



# Reading Wikipedia to Answer Open-Domain Questions

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## Overview

**Goal:** build an end-to-end question answering system that can use full Wikipedia to answer any factoid question.

Large-scale QA + Machine comprehension of Text  
"Machine Reading at Scale" (MRS)

Our system **DrQA:**

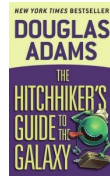
**Q** What is question answering? **A** a computer science discipline within the fields of information retrieval and natural language processing

**Q** Who was the winning pitcher in the 1956 World Series? **A** Don Larsen

**Q** What is the answer to life, the universe, and everything? **A** 42

Try it out yourself!

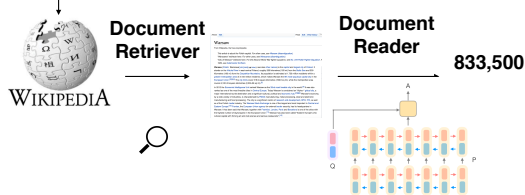
<https://github.com/facebookresearch/DrQA>



## Document Retriever + Document Reader

- Document retriever: finding relevant articles from 5 million Wikipedia articles
- Document reader (reading comprehension system): identifying the answer spans from those articles

Q: How many of Warsaw's inhabitants spoke Polish in 1933?



- Datasets:
  - SQuAD (Rajpurkar et al, 2016)
  - TREC (Baudiš and Šedivý, 2005)
  - WebQuestions  $\approx$  Freebase (Berant et al, 2013)
  - WikiMovies (Miller et al, 2016)

## Approach

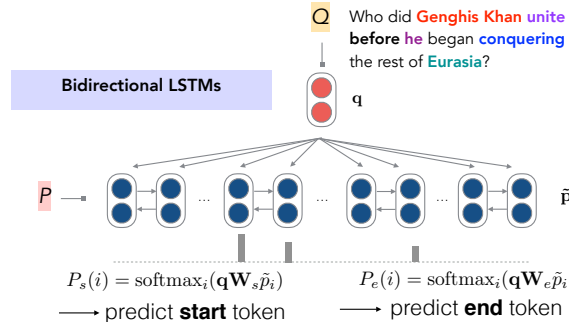
### Document Retriever

TF-IDF bag-of-words vectors + efficient bigram hashing (Weinberger et al., 2009)

### Document Reader

**Task:** given paragraph P and question Q, the goal is to find a span A in the paragraph which answers the question.

**Model:** similar to AttentiveReader (Hermann et al, 2015; Chen et al, 2016). We aim to keep it **simple!**



The input vectors consist of:

- Word embeddings
- Exact match features: whether the word appears in question
- Token features: POS, NER, term frequency
- Aligned question embedding

### Data: SQuAD + Distantly Supervised Data

(Q, A)  $\rightarrow$  (P, Q, A) if P is retrieved and A can be found in P

**Q:** What part of the atom did Chadwick discover? **WebQuestions**

**A:** neutron

### Atom

From Wikipedia, the free encyclopedia

The **atomic mass** of these isotopes varied by integer amounts, called the **whole number rule**.<sup>[23]</sup> The explanation for these different isotopes awaited the discovery of the **neutron**, an uncharged particle with a mass similar to the **proton**, by the physicist **James Chadwick** in 1932. Isotopes were then explained as elements with the same number of protons, but different numbers of neutrons within the nucleus.

## Results

### Finding Relevant Articles

|              | Wiki Search | unigram     | +bigram     |
|--------------|-------------|-------------|-------------|
| SQuAD        | 62.7        | 76.1        | <b>77.8</b> |
| TREC         | 81.0        | 85.2        | <b>86.0</b> |
| WebQuestions | 73.7        | <b>75.5</b> | 74.4        |
| WikiMovies   | 61.7        | 54.4        | <b>70.3</b> |

**70-86%** of questions we have that the answer segment appears in the **top 5 articles**

### Performance on SQuAD (single model, Feb 2017)

|                                | EM   | F1   |
|--------------------------------|------|------|
| Logistic regression            | 40.4 | 51.0 |
| Fine-Grained Gating (Carnegie) | 62.5 | 73.3 |
| Match-LSTM (Singapore)         | 64.7 | 73.7 |
| DCN (Salesforce)               | 66.2 | 75.9 |
| BiDAF (UW & Allen Institute)   | 68.0 | 77.3 |
| Ours                           | 70.7 | 79.4 |
| r-net (MSR Asia)               | 71.3 | 79.7 |
| State-of-the-art (July 2017)   | 75.7 | 83.5 |
| Human performance              | 82.3 | 91.2 |

Exact match features are important!

| Features                                | F1           |
|---|--------------|
| Full                                    | 78.8         |
| No $f_{token}$                          | 78.0 (-0.8)  |
| No $f_{exact\_match}$                   | 77.3 (-1.5)  |
| No $f_{aligned}$                        | 77.3 (-1.5)  |
| No $f_{aligned}$ and $f_{exact\_match}$ | 59.4 (-19.4) |

### Full Results

- Pre-trained SQuAD model
- SQuAD + fine-tuning on DS data
- Multi-task learning

Exact match (top-1 prediction)

