Question Rewriting?
Assessing Its Importance for
Conversation Question Answering

Gonçalo Raposo

INESC-ID, Instituto Superior Técnico, Universidade de Lisboa

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#### Introduction

Conversational question answering systems must correctly interpret interconnected interactions and generate knowledgeable answers, which may require the retrieval of external information. This work presents our conversational question answering system, leveraging neural language models, and reports a detailed analysis on various inputs for the question rewriting module. The results obtained emphasize the importance of the conversation context representation for the overall system performance.

### Methods

- Conversational question rewriting using a T5 model fine-tuned on the CANARD dataset.
- Retrieve the top-10 most relevant passages using the BM25 ranking function.
- Generate the answer using a fine-tuned Pegasus model. Answer can be seen as a summary of the retrieved passages, conditioned also on conversation data.
- Trained and tested using QReCC dataset.

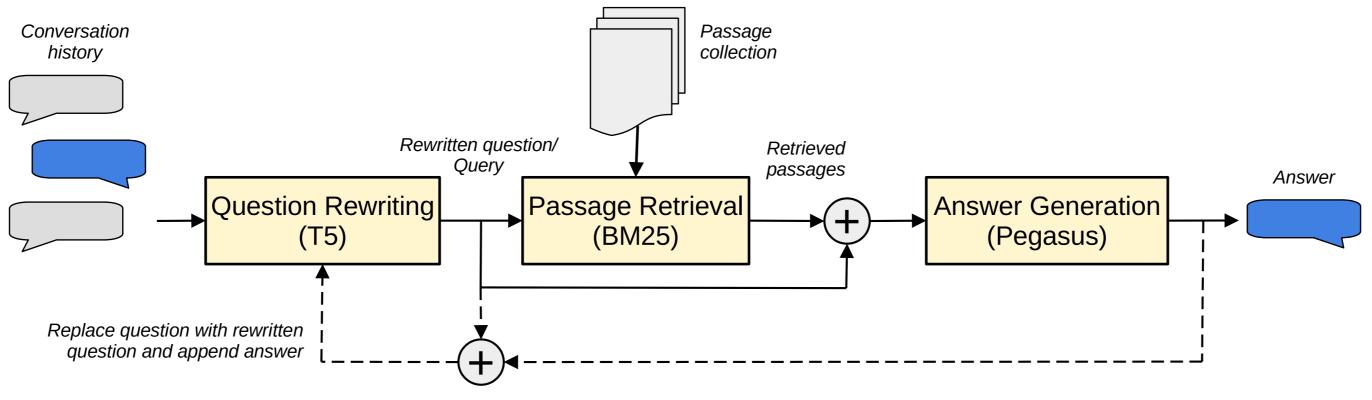


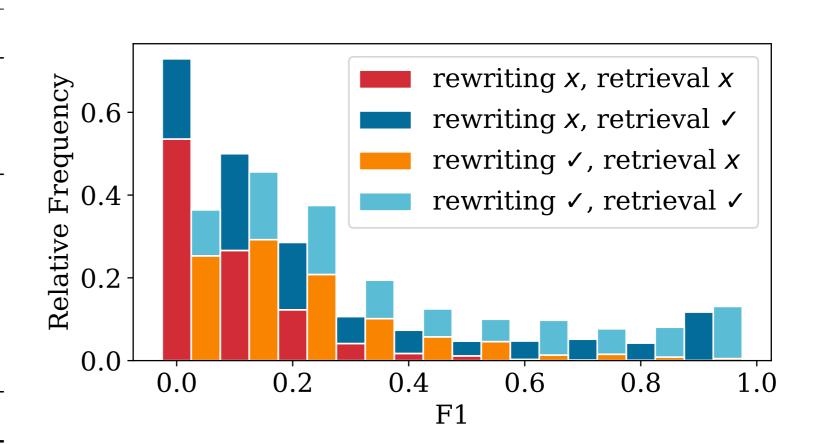
Fig. Proposed conversational question answering system.

#### Results

**Tab.** Evaluation of multiple variations of the input used in the question rewriting module.

Description	Rewriting Input	Rewriting	Retrieval		Answer	
		ROUGE1-R	MRR	F1	EM	ROUGEL-F1
SCAI baseline: question	-	-	-	0.117	0.000	0.116
SCAI baseline: retrieved	-	(0.571)	0.065	0.067	0.001	0.073
SCAI baseline: GPT-3	-	-	-	0.149	0.001	0.152
No rewriting $(h = 1)$	-	(0.571)	0.061	0.136	0.005	0.143
No rewriting $(h = 7)$	-	(0.571)	0.145	0.155	0.003	0.160
Questions	(Q) + Q	0.673	0.158	0.179	0.011	0.181
Questions + answers	(Q + MA) + Q	0.681	0.150	0.179	0.010	0.181
Rewritten questions	(MR) + Q	0.676	0.157	0.187	0.010	0.188
Rewritten + answers	(MR + MA) + Q	0.685	0.149	0.189	0.010	0.191
Ground truth rewritten	-	(1)	0.385	0.302	0.028	0.293
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**Fig.** Analysis of the influence of question rewriting on answer generation performance.



**E-mail:** goncalo.cascalho.raposo@tecnico.ulisboa.pt **GitHub:** https://github.com/gonced8/rachael-scai

## Conclusions

- 1st place in SCAI shared task 2021.
- Our system surpasses a GPT-3 baseline by 26.8% in answer accuracy.
- Question rewriting improved overall answer accuracy by about 20~30%.
- When the question rewriting succeeded, both the retrieval and answer generation improved – higher scores were about 2x more frequent.

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