

SQuAD-ing Up as a Winning Team: Character Embeddings and **Dynamic Coattention Networks**



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Introduction/Background

- Question Answering is a task in the Natural Language Processing community that has received a lot of attention due to its broad range of implications for issues ranging from web search to data analytics.
- Question Answering has further increase in popularity due in part to the robustness and reasoning-based nature of the SQuAD
- We first build upon the given baseline BiDAF model that contains only word-level embeddings by incorporating the Character-Level Embeddings as described in Seo et al. [2]
- We additionally implement Dynamic Coattention and create a model combining character-level embeddings with the Dynamic Coattention Network described in Xiong et al.^[3]
- We measure the success of our implementations using EM and F1 scores and further experiment upon our models with
- hyperparameter tuning experiments.
 We find our model performs best with the Coattention and Character-Level Embeddings configuration with a learning rate of 0.6 and a dropout probability of 0.15 and associated EM and F1 scores of 59.189 and 62.787, respectively, on the test set.

Methods

1. Bi-directional Attention Flow (BiDAF) Model



2. Character-Level Embeddings

- Pass preprocessed context/query words through a CNN to produce char embeddings.
- Maxpool, concatenate to word embeddings, pass to Highway Network
- 3. Dynamic Coattention Networks (DCN)



Results

Model	lr	dropout	F1	EM
Baseline	0.5	0.2	60.74	57.39
Char-level Embs	0.6	0.2	60.86	58.11
Char-level Embs	0.5	0.25	61.92	58.47
Char-level Embs	0.6	0.15	62.27	59.32
Char-level Embs + Coattention	0.6	0.2	61.78	59.06
Char-level Embs + Coattention	0.2	0.3	57.31	54.17
Char-level Embs + Coattention	0.6	0.15	63.91	60.43

Fig 1: Hyperparameter Tuning on Dev Set

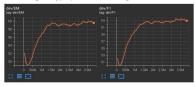


Fig 2: Best Model (Coattention + Char-level Embeddings + Hyperparameter Search) Performance

Model	F1	EM
Baseline on Dev	60.737	57.385
Char-Level Embs + Coattention + Tuned Parameter - Dev	63.914	60.427
Char-Level Embs + Coattention + Tuned Params - Test	62.704	59.138

Fig 3: Model Performance (EM and F1 scores)

Analysis



Model pulled more data than necessary from the context; however, the extra information was also relevant to the ground truth answer.



- Able to pull answer from context despite human labeling error

Conclusions

- Adding character-level embeddings to the baseline BiDAF increased performance.
- Substituting the attention layer in BiDAF with coattention marginally increased performance.
- We were able to improve and achieve higher accuracy after performing a hyperparameter search.
- Our best model achieved an F1 score of 62.704 and an EM score of 59.138 on the test set.
- For future improvements, we are curious to see what would happen with a feedforward model that separates question types.
- We also suggest implementing data augmentation such as Easy Data Augmentation or more types of attention such as self-attention.

References

[1] Pranav Rajpurkar, Robin Jia, and Percy Liang, Know what you don't know: Unanswerable questions for SQuAD. In Association for Computational Linguistics (ACL), 2018. [2] Mirijoon Seo, Aniruddha Kembhavi, Ali Farhadi, and Hannaneh Hajishirzi. Bidirectional attention flow for machine comprehension, 2016. [3] Caiming Xiong, Victor Zhong, and Richard Socher. Dynamic coattention networks for question answering. In arXiv, 2018