Word Games

J Chiu

March 15, 2022

Dialogue and information gathering

- Resolve ambiguity and coordinate through dialogue
- OneCommon: Interactive, symmetric reference game
 - Isolates info gathering (and coordination)
 - Environment (dots) are completely static
 - Dynamism comes from dialogue only
- 20 questions with symmetric information constraints

Previous SotA

- Purely supervised
- Uncalibrated beliefs: overconfidence
 - Pushes for to select a dot that will not work
- Research goal: Improve purely supervised models via model-based planning

Fixing strategy with planning

- Prior: Fully supervised neural encoder-decoder
 - Encode past interactions with a neural net
 - Generate what to say with a neural net
 - Brittle strategy, less brittle language
- Next: Model-based planning
 - Choose what to say by imagining how partner would respond
 - Say utterance with best expected outcome
 - Potentially stronger player than expert demonstrations

Challenges in model-based planning

- Partner modeling is hard
 - ► Variable amount of information
 - Random strategies
- Multi-turn planning
 - Accuracy of planning depends greatly on the partner model
 - Errors from the partner model will compound over time
- Single-turn planning
 - Removes compounding errors
 - Optimize a dialogue progress heuristic: uncertainty reduction
 - Requires belief
- Use what dots partner also sees as belief

Planning

▶ Plan by imaging partner response

$$\max_{u} \mathbb{E}_{p(r|h,u)} \left[\mathsf{Utility}(h,u,r) \right]$$

- ightharpoonup Utterance u, response r, history h
- Utility should approximate dialogue progress
 - Goal of dialogue is information gathering and coordination
 - Focus on information gathering
- Utility a function of belief

Planning with Belief

- ▶ Introduce belief state p(s | h)
 - ▶ State *s* is what dots partner can also see
- ► Incorporate belief in planning

$$\max_{u} \mathbb{E}_{p(r|h,u,s)p(s|h)} \left[\mathsf{Certainty}(p(s\mid h,u,r)) \right]$$

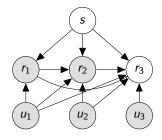
Obtain

$$p(s \mid h, u, r) = \frac{p(r \mid h, u, s)p(s \mid h)}{\sum_{s} p(r \mid h, u, s)p(s \mid h)}$$

▶ Response model p(r | h, u, s) now has more conditioning

Partner model

- ▶ Static latent state *s*: which dots do they also see
 - Alternative: actual field of view
- ▶ Uniform prior p(s) over 7 choose 4 dots partner also sees
- ▶ Partner response model $p(r_t \mid u_{1:t}, r_{1:t-1}, s)$
 - ▶ Pick *s* that is observed during training
 - ► History $h_t = (u_{1:t-1}, r_{1:t-1})$



Belief update

- Interaction history h_t contains all previous utterances and responses
- ► Given an initial belief $p(s | h_t)$ + next utterance/response, obtain next belief via

$$p(s \mid h_t, a_t, o_t) \propto \underbrace{p(o_t \mid h_t, a_t, s)}_{\text{response model}} p(s \mid h_t)$$

Belief calibration depends on accuracy of observation model

Response model

- Example exchange
 - ► Action: Do you see a red dot?
 - Observation: No, but I see a blue one.
- Utterances are multifaceted
 - Responses contain more information than asked
 - New information injected by partner not constrained
 - Very difficult to model new information

Incorporating informativity

- Add partner-dependent informativity i
- Determines partner willingness to give more information

$$\max_{u} \mathbb{E}_{p(r|h,u,s,i)p(s|h)p(i)} \left[\text{Certainty}(p(s \mid h, u, r)) \right]$$

- More information improves game-play
- ▶ Introduce uncertainty set over $i \in [low, high]$
- Optimize worst-caset informativity

Response extraction

- ▶ Heuristic: Use repeated mentions from response
 - Do you see a red dot? Yes, the one next to the blue one?
- ► Generalization: TBD
- ► Recap: we have
 - ▶ Belief over shared dots $p(y \mid h)$
 - ▶ Observation model p(o | h, a, y)
 - ightharpoonup Update $p(y \mid h, a, o)$
 - Reduced to assymmetric case by extracting response only
- Next: Single-turn planning

Planning: Use prior work in assymmetric setting

Given history h, we need to chose an action a by optimizing heuristic utility

$$\max_{a} U(h, a)$$

- lacktriangle Utility U= information gain utterance pragmatic cost
 - ► IG: Reduce uncertainty
 - Utterance cost: Can't send a full paragraph
 - Pragmatic cost: Want utterance to be accurate
- Ideally would estimate and optimize future reward directly
 - Heuristic approximation of future reward U
 - Limited-horizon planning to minimize impact of model error

Expected information gain

 Maximizing expected information gain equivalent to minimizing uncertainty

$$\min_{a} \sum_{o} \sum_{y} \underbrace{p(o \mid h, a, y)}_{\text{observation model}} \underbrace{p(y \mid h)}_{\text{belief}} \text{Uncertainty} \underbrace{(p(y \mid h, a, o))}_{\text{new belief}})$$

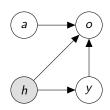
Summary

- ► Goal: Extend methods from 20 questions to symmetric, language setting
- Extract relevant information from partner utterances
- ▶ Use explicit belief state + single-turn planning heuristic

End

Information Gain

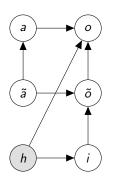
- A good action should decrease uncertainty
- Requires
 - ▶ Belief distribution over selection item given history $p(y \mid h)$
 - Partner response model p(o | h, a, y)
- Represent a turn as



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



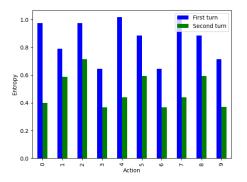
▶ Abstract observation $\tilde{o} \perp \!\!\!\perp h \mid \tilde{a}, i$

Experiments

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

Information gain issues

- Best info gain could be to ask the same question twice
- Usual fix: Limit to asking once only
- Would be nice to have a principled way to deal with correlated features though



Second turn after taking action with lowest entropy

Related work: 20 questions

- ► Padmakumar and Mooney (2020)
 - ► Attribute-based classification (string heuristic to map to description) + activate learning about attributes
 - Info gain (on top of binary unweighted logistic regression) as feature for RL policy
- Yu et al. (2019)
 - Question-based classification (attributes)
 - Learn weights of features
 - Do not consider feature correlations
- More interesting language, symmetric setting
- Learn weights, account for correlation
- Symmetry, deal with unexpected 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 optimistism

End

Expected Information Gain

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

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

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

Citations I

- Padmakumar, A. and Mooney, R. J. (2020). Dialog policy learning for joint clarification and active learning queries. *CoRR*, abs/2006.05456.
- Yu, L., Chen, H., Wang, S. I., Artzi, Y., and Lei, T. (2019). Interactive classification by asking informative questions. *CoRR*, abs/1911.03598.

Games

Friends of agent A:

Name	School	Major	Company
Jessica Josh	Columbia Columbia	Computer Science Linguistics	Google Google

A: Hi! Most of my friends work for Google

B: do you have anyone who went to columbia?

A. Hello?

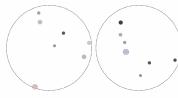
A: I have Jessica a friend of mine

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.

Sample of prior work in model-based planning

- ➤ 20 questions (Yu et al., 2019; Padmakumar and Mooney, 2020)
 - Sym: Assymmetric questioner + answerer
 - ► Turns: Multi-turn game
 - Lang: Closed class answers (observations)
 - Heur: Expected info gain heuristic
- ► EVPI (??)
 - ► Sym: Assymmetric questioner + answerer
 - ► Turns: No interaction, single turn game
 - Lang: Open
 - Heur: Expected utility heuristic
- RSA reference game (?)
 - Sym: Symmetric
 - ► Turns: Multi-turn game
 - Lang: Symbolic language
 - Heur: Bounded depth search

Conditioning in partner modeling

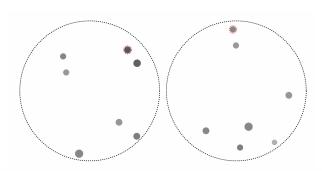
- Assuming conditional independence $p(o \mid h, a, y) = p(o \mid a, y)$ is harmful
- ▶ If you ask the same question twice, your belief changes both times!
 - ▶ $p(\text{yes} \mid h = \emptyset, a = \text{red dot}?, y)$ can vary depending on the latent y
 - $p(\text{yes} \mid h = (\text{red dot}?, \text{yes}), a = \text{red dot}?, y) = 1$, since we just asked!
- 'Questions with correlated answers' and deficient observation model lead to uncalibrated beliefs, and therefore poor strategy
- Contribution: relax independence assumption
 - Let past obs vote on current one (weighted by action similarity)
 - Probably solved by Transformers¹

¹Copy attention, depends on amount of data

Example dialogue 1: Overconfidence

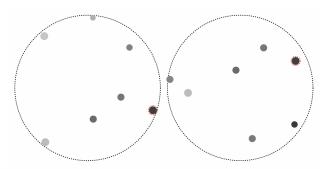
i have a large black dot, it 's not the same size, but it is not the darkest Mine is the darkest i see it . i think it is the one . I have multiple dark dots so I am not sure i have a large black dot, it is in the center I have a large dot in the lower part with a dark smaller dot at seven o'clock is the black dot the darkest? if so click it I have three the same darkness pick the lightest one Ok but our odds are not good. pick the light one

Example dialogue 2: Overconfidence



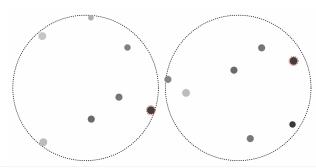
- 0 i have two dark dots, one on top and slightly smaller than the other
- 1 i see it. pick the top one?
- 0 ok

Example dialogue 3: Good humans



- 1 I have a large black dot by itself
- I see a large, very dark dot on the edge of my screen (so I won't be able to see anything to its right). Can you see anything on the left of your large black dot?
- 1 Yes, my large dark dot is on the edge of the right side
- Ok, to the left of the dark dot, and slightly above it, do you see a slightly-smaller, slightly lighter dot?
- 1 yes

Example dialogue 3: Good humans



- and then far above (and a bit to the right of) that lighter one, do you see a slightly smaller, identically colored dot?
- 1 No, the first lighter dot is the closest dot to the top of mine
- 0 Okay. what do you see to the left of that lighter dot?
- 1 A slightly darker dot that is below it just a bit
- 0 ok, I think we're in the same place. Let's click that original, blackest dot
- 1 Okay sounds good