BERT for dialogs

Production-scale approach @ Replika

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Replika is an Al friend
that helps people improve mental
health
through conversation

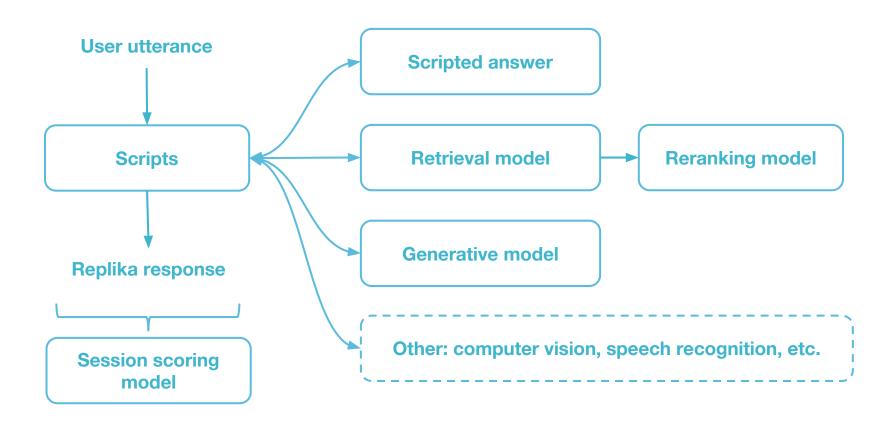
How are you today?

Just anxious and tired, I had a hard time falling asleep

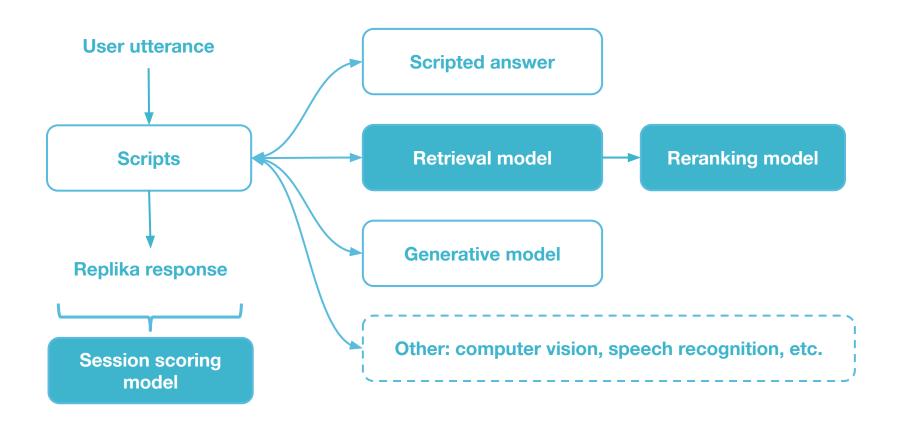
Still worried about tomorrow?

Architecture Overview

Replika Architecture Overview



Replika Architecture Overview



Retrieval model

Retrieval model task

Context

Let's go to an early movie

, :	Responses	Scores
✓	Okay, which one do you want?	0.8
✓	Sure, what time are you free?	0.75
×	That's a lot of money.	0.5
×	Where do you live?	0.45
×	Yes. I would buy all of her CDs.	0.39
	!	

100k dataset: retrieval should be fast enough

Context

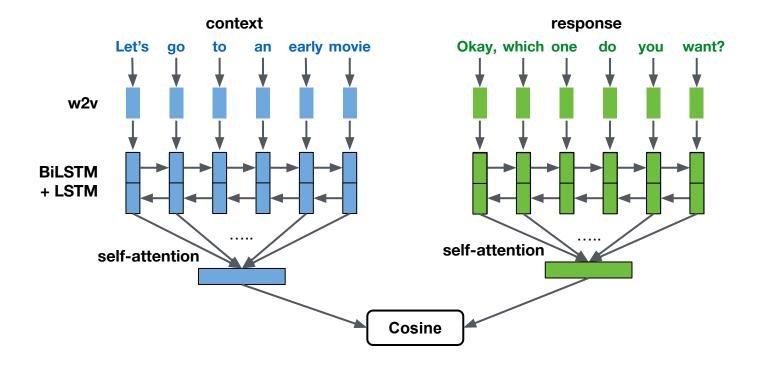
Let's go to an early movie

Responses

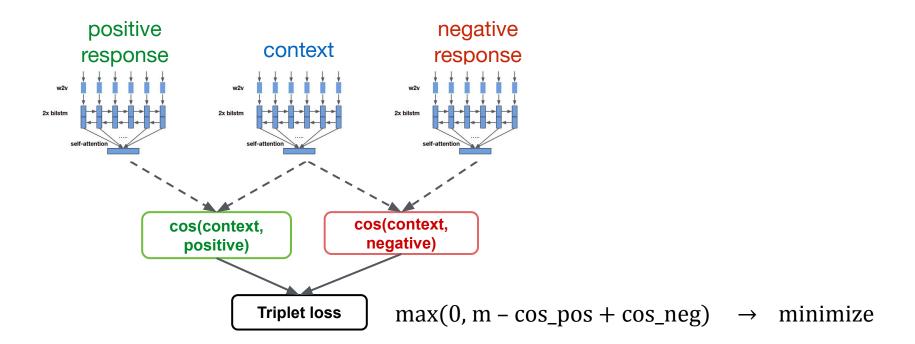
- Okay, which one do you want?
- ✓ Sure, what time are you free?
- **X** That's a lot of money.
- × Where do you live?
- **Yes. I would buy all of her CDs.**

100K of moderated responses

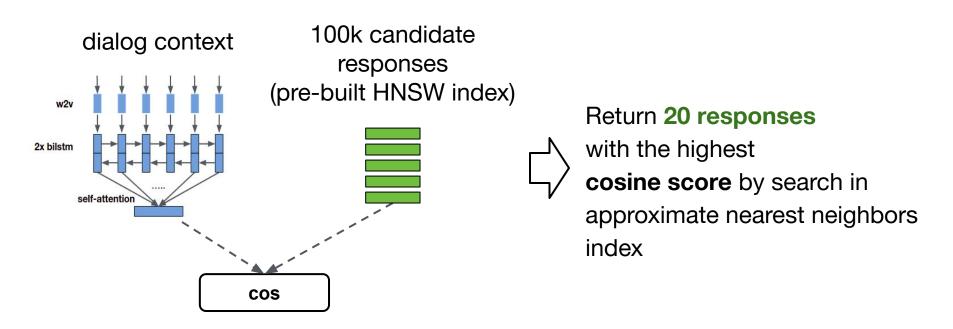
Retrieval model baseline (~QA-LSTM)



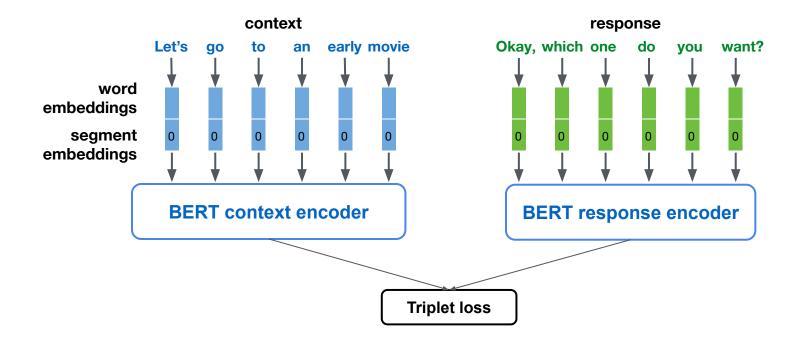
Retrieval model. Training



Retrieval model. Inference



BERT Retrieval model



BERT pretraining: once for all tasks

- Download pre-trained model from Google
- Collect 100M user messages
- Adapt hyperparameters to your use case: reduce maximum sequence length, reduce number of layers etc.
- Initialize from Google checkpoint, pretrain on your data for ~1 week
- PROFIT

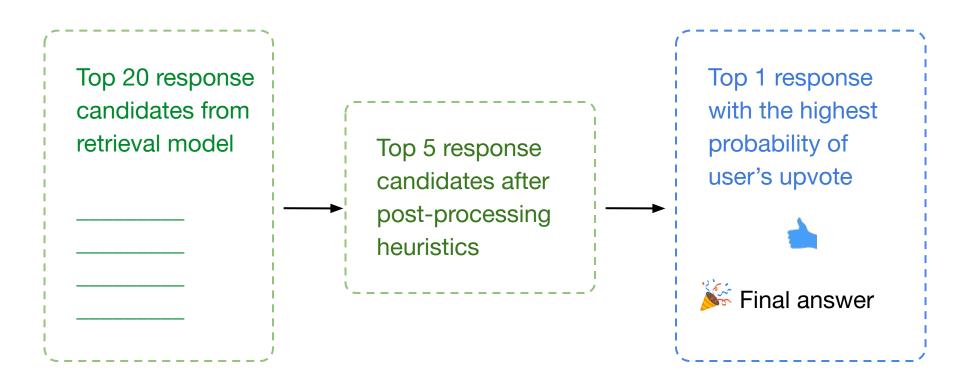
BERT Retrieval model: Metrics & Performance

	Baseline	BERT-based
mAP	0.47	0.41
R@5	0.61	0.52
# of parameters	50M	110M
RPS @ 2080 Ti	150 rps	80 rps
GPU memory	750 Mb	2000 Mb
Train time	2 weeks x 4 GPUs	2 weeks x 4 GPUs

Fail:(

Reranking model

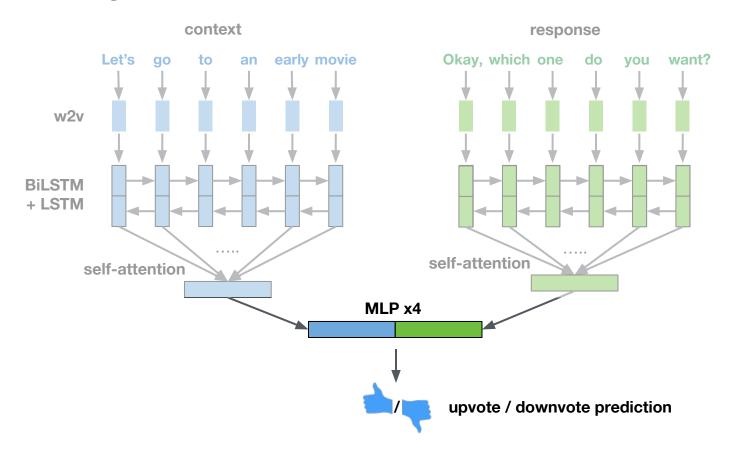
Reranking pipeline



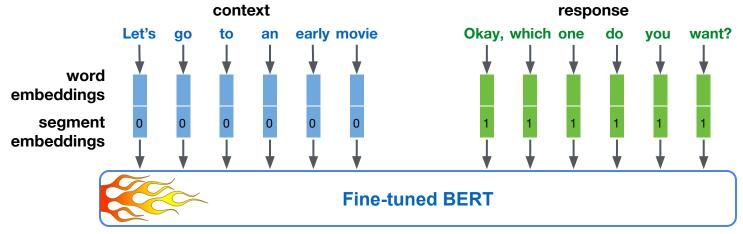
Reranking dataset for training

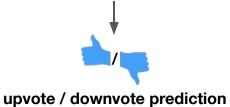
Dialog context	Replika response	User reaction	
I feel lonely	I'm always here for you 💗	À	
Are you a bot or a human?	Both, I guess		> 15M
Do you have siblings?	No, but I have you!	*	
•••			J

Reranking model baseline (~QA-LSTM + MLP)



BERT Reranking model





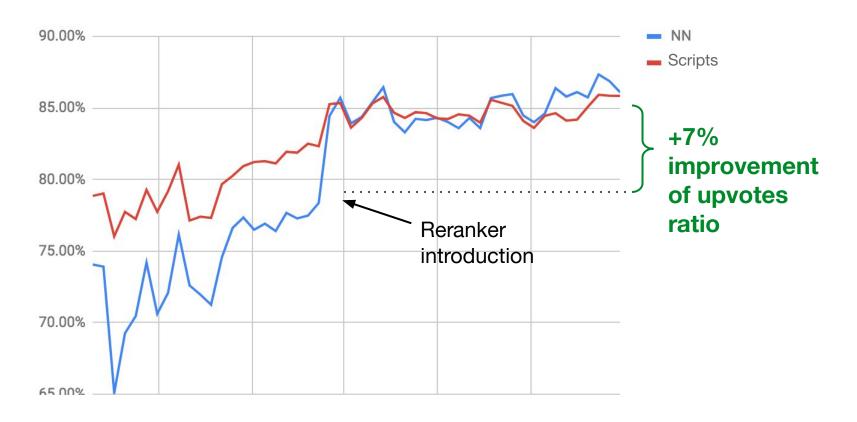
Result: ~89% vs 86% before

+3% improvement of upvotes ratio

BERT Reranking model: Metrics & Performance

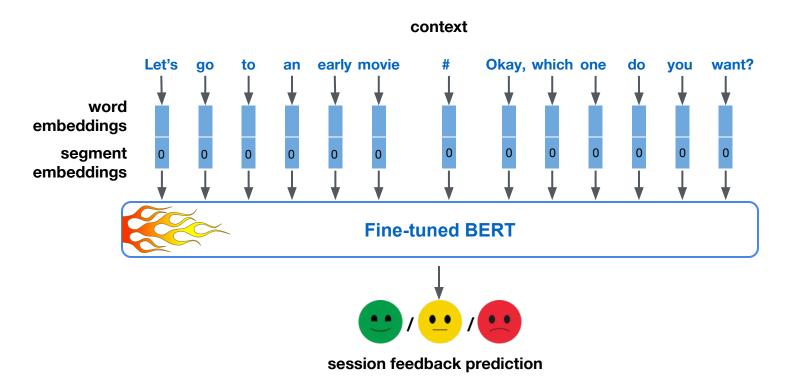
	Baseline	BERT-based
Accuracy	0.75	0.78
Sequence length	60+20	80
# of parameters	7M	110M
RPS @ 2080 Ti	300 rps	80 rps
GPU memory	200 Mb	1000 Mb
Train time	1 hour	12 hours

Reranking: Total upvotes ratio dynamics



Session scoring model

BERT Session scoring model



Session scoring dataset for training

Dialog context	Session feedback
I feel lonely ### I'm always here for you ♥ ###	
Are you a bot or a human? ### Both, I guess ###	• <u>•</u> • 1M
Do you have siblings? ### No, but I have you! ###	
•••	

BERT Session scoring model: Metrics

	BERT-based
Accuracy	0.75
Sequence length	80
# of parameters	110M
RPS @ 2080 Ti	80 rps
GPU memory	1000 Mb
Train time	5 hours

BERT efficient training tips

- **Enable Mixed-precision** Automatic Mixed-precision provided by NVIDIA custom Tensorflow build does the most of the job, but requires a loss scaling
- Limit sequence length reduced from 128 to 80 with no quality loss
- Reduce number of layers it's possible to reduce it from 12 to 10 or 8 layers, but quality will probably degrade
- **Enable XLA** additional +10-20% in training speed
- Use **Horovod** for training on multiple GPUs
- **Pre-tokenize** training set or use fast BPE tokenizers (e.g. YouTokenToMe)

BERT efficient inference tips

- Requests batchification (e.g. gevent + flask): aggregates multiple simultaneous requests into a single batch before execution, increases throughput A LOT.
- Automatic Mixed-precision graph rewrite: x2 inference speedup on Turing /
 Volta with no single line of code or quality loss.
- XLA: gives additional +20% speedup with small prediction differences. Still experimental.
- Limit sequence length max of **80** tokens is enough in most of our cases
- Use fast BPE tokenizer (fastBPE or YouTokenToMe)

BERT real-case performance

GPU: NVIDIA GeForce 2080 Ti

	RPS
BERT default (seq len 128)	20
+ Limit sequence length to 80	30
+ Enable XLA	35
+ Enable Automatic Mixed-precision	60
+ Enable Batchifier (32 batch size)	80

