

#### Introduction

- **Goal:** Build a system for question answering as well as identify unanswerable questions in the SQuAD 2.0 dataset.
- **Question answering:** Identifying a sequence in a context paragraph that correctly answers an input query.

#### Approach

- **Evaluation Metric:** F1 and EM scores on the Dev and Test set.
- **Data:** SQuAD 2.0, a dataset of over 150,000 questions, both answerable and unanswerable.
- Word Embeddings: Stanford's pretrained GLoVE embeddings.
- Experimental Details:
  - Adam Optimizer with learning rate warmup and weight decay.
  - Character-level embedding Ο ablation.
  - Gradient accumulation for Ο larger effective batch sizes.
  - Variation in number of Ο heads, hidden size, encoder repetition.
  - Stochastic Depth Dropout. Ο
  - Weight initialization ablation Ο testing.

#### Recurrency-Free Question Answering with **Unanswerable Questions** Michael Hazard, Ian Hodge, Daniel Semeniuta







#### **QANet Model:**

- Embedding Layer
- Embedding Encoder
- **Context-Query Attention Layer**
- Model Encoder Layer
- Output Layer

#### **References Cited**

Adams Wei Yu, David Dohan, Minh-Thang Luong, Rui Zhao, Kai Chen, Mohammad Norouzi, and Quoc V. Le. Qanet: Combining local convolution with global self-attention for reading comprehension. CoRR, abs/1804.09541, 2018. Minghao Hu, Furu Wei, Yuxing Peng, Zhen Huang, Nan Yang, and Ming Zhou. Read + verify: Machine reading comprehension with unanswerable questions. CoRR, abs/1808.05759, 2018.



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## Results

Model	F1 Score	EM
Baseline BiDAF	58	55
Tuned BiDAF	62.814	59.267
Tuned BiDAF + Char Embeddings	r 65.731	62.225
QANet	55.64	52.26
QANet + Char Embeddings	56.91	53.18

## Model

#### **Answer Verifier:**

- Embedding Layer
- Transformer Layer
- Output Layer
  - Option to pre-train as a language model

# **Challenges/Conclusions**

- The model was too large to fit onto any of the GPUs training took much longer than anticipated. This defeated the purported speed gains of the non-sequential architecture.
- and tuning very difficult. We often had to try multiple changes at once which made it difficult to narrow down our problems.
- We found the transformer architecture to be extremely fragile and small tunes to the hyper parameters made drastic positive and negative changes.
- Future work would include incorporating the verifier with our BiDAF and QANet models, with a possible ensemble approach.

## **Error Analysis**

• Misidentified Specifier: Question: What sort of motion did Newcomen's steam engine continuously produce? **Context:** In 1781 James Watt patented a steam engine that produced continuous rotary motion. Answer: N/A **Prediction:** rotary • Incorrect Answer Length: Question: What is the Chinese name for the Yuan dynasty? **Context:**The Yuan dynasty (Chinese:元朝; pinyin: Yuán Cháo), officially the Great Yuan (Chinese:大元; pinyin: Dà Yuán; Mongolian: Yehe Yuan Ulus[a])... Answer: Yuán Cháo **Prediction:**元朝; pinyin: Yuán Cháo), officially the Great Yuan

provided with a reasonable batch size. As a result,

The long training cycle (~12 hours) made debugging