

TriviaNLQ: Natural Language Query for trivial questions

M2 TAL Software Project

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SPARQLLess queries

Several issues came up while trying to run `sparqling-queries` [1]:

- ▶ broken dependencies
- ▶ complicated infrastructure, requires a database
- ▶ hard to adapt code, only works with the given dataset

Overall, it didn't work out.

Back to code generation

Turn this into Wikidata SPARQL query: "What instrument does Kermit the Frog play?".

```
SELECT ?instrument WHERE {  
  wd:Q717 wdt:P361 ?instrument .  
  FILTER contains(?instrument, "Kermit the Frog")  
}
```

Figure: GPT-3 composed query

Looks OK at the first glance and even executes but:

- ▶ Q717 is Venezuela
- ▶ P361 is a relation "part of"
- ▶ it keeps only the results that have "Kermit the Frog" in its description

GPT-3 composed queries

- ▶ don't execute (syntax errors, maximum limit reached)
- ▶ don't match objects and predicates
- ▶ aren't deterministic

- ▶ Convolutional Sequence-to-Sequence model
- ▶ Based on TNTSPA [2]
- ▶ Best performing model compared to RNN-based models (NSpM, NSpM+Att1, NSpM+Att2, LSTM Luong, GNMT-4, GNMT-8) and Self-attention models (Transformer) for text-to-sparql [2]

ConvS2S

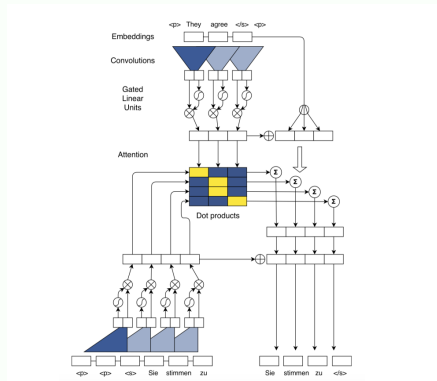


Figure: Convolutional Sequence-to-Sequence model [GAG + 17][2]

Dataset

- ▶ Monument dataset
- ▶ Better results over the two other main datasets in literature (LC-QUAD and DBNQA) [2]

Dataset

	Monument
Instance	14,788
English vocab	2,500
SPARQL vocab	2,200

Figure: Size of the English-SPARQL Monument dataset [2]

Dataset Preprocessing

SPARQL	<pre>SELECT DISTINCT ?uri WHERE { <http://dbpedia.org/resource/Sam_Loyd> <http://dbpedia.org/ontology/knownFor> ?uri . <http://dbpedia.org/resource/Eric_Schiller> <http://dbpedia.org/ontology/knownFor> ?uri . }</pre>
Encoded	<pre>select distinct var_uri where brack_open dbr_Sam_Loyd dbo_knownFor var_uri sep_dot dbr_Eric_Schiller dbo_knownFor var_uri sep_dot brack_close</pre>

Figure: SPARQL Encoding [2]

Dataset Splitting

- ▶ Splitting of the Monument dataset into 80%/10%/10% training/validation/testing set [2]

ConvS2S

Frameworks

- ▶ fairseq based on PyTorch (sequence modeling toolkit that allows researchers and developers to train custom models)
- ▶ implemented on ConvS2S

Hyperparameters

- ▶ Recommended hyperparameters from fairseq [2]

Results

Model	Validation	Test
ConvS2S	94 99	95 96

Figure: Accuracy (in %) of syntactically correct generated SPARQL queries | F1 score)[2]

Next Steps

- ▶ Wrapper for the model
- ▶ Script for turning the encoded output into SPARQL
- ▶ Front-end
- ▶ Script for querying DBpedia
- ▶ Check the performance on targeted trivia questions

References

- [1]Saparina, I., Osokin, A. (2021). Sparqling database queries from intermediate question decompositions. arXiv preprint arXiv:2109.06162.
- [2]Yin, X., Gromann, D., Rudolph, S. (2019). Neural machine translating from natural language to SPARQL. Future Generation Computer Systems, 117, 510-519.