

Building a Robust QA System for IID Squad Using Coattention, Self-Attention, and Answer Pointer Network

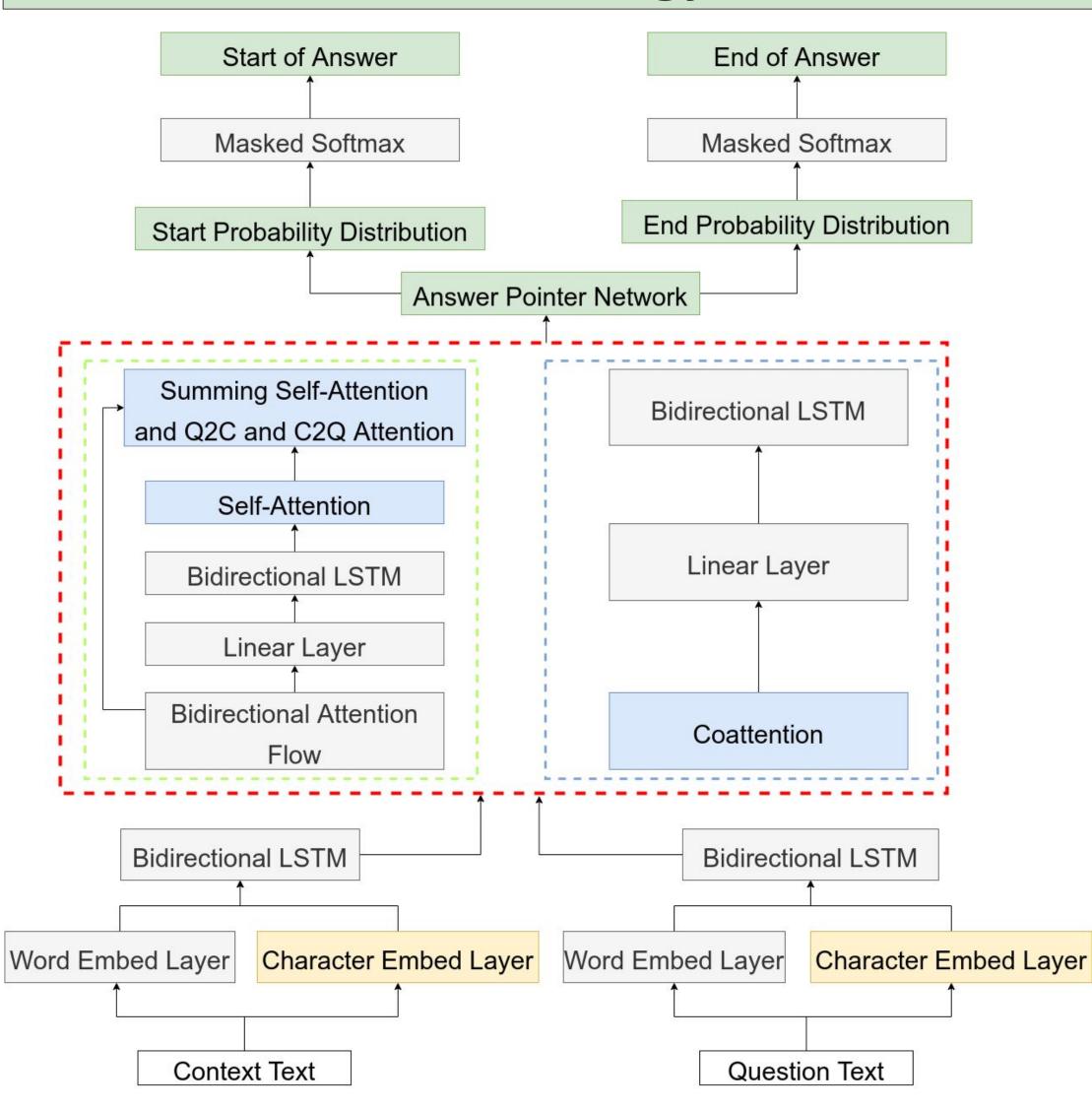
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Problem and Background

- Question answering: important as many NLP tasks can be formulated as question answering
- **SQuAD**: > 129k data points from Wikipedia
- Motivation: explore effects of implementing and modifying techniques to improve traditional BiDAF model
- Methods used:
 - Character-level embedding: consider morphology
 - \circ Attention: BiDAF \rightarrow coattention, self-attention
 - Answer pointer network: condition end predictions on start predictions
- Goal: implement different combinations of various techniques to improve baseline

Methodology

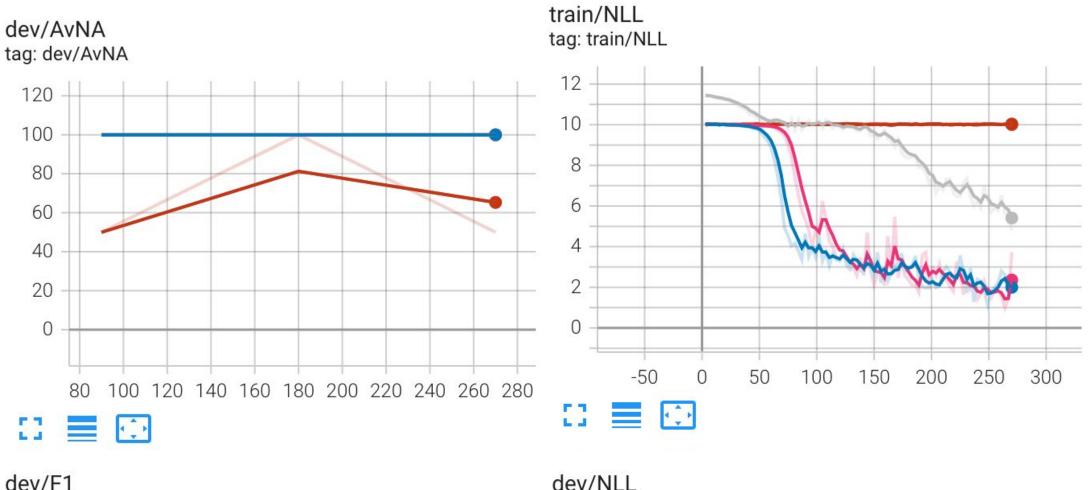


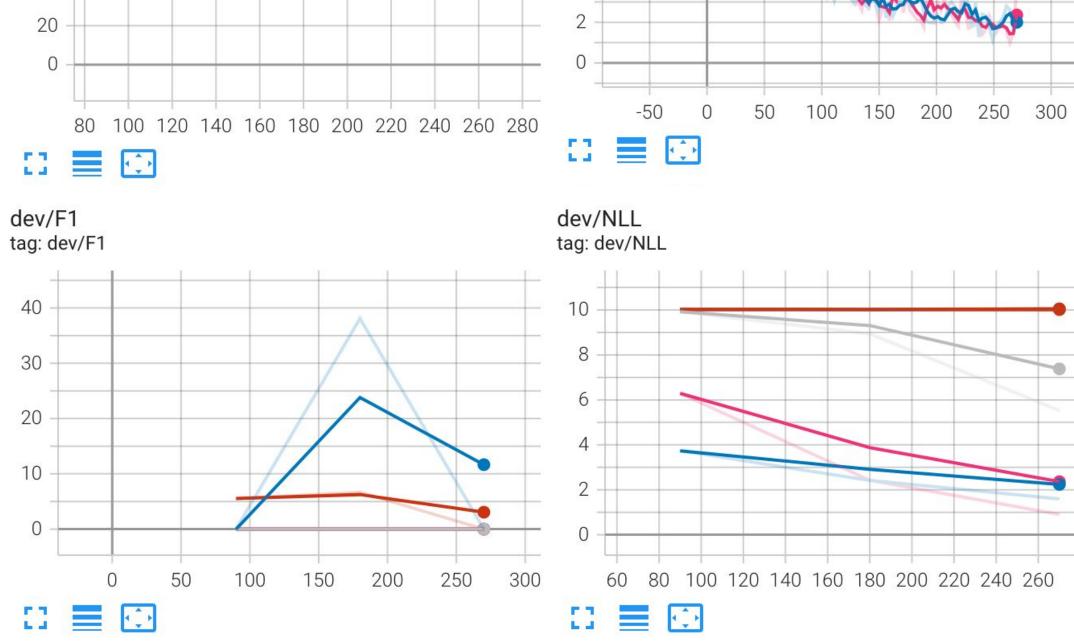
Abstract

- Combinations of various techniques aiming to improve performance in question answering on SQuAD dataset
- Best model is the combination of coattention, character embeddings, and answer pointer network
 - \circ EM = 52.20, F1 = 52.20 on training set
- No improvement yet, likely due to small training dataset used

Experiments

- 1. Baseline (baseline)
- 2. Character-level embedding (character)
- 3. Self-attention (selfattn)
- 4. Answer pointer (combined)
- 5. Self-attention + answer pointer + character (everything)
- 6. Coattention (coattention)
- train/baseline-02
 train/selfattn-01
 train/character-03
- train/combined-02
- train/everything-01
- train/coattention-01

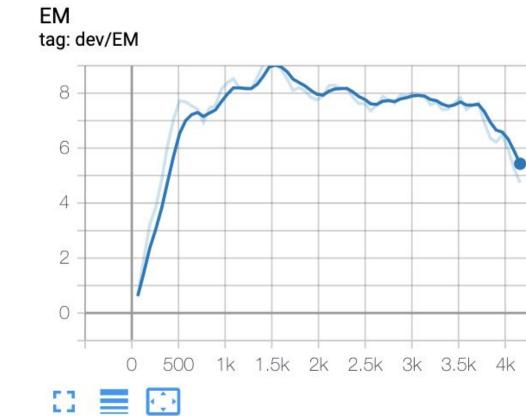




Results and Analysis

- Loss neared 0 at end of training 2e-4: overfit
- Reduced size of train
 dataset to ~1/10, increased
 # epochs (Colab's
 hardware limitations when
 Azure stopped working)
- May overfit due to training many epochs on small subset of train dataset

tag: dev/F1



character embedding

results (~5k)

ag: dev/NLL

12.5

11.5

11.5

6 10.5 10 10 0 500 1k 1.5k 2k 2.5k 3k 3.5k 4k 0 \$\equiv \text{LI} \equiv \text{LI} \text{LI} \text{LI} \text{LI} \text{LI}

Conclusion and Future Work

- Able to segment dataset and train different implementations of coattention, self-attention, character embeddings, and conditioning
- Identified the best combination of various techniques
- Learned how to read, understand, and implement NLP research papers; understanding large codebase; and using real-world datasets for interesting NLP applications
- With small training dataset, no performance improvement is found.
 - Future work: increase training size to better conclude performance

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