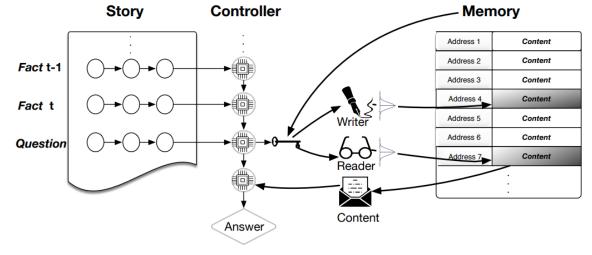


A Memory Model for Question Answering from Streaming Data Supported by Rehearsal and Anticipation of Coreference Information

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Memory Networks for NLP

- Memory networks mimic the human process of incremental comprehension and compression of information in a memory (Graves et al., 2014, 2016).
- Memory models can read contiguous text (stories) and, via a controller, write and read information to and from memory to solve a task (e.g., answer a question).



Source: (Gulcehre et al., 2017)



Memory Networks for NLP

Problem:

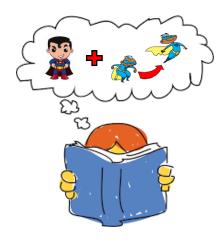
• Memory networks rely on a single task to maintain memory.

Human memory:

- Humans are endowed with mechanisms for memorization.
 - Anticipation (Hawkins and Blakeslee, 2004).
 - Memory rehearsal (Waugh and Norman, 1965)
- Complementary mechanisms (Ferreira, 2018).

Proposed Solution:

• Support a memory model with anticipation and rehearsal.





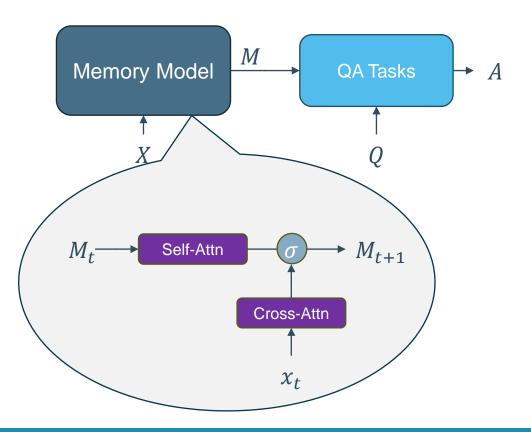
RAM: A memory network augmented with rehearsal and anticipation mechanisms.

RAM Method:

Main task: Question answering (QA) from • streaming data:

$$M = F(X)$$
, where $X = [x_1, x_2, x_3, ..., x_T]$
 $A = G(M, Q)$

- Improved Memory Cell:
 - Memory integration based on cross-attention.
 - Gating (σ) for allowing forgetting.



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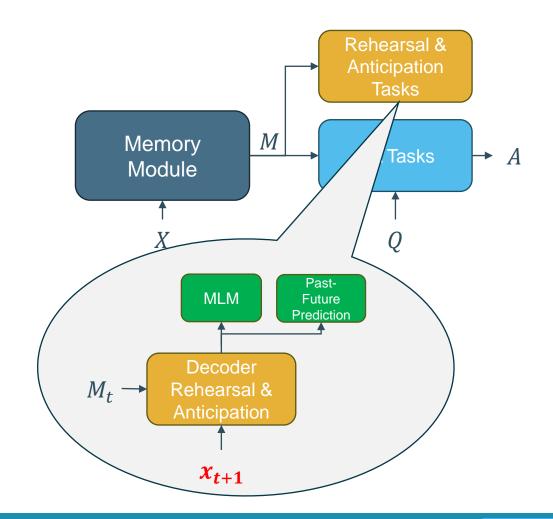


RAM Method:

- Pretext tasks as masked modeling:
 - Rehearsal as a prediction of previous elements.
 - Anticipation as a prediction of future elements.
- Rehearsal and anticipation are related and might be using the same machinery (Cole et al., 2015)
 - Shared decoder with auxiliary task.

What to rehearse and anticipate?

- Coreference information guides memory retrieval and expectations (Jaffe et al., 2020).
 - coreference-related tokens (nouns, pronouns, verbs)





Findings:

- Improved performance across tasks (also video).
- Coreference information is important.
- Rehearsal and anticipation improve memorization.
- Rehearsal has more impact than anticipation.

Method	bAbI	NarrativeQA	ActivityNet-QA
	(Mean Error)	(MRR)	(Acc)
DNC	16.70 ± 7.60	25.2	30.3
NUTM	5.60 ± 1.90	27.2	33.1
HMem	8.93 ± 0.73	25.5	31.9
DAM	1.53 ± 1.33	27.5	32.4
\mathbf{STM}	0.39 ± 0.18	26.7	33.7
\mathbf{CT}	0.81 ± 0.26	28.3	35.4
$\mathbf{R}\mathbf{M}$	0.33 ± 0.15	28.7	36.3
RAM	0.25 ± 0.16	30.9	37.4

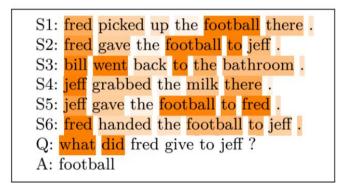
*Error rate: lower is better

Method	bAbI (Error Rate)	NarrativeQA (MRR)	ActivityNet-QA (Acc)
RAM	0.25 ± 0.16	30.9	37.4
RAM (random masking)	0.28 ± 0.19	30.6	37.4
RAM w/o $\mathcal{L}_{anticipation}$	0.31 ± 0.17	28.9	35.9
RAM w/o $\mathcal{L}_{rehearsal}$	0.34 ± 0.17	29.7	36.4
RAM w/o \mathcal{L}_{ssm}	0.37 ± 0.18	28.0	35.0

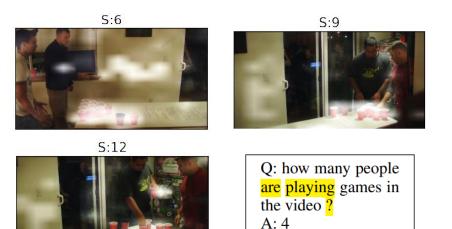


What is memory storing?:

- Memory slots capture information related to the entities and activities.
- Also, the relation between entities and objects.



• Memory behaves the same for visual information.





Conclusions

Anticipation and Memory Rehearsal are complementary and improve language modeling (an increase of ~11%).

- Predicting the future and rehearsing past elements improve memorization (an increase of ~16%).
- Coreference information plays an important role (an increase of ~4%).





8

Thank you

