Word Games

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Natural Language Interaction

- Interaction (through language) is important
 - Cannot fully automate every task, i.e. task-oriented or information seeking dialogues require human input
 - Must handle diverse non-expert human input, although input may map to a low-dimensional manifold
 - ► High levels of ambiguity must be resolved via interaction
- Interaction (through language) is hard
 - Human input is expensive, so supervision is limited
 - In order to make certain problem aspects tractable, must make sacrifices in other areas (toy domain = out of distribution for pretrained models)
- What are the main challenges in interaction, and what are the tradeoffs of different approaches?

Types of Dialogue Games

- ► Task-oriented: Wizard of Oz (WoZ)
 - ► Tseng et al. (2019): Wizard obtains task from human then executes it.
- Deliberation / reference / signal
 - ▶ Udagawa and Aizawa (2019): Visual reference game with latent translated views. Each player gets a different petri dish view of the same underlying game board, and players must select the same object on the board.
- Information seeking / inquiry
 - Yu et al. (2019): WoZ-style answer providing where asker does not know exact question. Latent true question (to all), WoZ must answer
- ► Persuasion / negotiation
 - ▶ Lewis et al. (2017): Negotiation over an observed set of item with latent utilities for each agent.

Types of Dialogue Games

In all cases, the game can be (indirectly) solved by resolving a latent variable

- When is this tractable, and why do no new methods do this?
- ▶ New (ie basically all) methods rely on supervision
- ▶ If they do not, it is because the game has a trivial solution

Types of Dialogue Games: Latent goals and strategy

What are the latent variables in each type?

- ► Task-oriented
 - Latent task slots
- ► Deliberation / reference / signal
 - Varies per game
- ► Information seeking / inquiry
 - ► Infer true question, find answer
- Persuasion / negotiation
 - ► Infer utilities, exploit

Types of Dialogue Games: Latent goals and strategy

- ► Tasks must be interesting enough so that latent quantities cannot be inferred with a single utterance, reducing them to single turn games
 - ► High degree of ambiguity / distractors or large number of slots to fill (combinatorial)
- Break down latent quantities and use heuristics to make assumptions on structure
 - ► For example, choosing an ordering of WoZ slots: When choosing a restaurant, first figure out time, then cuisine, and finally price
 - Will likely remain task-specific
- ▶ What other parts can we learn?

Belief State Tracking

- ► The incremental inference procedure is known as belief state tracking (BST)
- ► Local semantics are aggregated into belief state, which informs high-level strategic decisions
- Seems difficult to learn language, high level strategy, belief state updating, and low level parsing at the same time
- ► Ablate how structure influences each of these

Language Games

- ► Games offer a testbed for the development of methods
 - Allow designers to control difficulty and simplicity
- ► Allowing interaction through language increases the population of players

3 Challenges in Language Games

- 1. Game design
 - Simplification of or step towards a realistic task
 - Must be fun or useful
- 2. Meaning representations
- 3. Search
 - A
- 4. Modeling other agents
 - A

Axes of Tractability

- ► Horizon length
 - **▶** B
- ► Action space
 - ▶ B
- A
 - ▶ B

Meaning reps

- ► In full generality, this problem is often encountered in hierarchical RL
 - Less bleak in the language gamesetting
 - Games are often very simple and can be constrained to small horizons, for example He He engineered a parser and policy that basically solves the negotiation task
- Many text-specific meaning representations (MR) to choose from
 - Many are too complex
 - Can we leverage existing MRs to learn a minimal task-specific representation that balances utility and expressivity?

Learning to plan



Contributions

- Under a unified Bayesian game perspective of dialog games, present formulations for different classes of dialogs: signaling, negotation?
- Provide pipelined variational Bayesian framework for learning to play dialog games from offline data, with and without granular annotations
- ► Good results

Tasks

- ► Task-oriented dialogue: Multi-WoZ
- ► Negotation: Deal-or-no-deal
- ► Reference: OneCommon
- ► Information-Seeking: Birds

Citations I

- Lewis, M., Yarats, D., Dauphin, Y. N., Parikh, D., and Batra, D. (2017). Deal or no deal? end-to-end learning for negotiation dialogues. *CoRR*, abs/1706.05125.
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