

# Word Games

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# Dialogue

- ▶ Communication is rarely unambiguous
  - ▶ Ambiguity resolution through dialogue
  - ▶ Clarification questions
- ▶ Interactive, symmetric reference games
  - ▶ Isolates ambiguity resolution
  - ▶ Both give and request information

# Games

Friends of agent A:

| Name    | School   | Major            | Company |
|---------|----------|------------------|---------|
| Jessica | Columbia | Computer Science | Google  |
| Josh    | Columbia | Linguistics      | Google  |
| ...     | ...      | ...              | ...     |

A: Hi! Most of my friends work for Google

B: do you have anyone who went to columbia?

A: *Hello?*

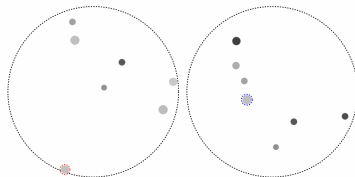
A: I have Jessica a friend of mine

A: and Josh, both went to columbia

B: *or anyone working at apple?*

B: SELECT (Jessica, Columbia, Computer Science, Google)

A: SELECT (Jessica, Columbia, Computer Science, Google)



Human A's view      Human B's view

Human B: three light grey dots in a diagonal line

Human A: i dont have that but i have a black dot neer the top to the right, the only black dot in the circle

Human B: i have two black dots. find something else

Human A: ok i have a light grey dot by itself at the bottom to the left. right on the line

Human B: how big is it

Human A: its one of the bigger ones

Human B: okay just pick it then

Human A: ok

Human B: SELECT blue

Human A: SELECT red

## Mutual Friends and OneCommon

## Issue: Poor neural reasoning

From Mutual Friends: Neural + Human

- ▶ A: Know anyone who likes chess?
- ▶ B: None of my friends like chess.
- ▶ (conversation continues)
- ▶ A: Crocheting?
- ▶ B: None like crocheting.
- ▶ A: Chess?
- ▶ B: None like chess either, haha.

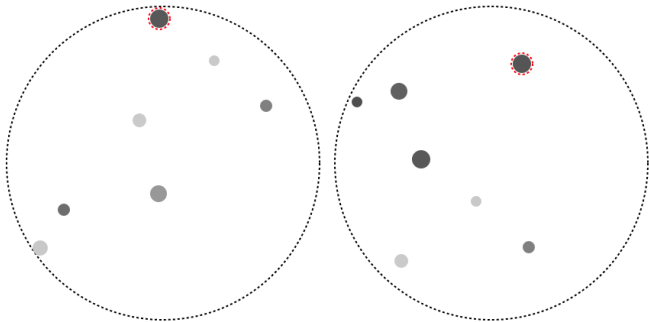
# Issue: Poor neural reasoning



Agent 0: human || 1: human

|   |                                                                                                         |
|---|---------------------------------------------------------------------------------------------------------|
| 0 | 1                                                                                                       |
| 0 | 1 I see a large grey dot with a smaller black dot right below it                                        |
| 1 | 0 is the smaller black dot to the right and below                                                       |
| 2 | 1 No, it is slightly to the left and below                                                              |
| 3 | 0 Might not be the same one. Do you have a lot of dots in a fairly vertical conformation                |
| 4 | 1 I have three dots in a kinda vertical line with different shade of gray but close in size             |
| 5 | 0 Is the middle dot curved to the left or right a little bit?                                           |
| 6 | 1 the middle dot is slightly to the left of the other two and is also a little bit darker than the rest |
| 7 | 0 Yes pick that middle dot                                                                              |

# Issue: Poor neural reasoning



Agent 0: pragmatic\_confidence || 1: human

|   |   |                                                         |
|---|---|---------------------------------------------------------|
|   | 0 | 1                                                       |
| 0 | 0 | i have one large black dot by itself . do you have it ? |
| 1 | 1 | Yes, I do have that.                                    |
| 2 | 0 | let 's pick that one                                    |
| 3 | 1 | ok                                                      |
| 4 | 0 | ok                                                      |

## Issue: Scaling rule-based

- ▶ Rule-based text generation and understanding is somewhat viable for Mutual Friends
  - ▶ Very optimistic selection, but can be tuned
- ▶ Continuous and spatial nature of OneCommon makes writing rules difficult
  - ▶ Size, color, and positions all continuous
  - ▶ Descriptions are relative

## Current approaches: Two extremes

- ▶ Neural encoder-decoder
  - ▶ Encode past interactions with a neural net
  - ▶ Generate what to say with a neural net
  - ▶ Brittle strategy, less brittle language
- ▶ Rule-based
  - ▶ Encode past interactions in a table
  - ▶ Use rules for what to say next
  - ▶ Nonparametric lookup of utterances
  - ▶ Brittle language, less brittle strategy
- ▶ Meet in middle with interpretable planning + neural language



# A dialogue turn

- ▶ Engaging in dialogue requires
  - ▶ Inference: What do I know? How do I represent it?
  - ▶ Planning: What should I do and say?
- ▶ Formulate as model-based optimization
  - ▶ Plan what to say through a simple model of our partner
  - ▶ Model of partner conditions on past information

# Problem setup

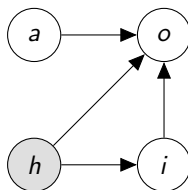
- ▶ Goal: Mutually select the same item as partner
  - ▶ Row in knowledge base, dot
  - ▶ Coordinate through dialogue
- ▶ Given history  $h$ , we need to choose an action  $a$  by optimizing value

$$\max_a V(h, a)$$

- ▶ Value  $V$  = information gain + utterance + pragmatic cost
  - ▶ IG: Entropy reduction of item selection probability
  - ▶ Utterance cost: Can't send a full paragraph
  - ▶ Pragmatic cost: Want utterance to be accurate
- ▶ Represent  $h, a$  using attributes
  - ▶ Columns of knowledge base, spatial configuration of dots

# Information Gain

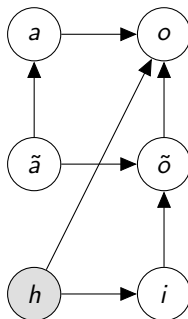
- ▶ A good action should move us closer to game success
- ▶ Game success depends on our knowledge of our partner's context
- ▶ Requires
  - ▶ Belief distribution over selection item given history  $p(i | h)$
  - ▶ Partner response model  $p(o | h, a, i)$
- ▶ Represent a turn as



- ▶ Language and planning coupled

# Decoupling language and planning

- ▶ Compress actions  $a$  and observations  $o$  into language and abstract representations  $\tilde{a}, \tilde{o}$ 
  - ▶ Language is high dimensional, redundant, and inefficient for planning
- ▶ Represent a turn as



- ▶ Abstract observation  $\tilde{o} \perp\!\!\!\perp h \mid \tilde{a}, i$

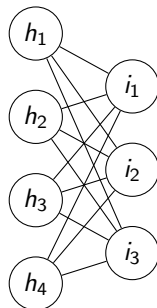
# State and belief: Representation

- ▶ History: whether attributes have been confirmed  $h \in \{0, 1\}^N$
- ▶ Items:  $i \in [M], N \gg M$
- ▶ Logistic regression with attributes as features

$$p(i | h) = \frac{\exp(\sum_n \psi(h_n, i))}{\sum_{i'} \exp(\sum_n \psi(h_n, i'))}$$

$$\psi(h_n, i) = W_{ni} 1(h_n(i))$$

- ▶ Generate per-game  $W = f(\text{context})$  from neural network
  - ▶ Many correlated features
  - ▶ How to (conditionally) sparsify?
- ▶ Dialogue = online feature selection



# Attributes

- ▶ Mutual Friends
  - ▶ Combinations of columns of knowledge base
  - ▶ Name, major, company
- ▶ OneCommon
  - ▶ Which dots are mentioned
  - ▶ Need to learn lower-level attributes
- ▶ Numerical reasoning?

# Experiments

- ▶ Mutual Friends
  - ▶ Augment rule-based (prior work) to optimize info gain
  - ▶ After OneCommon: Add neural on top
- ▶ OneCommon
  - ▶ Use attributes = raw mention configurations
    - ▶ Everything already in system, except for belief / info gain
  - ▶ Learn latent refinement on top of mention configurations
    - ▶ How to deal with redundancy? (i.e. correlation between features)

End



# Concerns

- ▶ Would a large LM solve all of this?
  - ▶ Fine tune on small onecommon dataset, are there still repeats?
  - ▶ Unlikely to solve strategy / over optimism

End

# Value: Information Gain

- ▶ drop slide
- ▶ Picture would be much better here...
- ▶ Value = expected information gain

$$IG(h, a) = H(i | h) - \mathbb{E}_{p(o|h,a)} [H(i | h, a, o)]$$
$$\mathbb{E}_{p(o|h,a)} [H(i | h, a)] = \sum_o \sum_{i'} p(o | h, a, i) p(i | h) H(i | h, a, o)$$

- ▶ Equivalent to minimizing expected uncertainty after receiving a response
- ▶ Cite Yu et al, White et al

# Citations I