Transformer-XH: Multi-Evidence Reasoning with eXtra Hop Attention

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Transformers in Modeling Natural Language

- Input sentence(s) in sequential form
- (New) standard solutions with pre-trained Transformers (e.g., BERT)



Transformer in Machine Reading

• (New) standard solutions with pre-trained Transformers (e.g., BERT)



Structured Text Sequence

Facebook was founded by <u>Mark</u> <u>Zuckerberg</u>, along with fellow <u>Harvard College</u> students and roommates.

Zuckerberg built a website called "Facemash" in 2003 while attending <u>Harvard</u> <u>University</u>. The site Harvard University is a private <u>Ivy League research</u> <u>university in Cambridge,</u> <u>Massachusetts</u>, with about 6,800 undergraduate

This Work

- Transformer-XH for *structured* text:
 - Transformer with eXtra Hop attentions
 - Global representations of **multiple connected** text pieces
- Strong performance on different multi-evidence reasoning tasks
 - Multi-hop QA (Hotpot QA)
 - Multi-evidence Fact Verification (FEVER)

Transformer-XH: In Sequence Attention



In sequence (τ) attention in layer l, token i:

$$h_{\tau,i}^{l} = \sum_{j} Softmax_{j}(\frac{q_{\tau,i}^{T} \cdot k_{\tau,j}}{\sqrt{d_{k}}}) \cdot v_{\tau,j}$$

Transformer-XH: Extra Hop Attention



In sequence (τ) attention in layer l, token i:

$$h_{\tau,i}^{l} = \sum_{j} Softmax_{j} \left(\frac{q_{\tau,i}^{T} \cdot k_{\tau,j}}{\sqrt{d_{k}}}\right) \cdot v_{\tau,j}$$

Extra hop attention between sequences $(\tau \rightarrow \eta)$:

$$\hat{h}_{\tau,0}^{l} = \sum_{\eta; e_{\tau\eta}=1} Softmax_{\eta} \left(\frac{\hat{q}_{\tau,0}^{T} \cdot \hat{k}_{\tau,0}}{\sqrt{d_{k}}}\right) \cdot \hat{v}_{\eta,0}$$

Transformer-XH: Layer Representation



In sequence (τ) attention in layer l, token i:

$$h_{\tau,i}^{l} = \sum_{j} Softmax_{j} \left(\frac{q_{\tau,i}^{T} \cdot k_{\tau,j}}{\sqrt{d_{k}}}\right) \cdot v_{\tau,j}$$

Extra hop attention between sequences $(\tau \rightarrow \eta)$:

$$\hat{h}_{\tau,0}^{l} = \sum_{\eta; e_{\tau\eta}=1} Softmax_{\eta} (\frac{\hat{q}_{\tau,0}^{T} \cdot \hat{k}_{\tau,0}}{\sqrt{d_{k}}}) \cdot \hat{v}_{\eta,0}$$

Combine two attentions:

$$\begin{split} \tilde{h}_{\tau,0}^{l} &= Linear(\left[h_{\tau,0}^{l} \circ \hat{h}_{\tau,0}^{l}\right]) \\ \tilde{h}_{\tau,l}^{l} &= h_{\tau,i}^{l}; \; \forall i \neq 0 \end{split}$$

Transformer-XH in Multi-hop QA

Input Question:

In which city was Facebook launched?



• Evidence graph construction

Transformer-XH in Multi-hop QA

Input Question:

In which city was Facebook launched?



- Evidence graph construction
- Transformer-XH

Transformer-XH in Multi-hop QA



- Evidence graph construction
- Transformer-XH
- Task specific Layers

Results on Hotpot QA Full-Wiki Test Set



Results on Hotpot QA Full-Wiki Test Set



Results on Hotpot QA Full-Wiki Test Set



Results on FEVER 1.0



Summary

- Transformer-XH: Simple but powerful adaption of Transformer for structured text
- It learns better representations via extra hop attention
- Strong performance on multiple tasks

Thanks!