Guided K-best Selection for Semantic Parsing Annotation

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* Equal contribution. Work performed during an internship at Microsoft Semantic Machines.

Semantic Parsing

User utterance

When's the lecture scheduled for in May?



Canonical utterance

start time of find event called something like "lecture" during May

Meaning representation

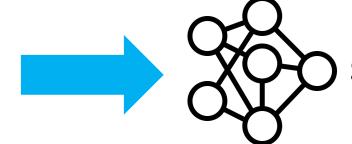
SCFG

```
(Yield :output (:start (singleton (:results
    (FindEventWrapperWithDefaults :constraint
  (EventDuringRange :event (Constraint[Event]
    :subject (?~= #(String "lecture"))) :range
(FullMonthofMonth :month #(Month "MAY")))))))))
```

Semantic Parsing Annotation via K-best

User utterance

When's the lecture scheduled for in May?



SCFG

Low-Res (1k instances)
Semantic Parsing Model

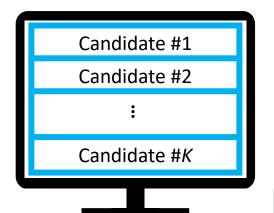


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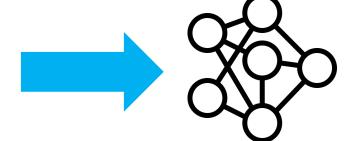




Semantic Parsing Annotation via Guided K-best

User utterance

When's the lecture scheduled for in May?



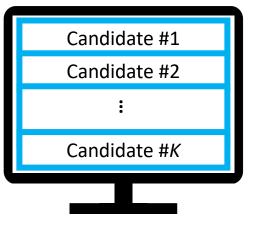
Low-Res (1k instances)
Semantic Parsing Model



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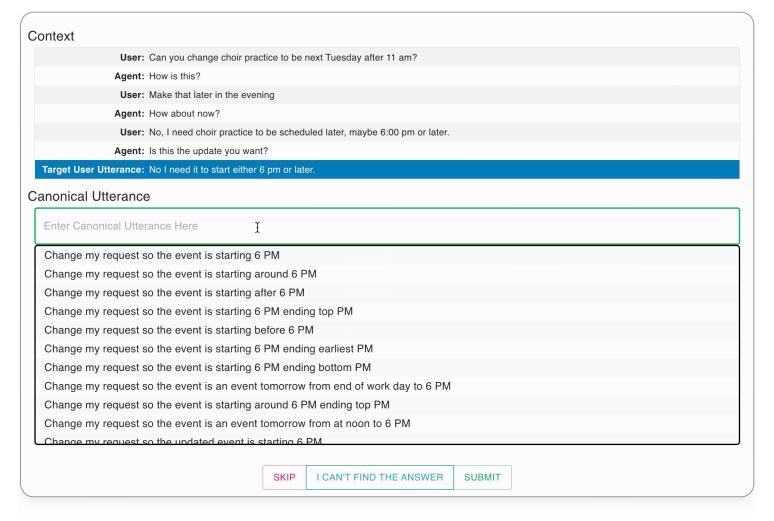






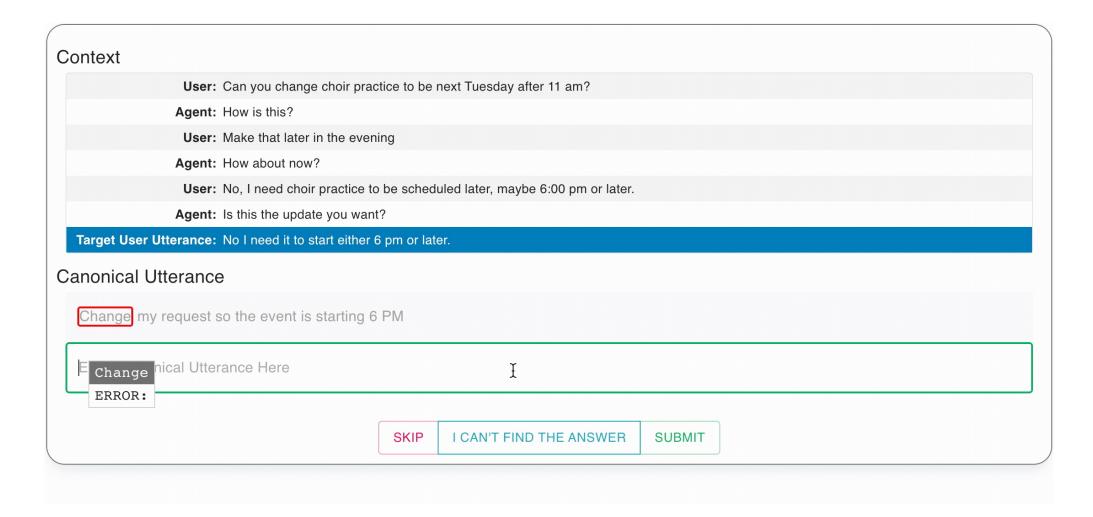
When *K* gets large, can we **guide** annotators to select the correct parse both **fast** and **accurately**?

Annotation Interface: Scroll

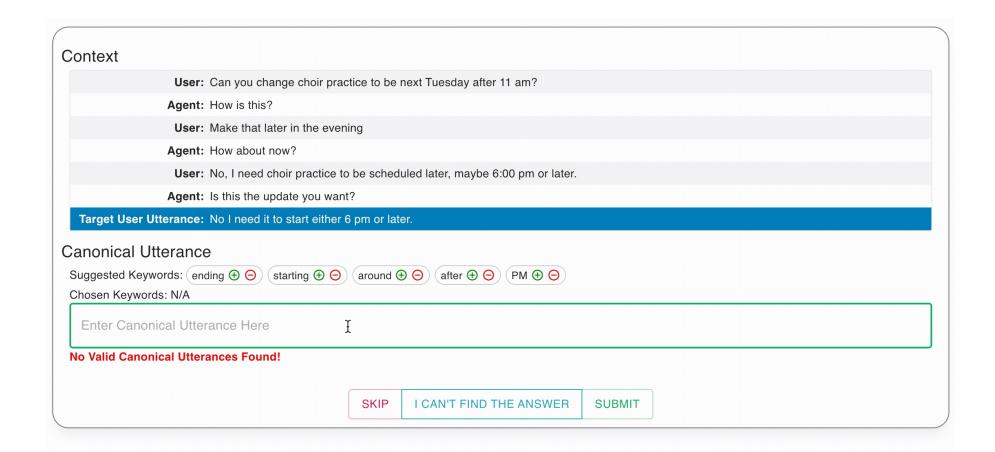


Example taken from SMCalFlow (Semantic Machines et al., 2020) dev set.

Annotation Interface: Autocomplete

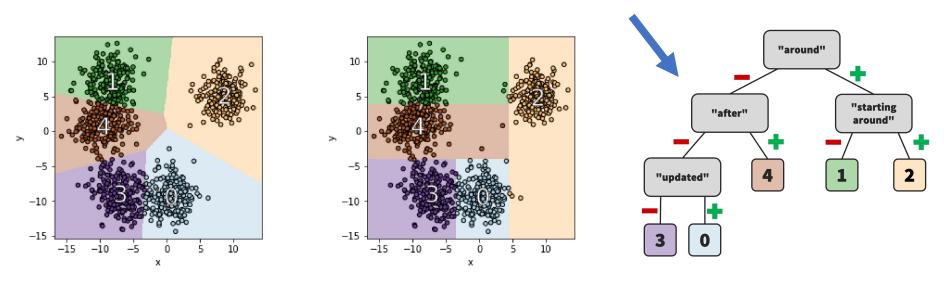


Annotation Interface: Search



Guiding Annotators via Keyword Suggestion

- Inspired by clustering-based methods for diverse paraphrasing (Hu et al, 2019)
- Instead of picking a cluster, users choose to include (+) or exclude (-) keywords
- We use explainable k-means (Dasgupta et al, 2020) to distill k clusters into k' < k keywords, which are generated from the intermediate nodes of the explanation tree



Hu et al. Large-scale, Diverse, Paraphrastic Bitexts via Sampling and Clustering. CoNLL 2019 Dasgupta et al. Explainable k-Means and k-Medians Clustering. ICML 2020

Experimental Settings

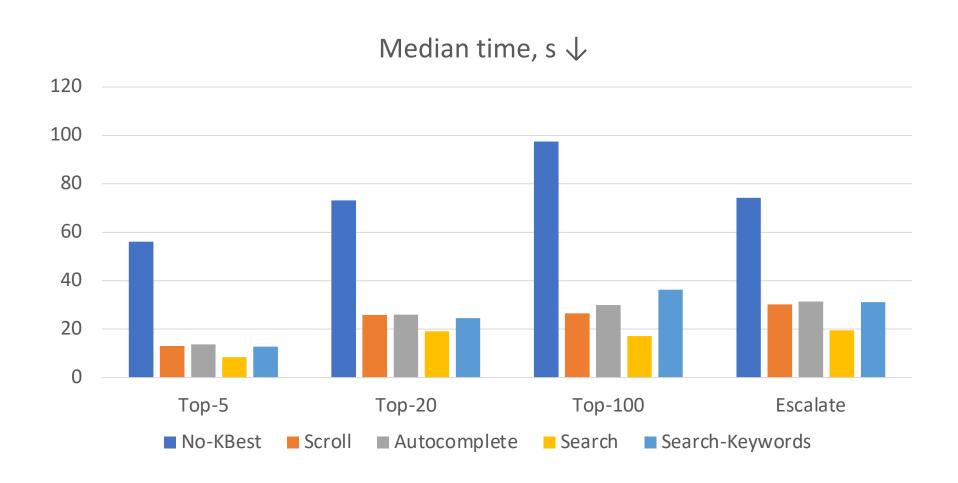
Interface comparison

- Model: VACSP (Platanios et al, 2021), generates *K*=100 best parses
 - Train set: SMCalFlow (Semantic Machines et al, 2020), 1000 utterances (sampled uniformly at random)
 - Eval set: SMCalFlow (Semantic Machines et al, 2020), 300 utterances (Top-5, Top-20, Top-100, Escalate)
- Participants: 5 annotators, randomly assigned to the specific (UI, data split)

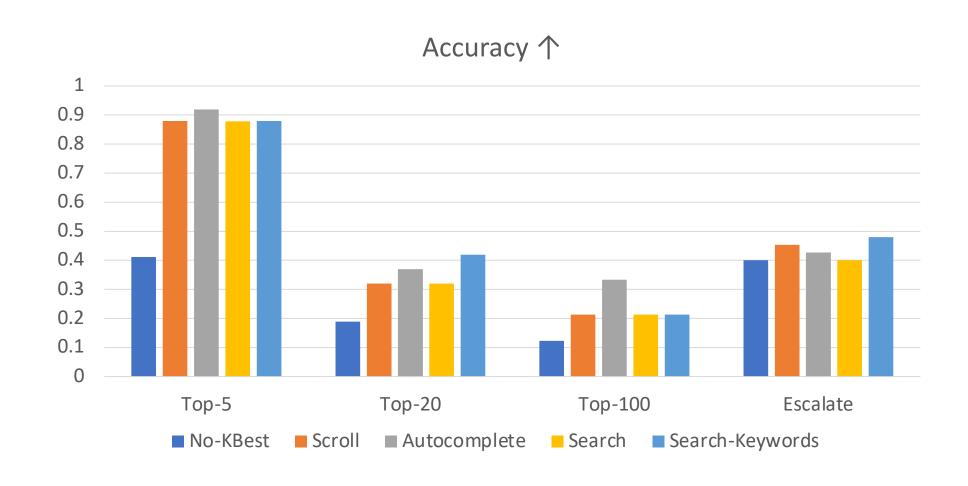
Guidance comparison

- Model: VACSP (Platanios et al, 2021), generates *K*=100 best parses
 - Train set: SMCalFlow (Semantic Machines et al, 2020), 1000 utterances (sampled uniformly at random)
 - Eval set: SMCalFlow (Semantic Machines et al, 2020), 12000 utterances

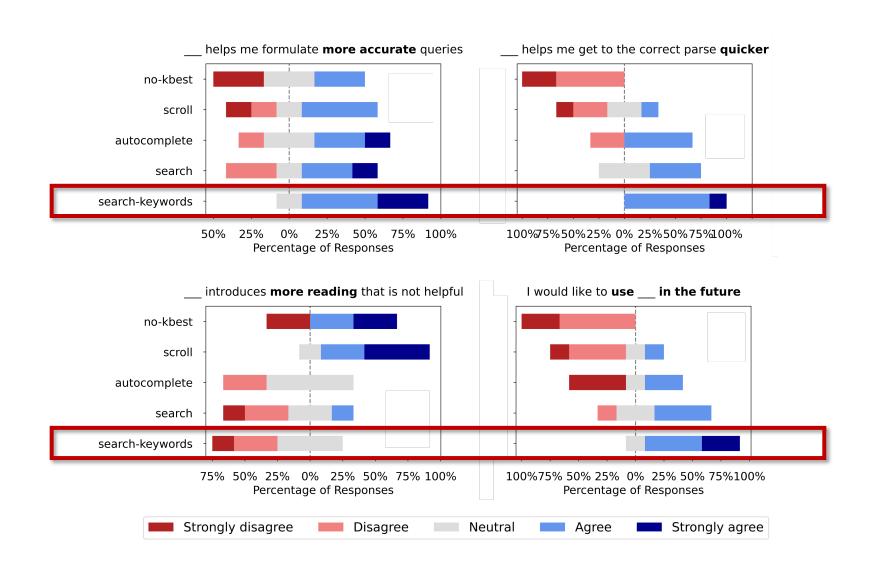
Results: Annotation Speed



Results: Annotation Accuracy



Results: Annotators' Feedback



Experimental Settings

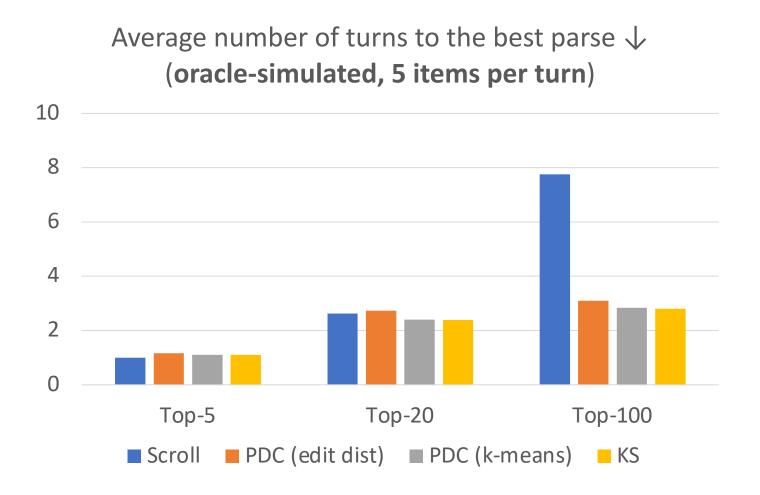
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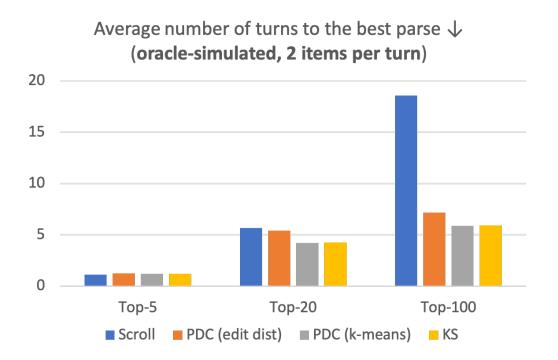
Guidance comparison

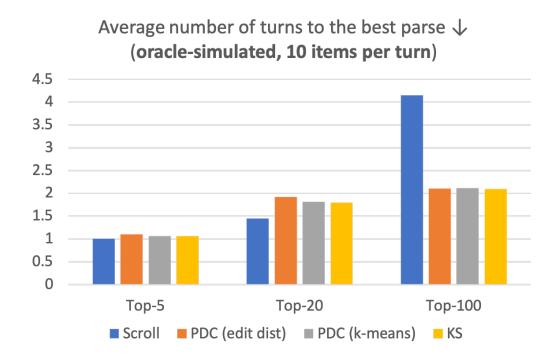
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Results: Guidance Comparison



Results: Guidance Comparison (Cont'd)





Takeaways

- Proposed human-in-the-loop annotation UIs that leverage guidance mechanisms over K-best lists generated by a low-resource model
- Guidance mechanisms are based on autocompletion (via prefix match) and search keyword suggestion (via explainable k-means clustering)
- Guidance-based UIs achieve higher annotation speed and accuracy than non-guided baseline interfaces, i.e. No-KBest and Scroll
- Among guidance-based UIs, Autocomplete achieves the highest accuracy, Search achieves the highest speed, and Search-Keywords receives the most positive feedback while balancing accuracy and speed