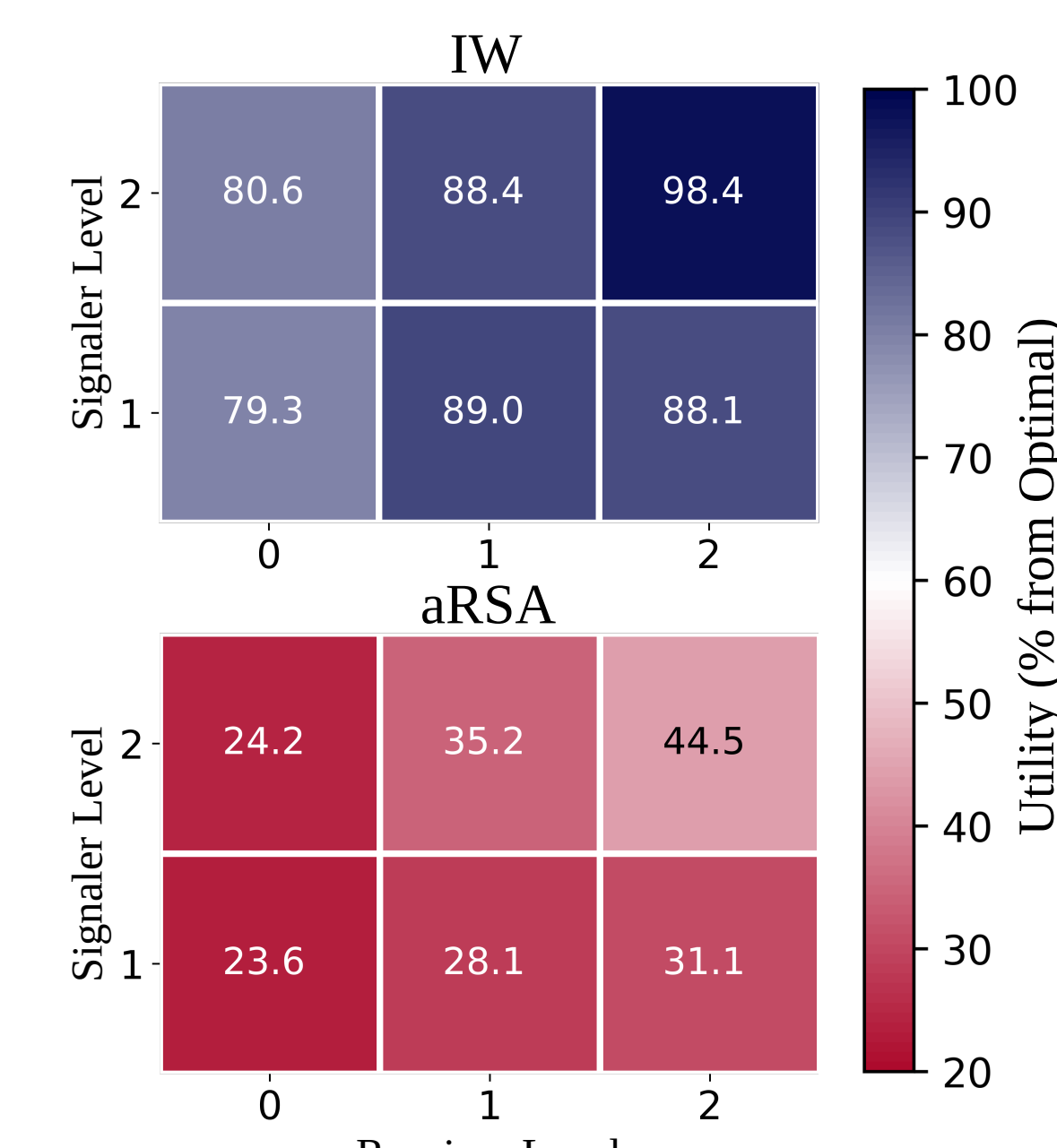
### Breakdown of Model Behavior



Achieved utilities are broken down to demonstrate the ways in which models behave differently.

1. JU always signals, but the receiver is unable to understand
2. aRSA predicts negative utilities and quits instead of communicating

### Utility Achieved Under Differing Recursive Reasoning Levels



When communicators of different reasoning complexities play each other, how does this affect utility?

1. Adding deeper reasoning to communicators improves performance
2. The least complex IW outperforms even the most complex aRSA, indicating deep reasoning may not be needed

## Conclusions

1. Leveraging Multiple Types of Context  
Both joint utility and joint pragmatics add to disambiguating overloaded signals
2. Robust Inferences  
The Imagined We framework outperforms competing models with many items and when highly overloaded
3. Less Reliance on Recursion  
Although recursion helps task performance, the IW does not need deep recursion to perform well because the burden of inference is spread to joint utility and cooperative assumptions

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