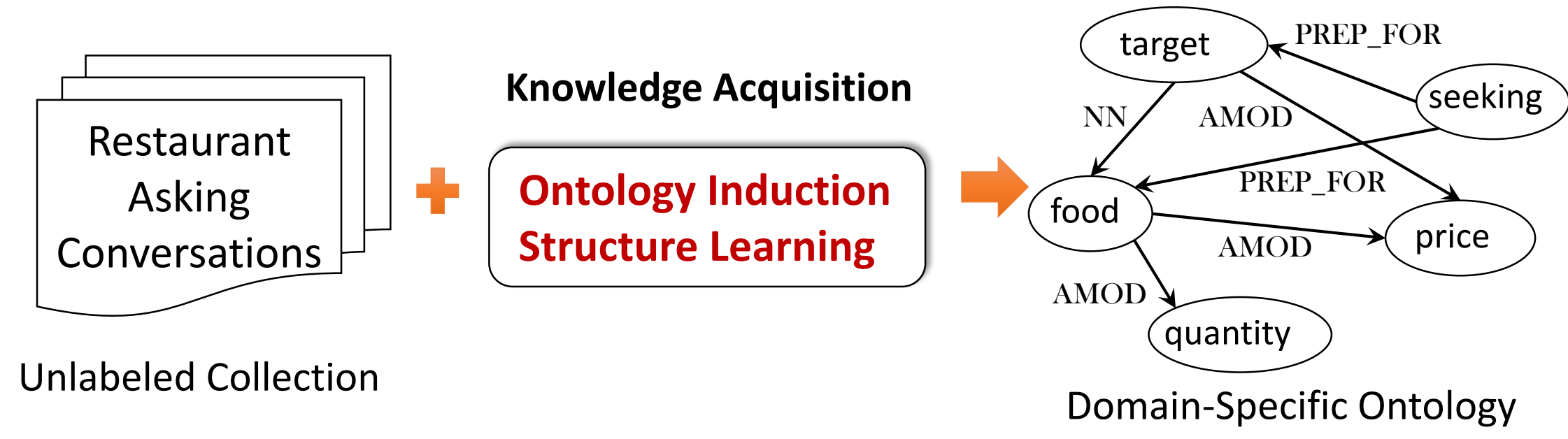


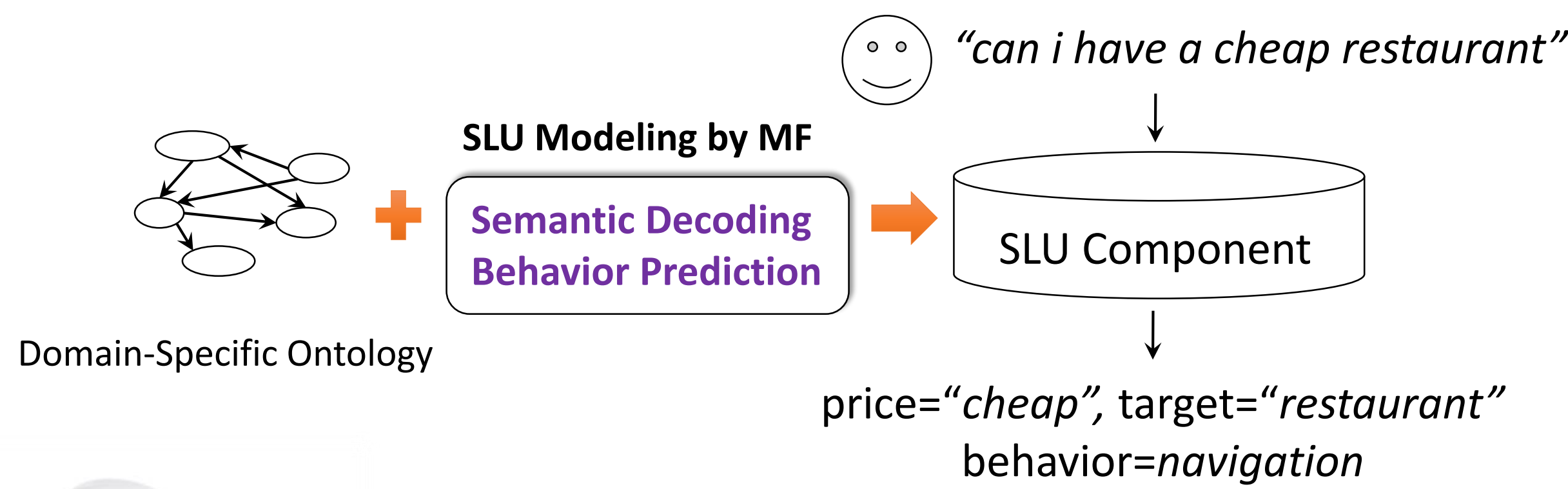


Goal

1) Given unlabeled conversations, how can a system automatically induce and organize domain-specific concepts?



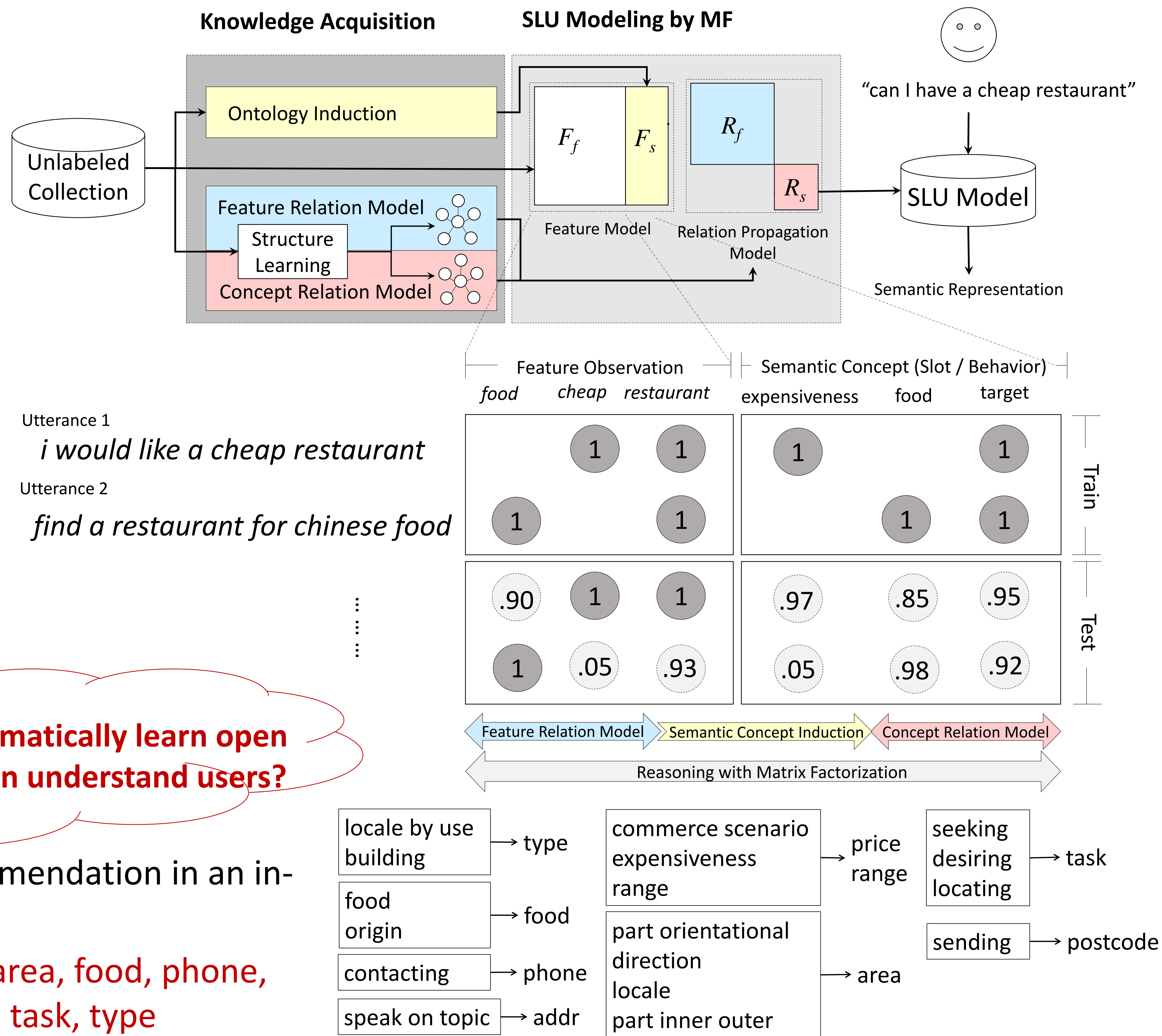
2) With the automatically acquired knowledge, how can a system understand utterances?



Can a dialogue system automatically learn open domain knowledge and then understand users?

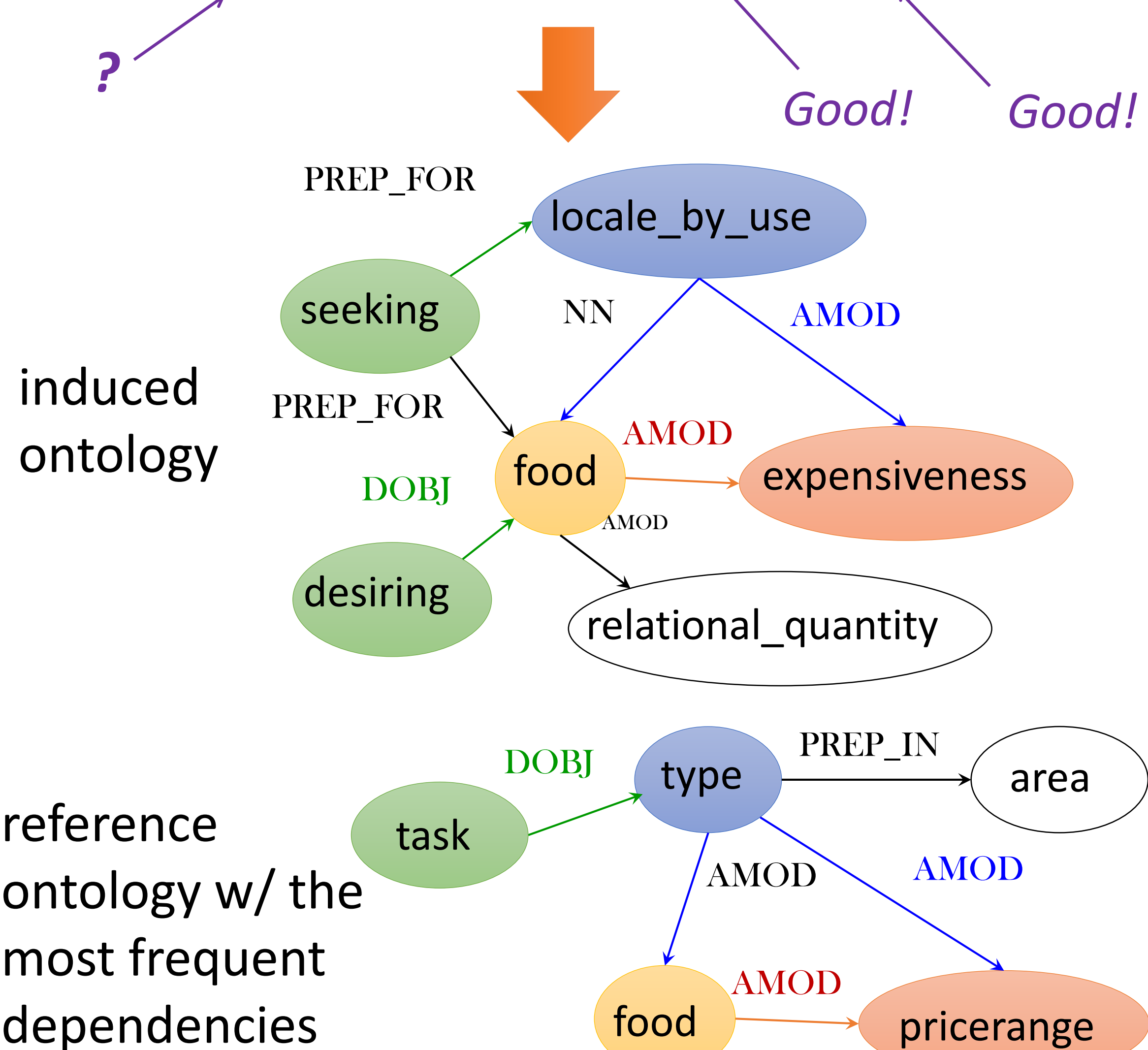
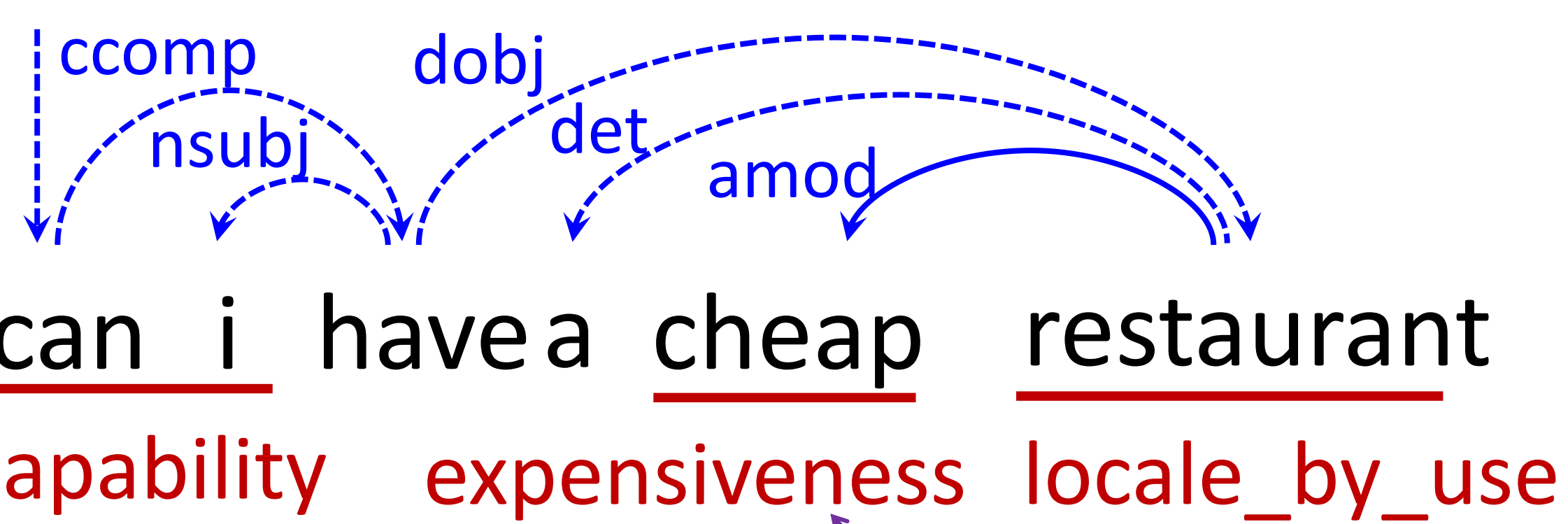
- Domain: restaurant recommendation in an in-car setting (WER = 37%)
 - Dialogue slots: **addr, area, food, phone, postcode, pricerange, task, type**

Framework



Knowledge Acquisition

- **Ontology Induction** (Chen et al., 2013 & 2014)
 - Frame-semantic parsing on ASR results (Das et al., 2013)
 - frame → slot candidate
 - lexical unit → slot filler
- **Structure Learning** (Chen et al., 2015a)
 - Typed syntactic dependencies on ASR
 - <https://github.com/yvchen/MRRW/>



SLU Modeling by Matrix Factorization

- **Semantic Decoding** (Chen et al., 2015b)
 - concept → semantic slot
- **Behavior Prediction**
 - concept → user behavior
- ❖ **1st Issue: How to induce domain-specific concepts?**
 - Relation Propagation Model
 - Feature Knowledge Graph
 - Concept Knowledge Graph
 - ✓ Assumption: The domain-specific features/concepts have more dependency to each other.

Relation matrices allow each node to propagate scores to its neighbors in the knowledge graph, so that domain-specific features/concepts have higher scores during training.

❖ **2nd Issue: Hidden semantics cannot be observed but may benefit understanding performance.**

- Matrix Factorization (MF)
 - Model implicit feedback $f^+ = \langle u, x^+ \rangle$, $f^- = \langle u, x^- \rangle \Rightarrow p(f^+) > p(f^-)$
 - $$p(M_{u,x} = 1 | \theta_{u,x}) = \sigma(\theta_{u,x}) = \frac{1}{1 + \exp(-\theta_{u,x})}$$
 - Objective: $\sum_{f^+ \in O} \sum_{f^- \notin O} \ln \sigma(\theta_{f^+} - \theta_{f^-})$

