

The Dialog State Tracking Challenge and Bayesian approach

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22.06.2016



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Outline

- Motivation
- Dialog State Tracking Problem
- Bayesian Approach
- Conclusion
- References

Motivation

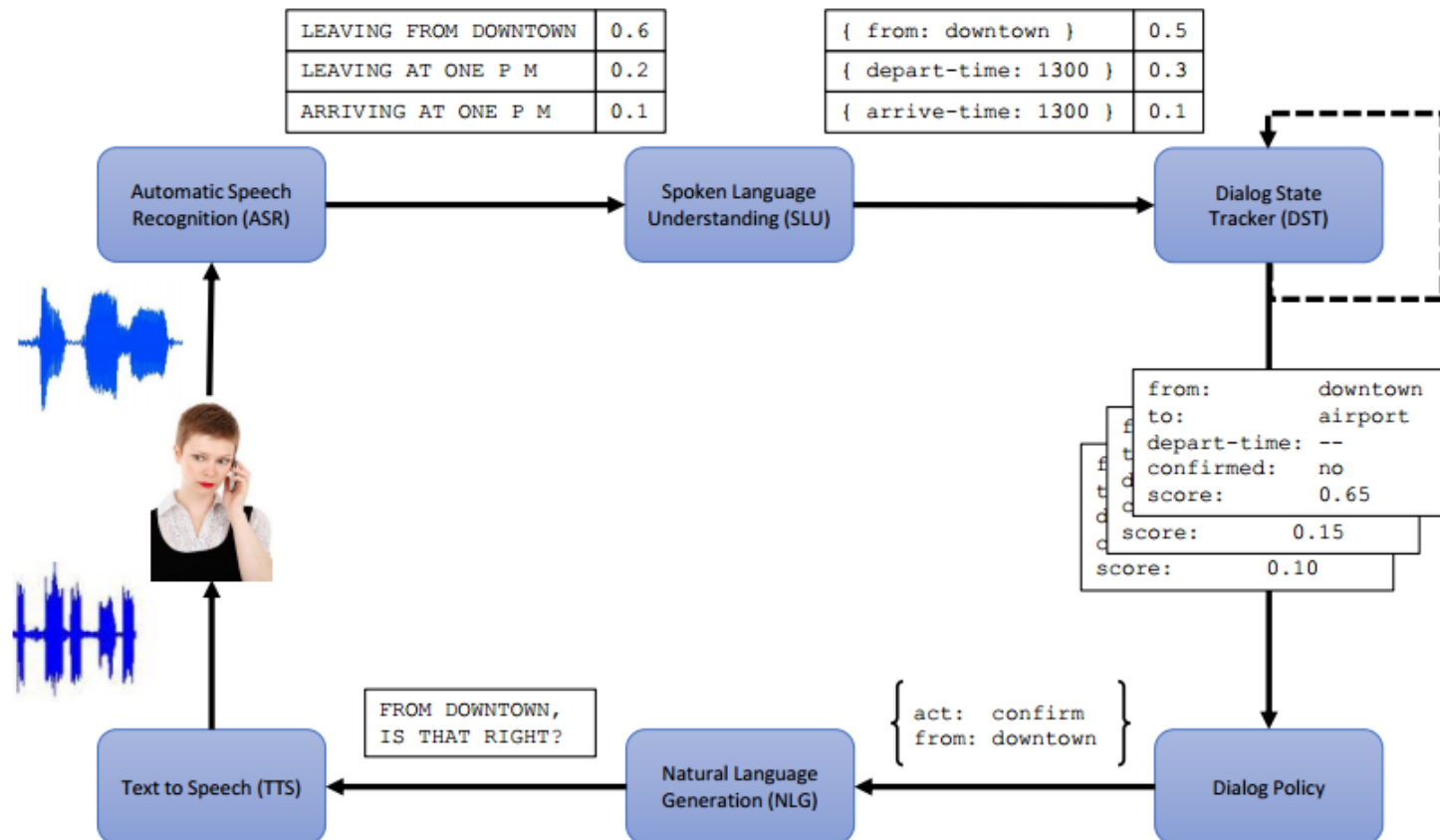
- From agent's perspective:
 - **What do we have so far?**
(common knowledge)
 - **What do the user need?**
(requested slots)
 - **What do I need from the user?**
(informable slots)
 - **What should I say?**
(conversation acts)

Motivation (2)

- Measuring the accuracy of agent's understanding
 - No common evaluation metrics
 - Different domain
 - Different techniques
 - Different (learning) data
- One state tracker for multiple strategies, multiple domain?

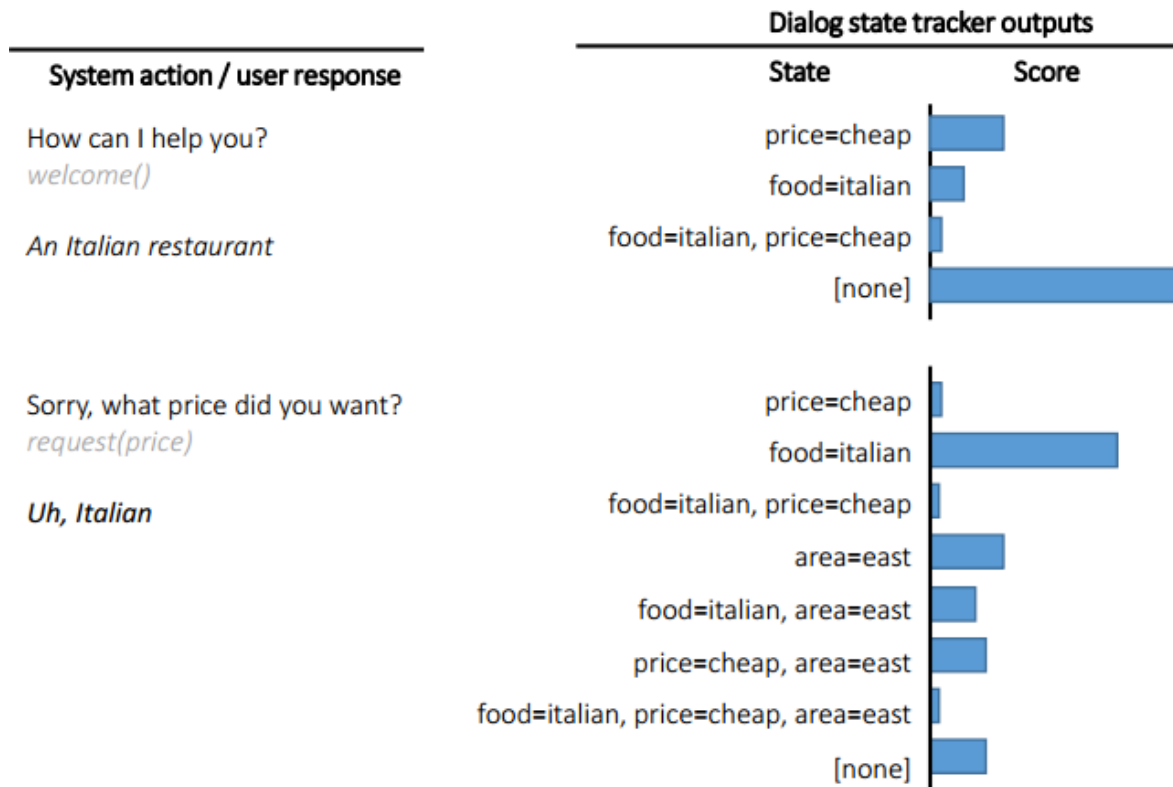
Dialog State Tracking Problem

- State of Dialog



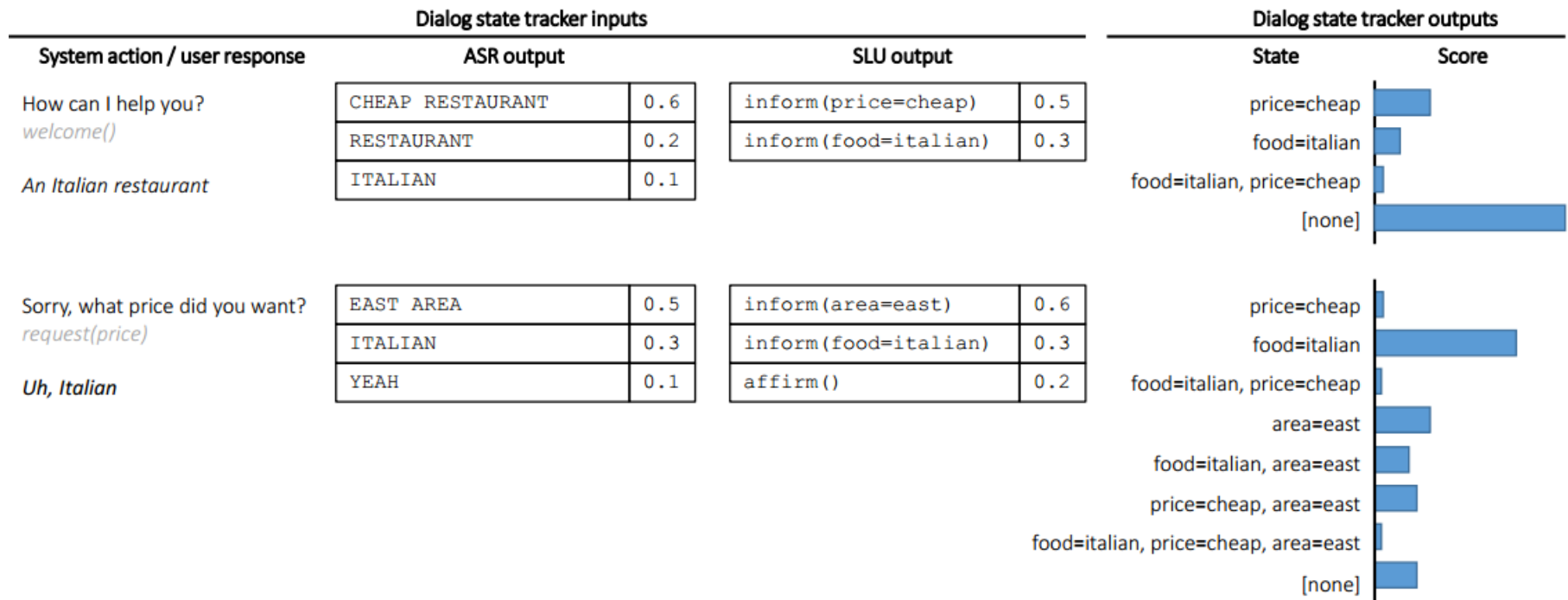
Dialog State Tracking Problem (2)

- Tracker's output



Dialog State Tracking Problem (3)

- Tracker's input (end-to-end system)



The DSTC challenge

- Different trackers – similar train and test data
 - Mixed strategies
 - Shared ontology?
- Baseline
 - Admissible heuristic
 - Check for improvement
 - Incorporate past solutions
- Evaluation metrics
 - Schedule – when to measure.
 - Accuracy – number of correct turns
 - L2 – vector distance with one-hot encoding
 - AvgP – average score

Bayesian approach

- How to build statistical frameworks?
- How to estimate Bayesian networks' parameters (tracker's experience)?
- Tractability?

Bayesian approach (2)

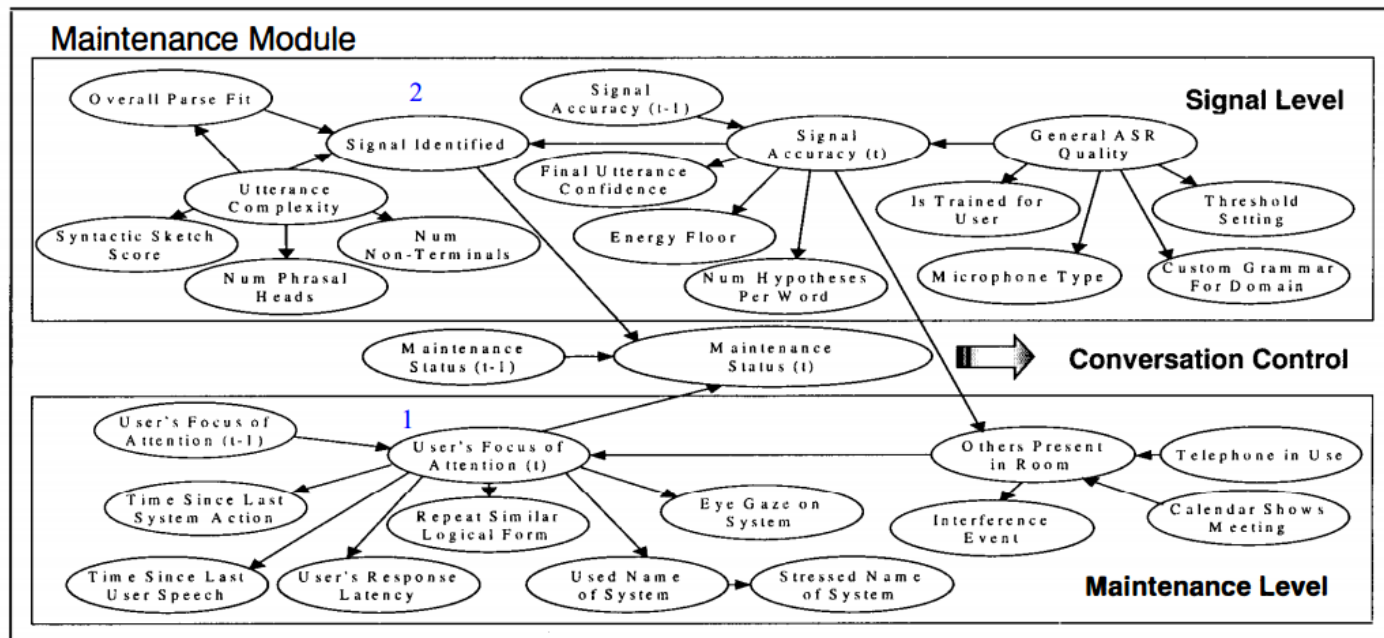
- Basic statistical framework (Williams et. al, 2007)

$$b'(s') = k \cdot p(o'|s', a) \sum_{s \in S} p(s'|a, s) b(s).$$

- Where
 - s and s' are dialog state
 - o' is agent's observations (ASR and SLU output)
 - a is agent's action (speech act)
 - k is normalization constant
 - b(s) and b'(s'): state distribution before and after action a

Bayesian approach (2)

- Tracker's experiences:
 - Hand-crafted nodes and edges (Paek et. al, 2000)



- Reinforcement learning: Optimal Policy (Williams et. al, 2007)
 - Value Iteration

Bayesian approach (3)

- Tracker's experiences (cont):
 - States with continuous values and/or search methods
 - Particle filters: Continuous -> Discrete (Williams 2007)
 - Beam search: M-best most likely states
- Tractability
 - Training phase + Value iteration approach: quadratic in number of states and number of actions (per iteration)
 - End-to-end system: quadratic in number of states

Bayesian approach (4)

- Hand-crafted nodes and rewards
 - Infeasible in complex domain
- Only works under certain assumptions
 - Independence
 - Short-term dependence (dialog history)

Conclusion

- Trackers are crucial in dialog systems
 - What the agents know
 - What the agents should do
- The DSTC challenge
 - Common testing framework
- Bayesian approach
 - Statistical framework -> room for tracking methods
 - Works under certain assumptions (Markov, short-term dependence, small state space)

Adaptation to

- DSTC2: user goal changes
- DSTC3: unseen concepts in test data

The End

Thank you for your attention.
Any question?

Literature:

- Williams, Raux and Henderson. The DSTC Series: A review. *Dialog and Discourse*, 2007
- Williams and Young. Partially observable Markov decision processes for spoken dialog system. *Computer Speech and Language*, 21(2):393-422, 2007a
- Paek and Horvitz. Conversation as Action Under Uncertainty. *UNCERTAINTY IN ARTIFICIAL INTELLIGENCE PROCEEDINGS*, 2000
- Williams. Using Particle filters to track dialogue state. In *Proc IEEE Workshop on Automatic Speech Recognition and Understanding (ASRU), Kyoto, Japan, 2007*