Supplementary Materials

We present screeshots of the user interface that was presented to users while performing our three stage online experiment.

| | Scholar using Natural Language |
|---|---|
| | Scholar is a database with facts about Papers together with their Keyphrases (e.g. Machine Translation), Datasets (e.g. Imagenet), Authors, Conferences and Journals. |
| | Note: Our database is incomplete, so results may be fewer than you expect. |
| ſ | E.g: papers by Michael I. Jordan <u>Run</u> |
| | E.g: keyphrases used by Michael I. Jordan Run |
| | E.g: How many papers does Michael I. Jordan have ? Run |
| | Write your natural language query here: (Try to capitalize noun phrases e.g. Semantic Parsing, instead of semantic parsing) |
| | Capitalize all names, keywords, years, conferences, paper titles etc. |
| | Execute |

Figure 1: Users were presented with example utterances and a text box to enter their own utterance.

Scholar using Natural Language

| | n) |
|--|---|
| le have seen a | similar query before! paper by "michael i. jordan"(AUTHOR) |
| VE HAVE SEEN a | |
| eedback: | |
| Correct | The result answered your question. |
| Wrong Types | The identified names/titles/keyphrase/years (in blue) are not what you intended. |
| Incomplete Result | The result answers your question but is incomplete. For eg. Missing papers or Low number of papers. |
| Wrong Result | The result shows something other than what you wanted, or, the result doesnt make sense. Eg. Authors instead of papers. |
| I can't tell | It looks correct to you, but you're not totally sure. |
| | title abstract numCiting numCitedBy year |
| ggle Columns: ow 10 \$ entri title | |
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| ow 10 \$ entri title Variational Prin | es |
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Figure 2: Once the user writes an utterance and pushes execute, they are presented with this screen. First, the identified entities and their types are highlighted to help them decide if the model is receiving the correct inputs. Second, in case the generated (anonymized) query already exists in the training set, it is presented as an alternate utterance, to give users additional confidence about the results that they are seeing. Third, they are presented with 5 feedback options as discussed in the main paper. Finally, they are presented with results and they can toggle columns with additional information.