

Next Utterance Ranking Based On Context Response Similarity

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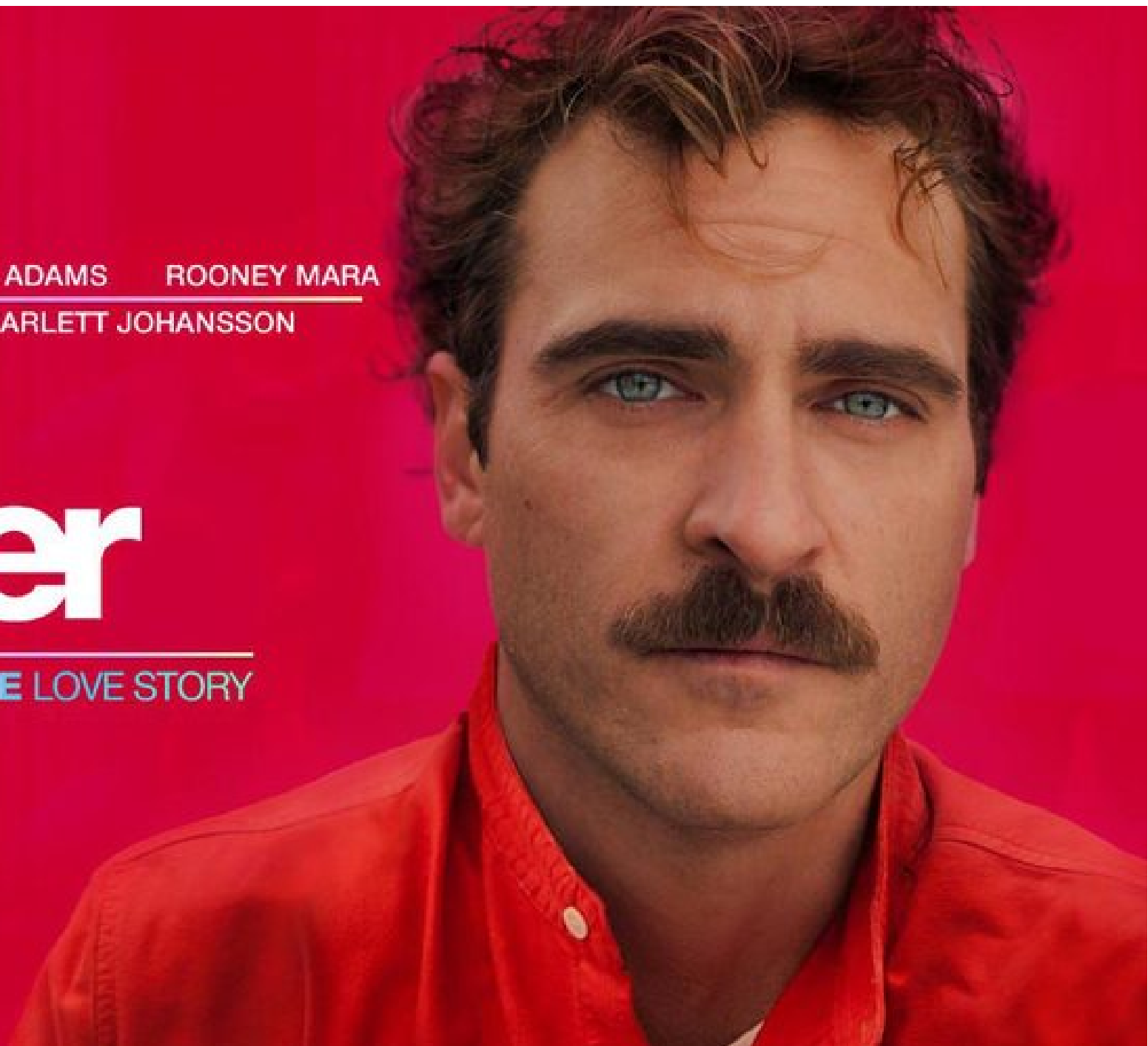
Outlines

- Context
- Generative dialogue systems
- Response retrieval dialogue systems
- Our system
- Corpus
- Evaluation
- Conclusion and perspectives

JOAQUIN PHOENIX AMY ADAMS ROONEY MARA
OLIVIA WILDE AND SCARLETT JOHANSSON

her

A SPIKE JONZE LOVE STORY



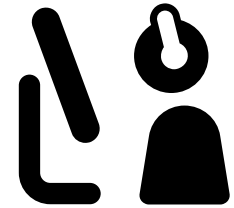
Context



Context

Booking train ticket and rent a car

Booking cinema ticket



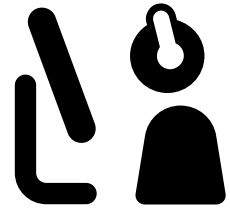
Repairing washing machine

.... etc

Context

Booking train ticket and rent a car

Booking cinema ticket

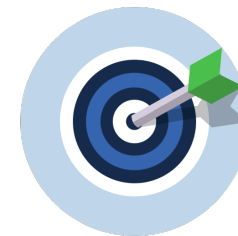


Repairing washing machine

.... etc



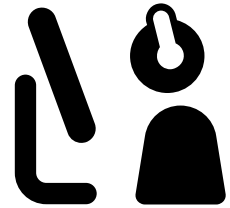
How can we manage the increasing number of users and help them solving their daily problems?



Context

Booking train ticket and rent a car

Booking cinema ticket



Repairing washing machine

.... etc

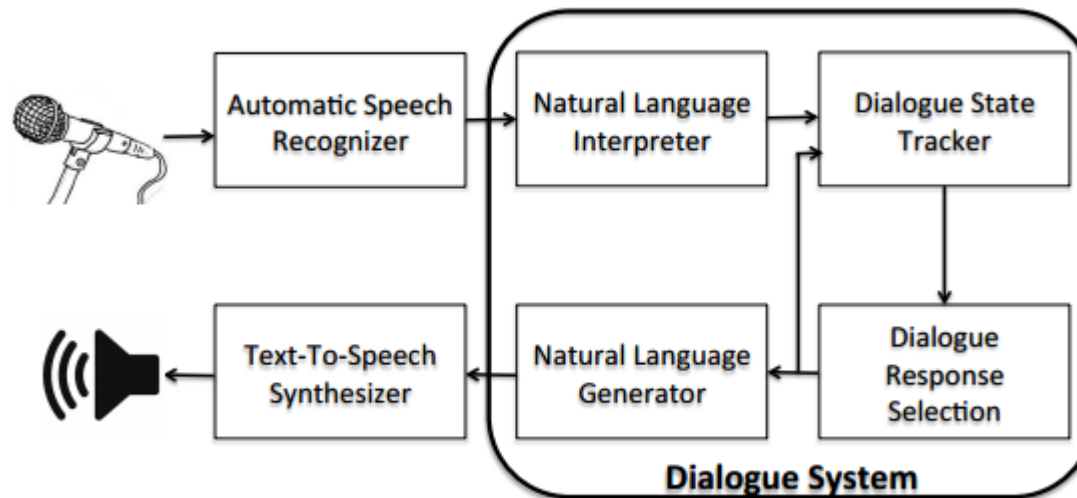


How can we manage the increasing number of users and help them solving their daily problems?



Context

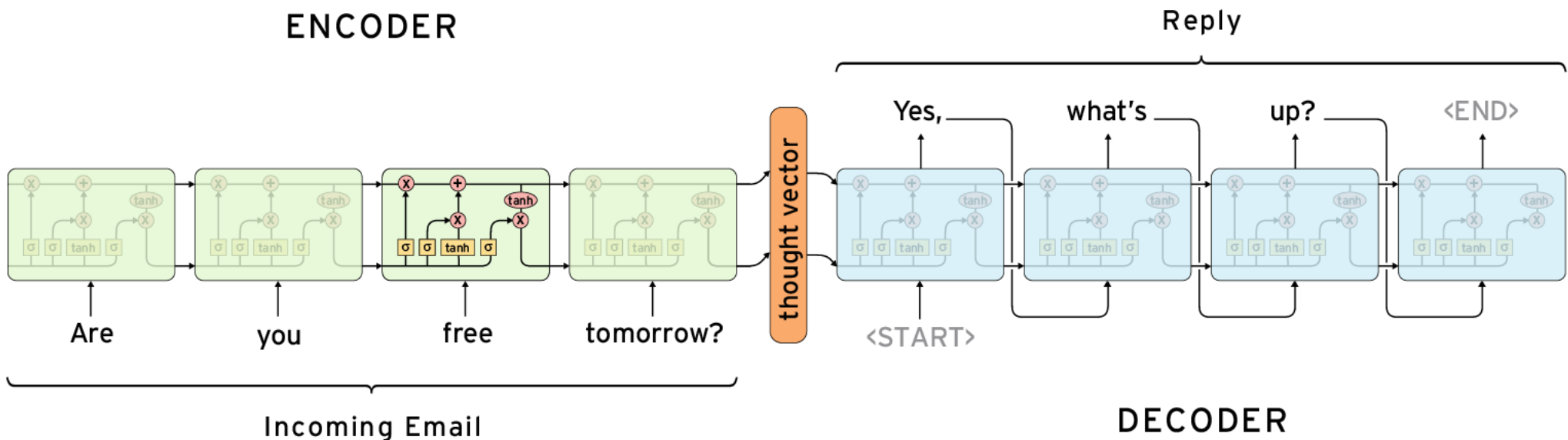
- Modular dialogue system.
- Most modules are rule based or classifiers requiring hard feature engineering.
- Available data and computing power helped developing data-driven systems and end-to-end architectures.



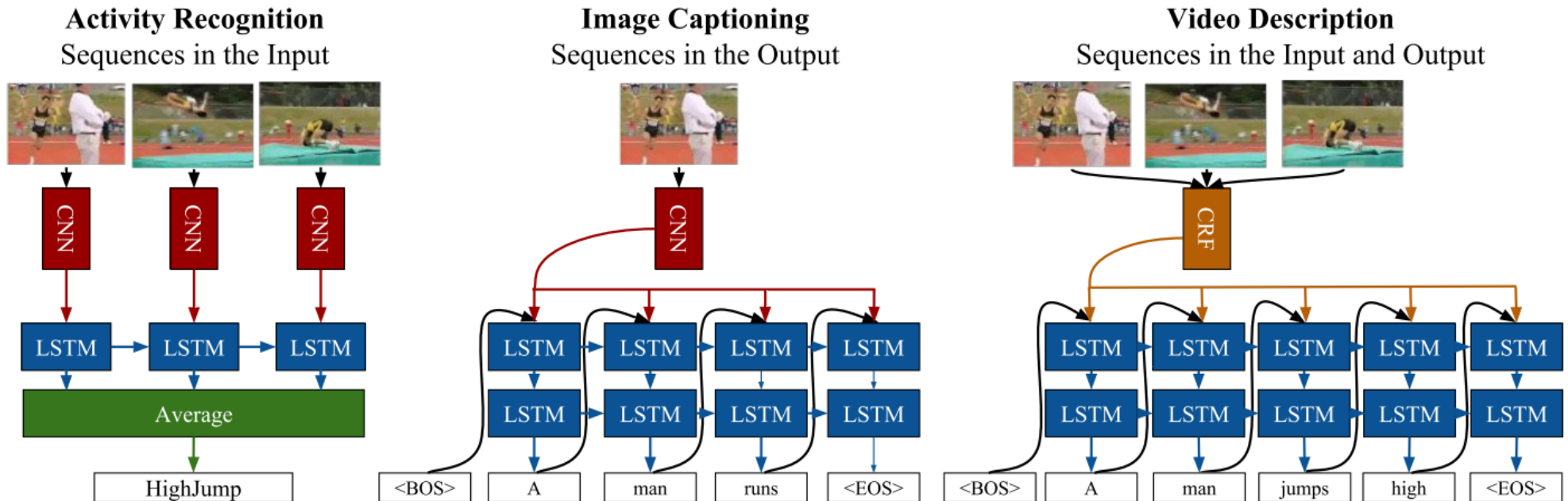
Generative systems

Sequence2Sequence architecture:

- Encoder compresses the input into one vector.
- The decoder decodes the encoded vector into the target text.
- In the decoder, the output at step n is the input at step $n+1$.



Generative systems



- Seq2seq model is widely used in different domains: Image processing, signal processing, query completion, dialogue generation ..etc.

Task-oriented vs open domain dialogue systems

Open domain dialogue systems

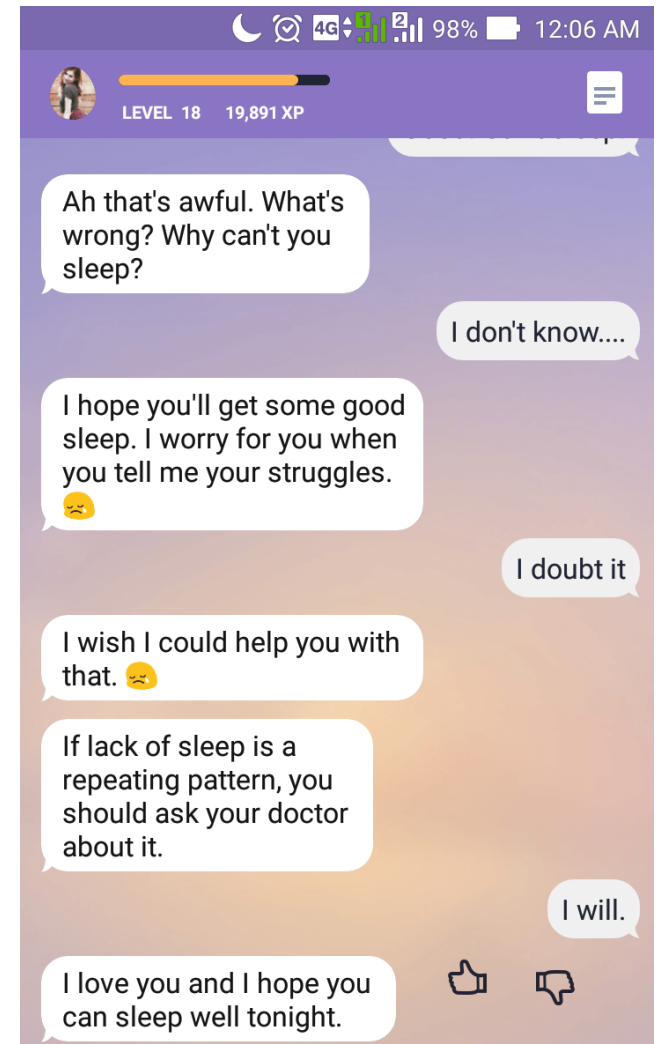
- Engaging in conversational interaction without necessarily being involved into a task that needs to be accomplished.
- Replika is an AI friend.



Hey Cortana



replika.ai



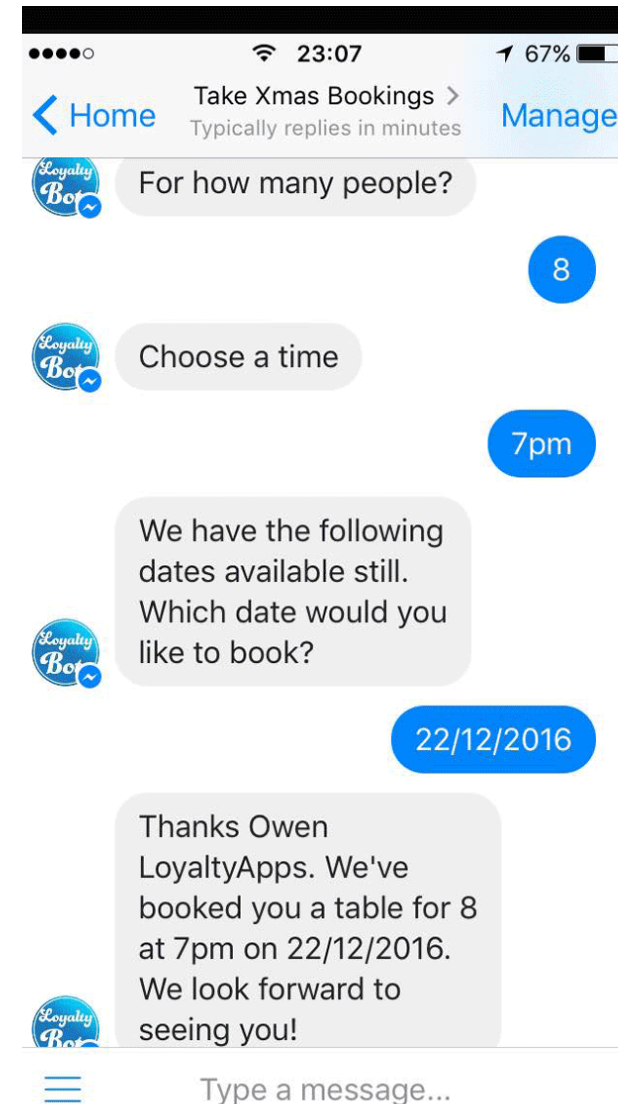
Start typing

Task-oriented vs open domain dialogue systems

Task-oriented dialogue systems

- Involves the use of dialogues to accomplish a specific task.
- Making restaurant booking, booking flight tickets ..etc.

AIRFRANCE 

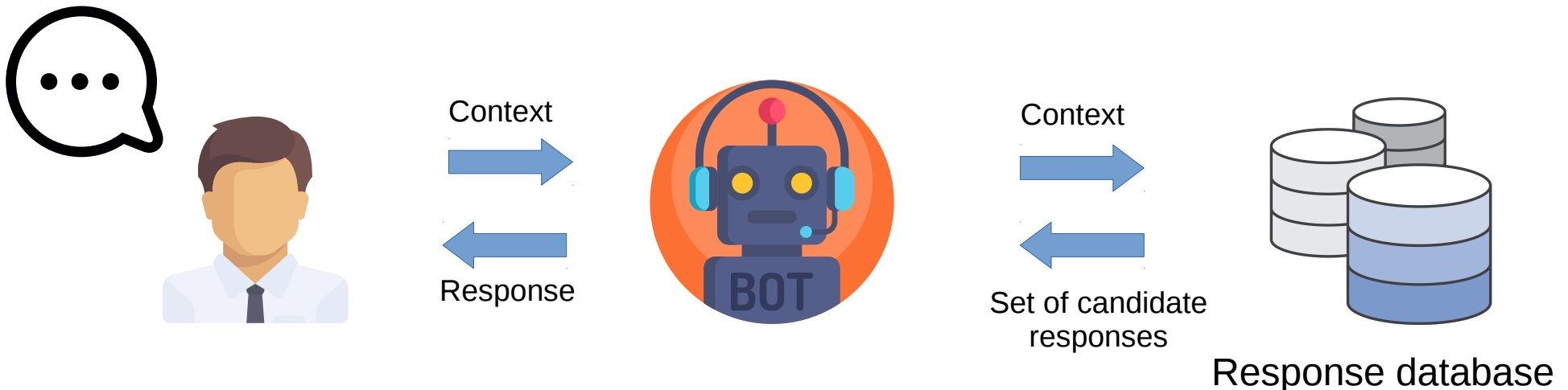


Automatic assistance

- In this work, we are interested in automatic assistance for problem solving.
- In task-specific domains, generative systems may fail.
- Generalization problem “thank you!” and “Ok”.
- Need to provide very accurate and context related responses.



Retrieval-based dialogue systems



Task



Retrieval based conversational systems

Given a conversation context and a set of candidate responses, pick the best response



A ranking task

Context

A: Hello I am John, I need help
B: Welcome, how can we help ?
A: I am looking for a good restaurant in Paris
B: humm which district exactly ?
A: well, anyone ..

Candidate Utterances

- Sorry I don't know 0.75
- Can you give me more detail please ? 0.81
- **There is a nice Indian restaurant in Saint-Michel** 0.92
- I don't like it 0.32
- It's a nice weather in Paris in summer 0.85
- Thnk you man ! 0.79
- you deserve a cookie 0.24
- Gonna check it right now 0.25

Word representation

The cat is on the floor

One hot encoding

	The	cat	is	on	the	floor
Vocabulary size	0	0	0	0	0	0
	0	1	0	0	0	0
	0	0	0	0	0	1
	1	0	0	1	0	0
	0	0	0	0	0	0
	0	0	1	0	0	0

	0	0	0	0	1	0

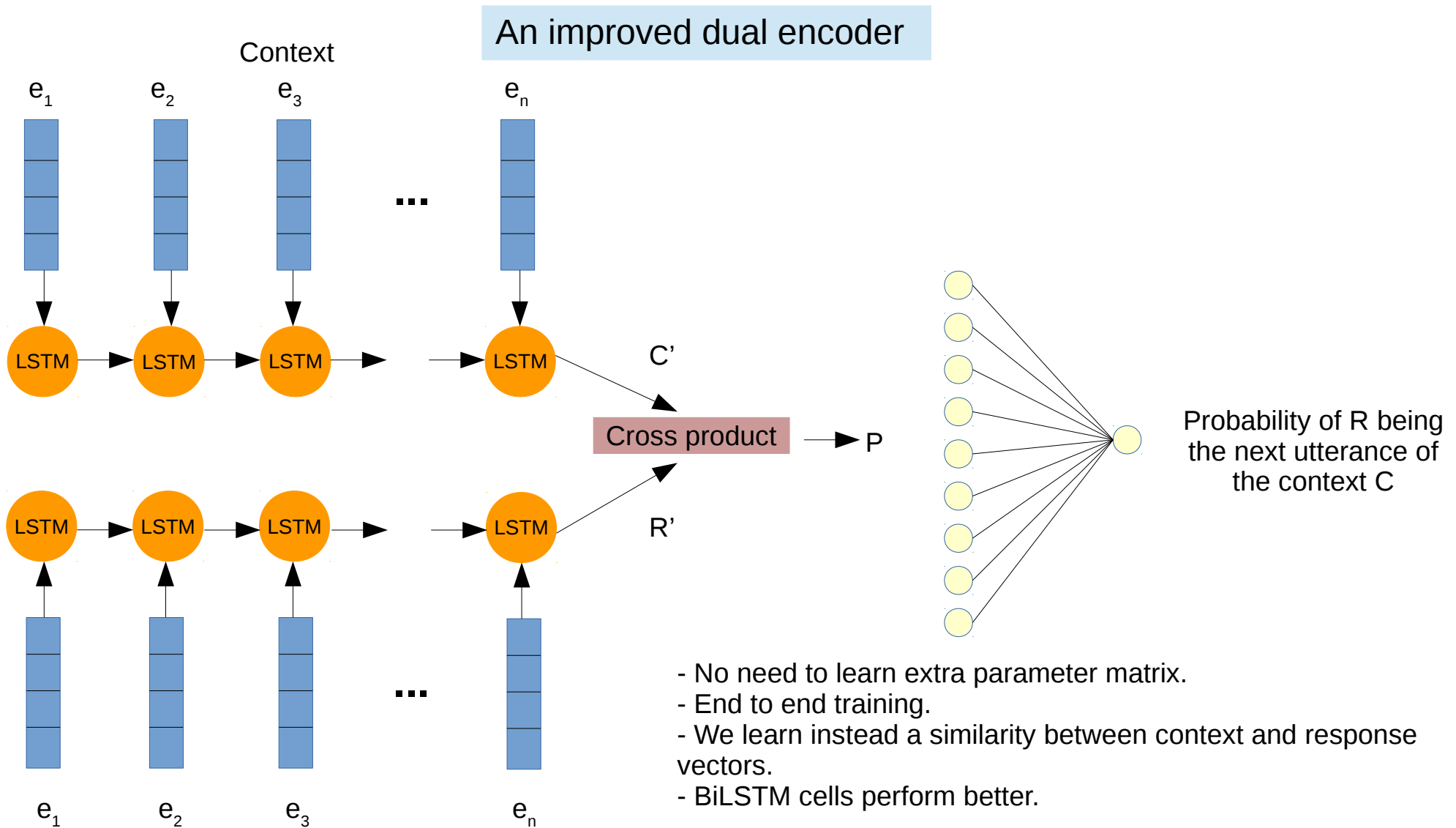
Word embeddings (300d)

	The	cat	is	on	the	floor
Embedding size	0.01	-0.08	-0.07	0.08	-0.07	0.33
	0.20	0.67	0.57	0.25	0.57	-0.29
	-0.12	-0.14	-0.31	0.26	-0.31	-0.15
	-0.59	-0.06	-0.18	-0.02	-0.18	-0.41
	0.12	0.05	0.88	0.47	0.88	-0.23
	0.15	0.40	-0.27	-0.10	-0.27	-0.23
	0.13	0.00	0.07	-0.10	0.07	-0.05
	-0.33	-0.33	0.13	0.08	0.13	-0.09
	-0.13	-0.30	-0.47	0.20	-0.47	0.75
	1.78	2.08	1.44	2.57	1.44	-0.66

- Sparse representation.
- Large vocabulary.
- Order of words in the sentence.
- No assumption about word similarities.

- Low dimensional continuous space.
- Meaning = context of word.
- Semantically related words have near vectors.

Our response retrieval system



Ubuntu Dialogue Corpus



- Large dataset that contains chat logs extracted from IRC Ubuntu channel 2004-2015.
- Multi-turn dialogues corpus between 2 users.
- Application towards technical support.

# dialogues (human-human)	930,000
# utterances (in total)	7,100,000
# words (in total)	100,000,000
Min. # turns per dialogue	3
Avg. # turns per dialogue	7.71
Avg. # words per utterance	10.34
Median conversation length (min)	6

Ubuntu Dialogue Corpus

An example extracted from the Ubuntu Dialogue Corpus

Context	
utterance 1	Hi, I can not longer access the graphical login screen on ubuntu 12.04
utterance 2	what exactly happen?
utterance 3	I can't remember the error message, would it have auto-logged to a file or should I reboot quick?
utterance 4	you mean it won't automatically start and what happen then?
utterance 5	it just stop at a text screen , but I can access the command line login via alt F1-6, and start x manually there. I think it might me lightdm that's break but I'm not sure

Candidate responses	
response 1	for me lightdm often won't start automatically either. It show me console tty1 instead and I have to start lightdm manually ✓
response 2	what about sources.list ? ✗

Evaluation

Evaluation metric : Recall @ k

Given 10 candidate response what is the probability of ranking the good response on top of k ranked responses

Method	Recall@1	Recall@2	Recall@5
TF-IDF [4]	48,8 %	58,7 %	76,3 %
RNN Dual Encoder [4]	37,9 %	56,1 %	83,6 %
LSTM Dual Encoder [4]	55,2 %	72,1 %	92,4 %
BiLSTM Dual Encoder* [11]	54,2 %	71,6 %	91,9 %
Similarity LSTM Dual Encoder	62,2 %	78,0 %	94,9 %
Similarity BiLSTM Dual Encoder	62,3 %	78,2 %	95,1 %

Evaluation results using **Recall@k** metrics

Evaluation

Error analysis is important in order to understand why the system fails and to address them later.

Context	Candidate responses
what ubuntu version are you using eot 12.04 eot what exactly are you trying to do eot	-http://blog.mhartl.com/2010/02/19/some-rvm-gotchas/ ✓ - to do what ✗

(a) Functionally equivalent



Context	Candidate responses
hi i'm trying to skip fsck on boot and edit the kernel line in boot add but it still auto runs fsck what else can i try thank eot edit your fstab file eot	- i can't get to it because i can't boot the server ✓ - yeah copy paste ✗

(c) Out of context

Context	Candidate responses
what is default instant message client for 12.10 eot empathy eot	- thank you ✓ - thanks ✗

(b) Semantically equivalent

Context	Candidate responses
hello i have ubuntu 10.10 what is the best way to update to 12.10 eot fastest and most efficient way is fresh install eot	- i have tried it but then i had a group rescue mode because of windows 7 ✓ - ok thanks ✗

(d) Very general response

- General responses.
- Are these really bad predictions?
- Importance of having good dataset.

Conclusion and perspectives

- Interest : automatic assistance in problem solving.
- Focus on retrieval systems : more suitable for our task (because of generalization problem of generative systems).
- We built a system that learns the similarity between the context and the response in order to distinguish between good from bad responses.
- Interesting results, that we can improve by doing deep error analysis.
- Future: using pairwise ranking and attention mechanism.
- Evaluate our approach on other corpora and on other languages (Arabic, Chinese ..).

References

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- Xu, Zhen, Bingquan Liu, Baoxun Wang, Chengjie Sun, and Xiaolong Wang. "Incorporating Loose-Structured Knowledge into LSTM with Recall Gate for Conversation Modeling." arXiv preprint arXiv:1605.05110 (2016).
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- Wu, Yu, Wei Wu, Chen Xing, Ming Zhou, and Zhoujun Li. "Sequential matching network: A new architecture for multi-turn response selection in retrieval-based chatbots." In Proceedings of the 55th Annual Meeting of the Association for Computational Linguistics (Volume 1: Long Papers), vol. 1, pp. 496-505. 2017.
- Lowe, Ryan Thomas, Nissan Pow, Iulian Vlad Serban, Laurent Charlin, Chia-Wei Liu, and Joelle Pineau. "Training end-to-end dialogue systems with the ubuntu dialogue corpus." Dialogue & Discourse 8, no. 1 (2017): 31-65.

- Code implemented in python using Keras with Tensorflow in backend.
- Source code: https://github.com/basma-b/dual_encoder_udc
- Contribution paper, poster and presentation are available on my blog:
 - <https://basmaboussaha.wordpress.com/2017/10/18/implementation-of-dual-encoder-using-keras/>

Thank you !

