# Exploring Embedding and Attention Improvements to BiDAF SQuAD Model

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## **BACKGROUND**

### **Question Answering**

- Cloze-style reading comprehension task: Given a question and context paragraph, predict an answer to the question using span of text from the context.
- Stanford Question-Answering Dataset (SQuAD) is recent benchmark reading comprehension data-set. SQuAD 2.0 introduces unanswerable questions.

### **BiDAF**

- Bi-Directional Attention Flow (BiDAF) was early state-of-the-art SQuAD model.
- Attention layer allows both Context-to-Question attention and Question-to-Context attention to flow into modeling layer.

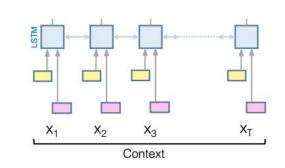
# **EXPERIMENTS**

- SQuAD metrics:
  - Exact Match (EM)
     Baseline: 55.49
  - o F1 (precision and recall) Baseline: 58.62
  - Answer vs No Answer (AvNA) Baseline: 64.68
- BiDAF + Character Embeddings tested against baseline
- Self-Attention tested varying input attention (context state, context-to-question attention, question-to-context attention)
- GRU tested with baseline and character-embeddings

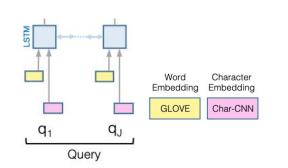
# **APPROACH AND MODELS**

### **BiDAF + Character Embeddings**

- Baseline BiDAF uses only word-level embeddings.
   Original BiDAF model combines word-level and character-level embeddings.
- (1) Embed character's into vectors
  - (2) Feed through CNN
  - (3) Max-pool output



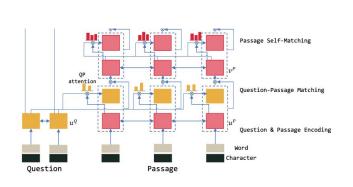
c2q + q2c (dark blue)



### **Self-Attention**

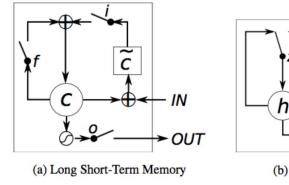
- Self-attention layer based on R-Net's "self-matching attention."
- For each question-aware context paragraph word representation compute attention pooling vector over whole context:

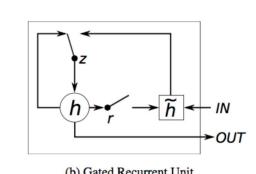
$$c_t = att(v^P, v_t^P)$$



### **BiDAF with GRU RNN**

- Baseline BiDAF and original BiDAF models both use a Bi-Directional Long-Short-Term (LSTM) RNN for encoder and modeling layer.
- R-Net uses Gated Recurrent Unit (GRU) RNN





# **RESULTS**

# Character Embeddings and GRU results. Baseline (pink), Character Embeddings (blue), Baseline with GRU (orange), Character Embeddings with GRU (red) dev/EM dev/EM dev/F1 dev/F1 dev/F1 self-Attention results. Baseline (light blue), context-to-question (c2q) (green), c + c2q (orange), c +

