# HAS-QA: Hierarchical Answer Spans Model for Open-domain Question Answering

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## Background

#### **Search Engine**



What is Machine Learning





#### What is Machine Learning? - An Informed Definition - TechEmergence

https://www.techemergence.com/what-is-machine-learning/ ▼ 翻译此页 2018年10月29日 - We asked 6 machine learning experts (including machine learning "godfather" Dr. Yoshua Bengio) to define "Machine Learning" as simply as ...

#### Machine learning - Wikipedia

https://en.wikipedia.org/wiki/Machine learning ▼ 翻译此页

Machine learning (ML) is a field of artificial intelligence that uses statistical techniques to give computer systems the ability to "learn from data, without being ...

Active learning (machine ... · Boosting (machine learning) · Online machine learning

#### What is Machine Learning? A definition - Expert System

https://www.expertsystem.com/machine-learning-definition/ ▼ 翻译此页

Machine learning is an application of artificial intelligence (AI) that provides systems the ability to automatically learn and improve from experience without being ...

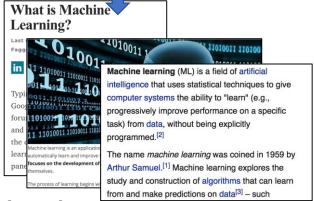
#### What is machine learning (ML)? - Definition from WhatIs.com

https://searchenterpriseai.techtarget.com/definition/machine-learning-ML ▼ 翻译此页 Machine learning (ML) is a category of algorithm that allows software applications to become more accurate in predicting outcomes without being explicitly ...

#### What is Machine Learning? - Introduction | Coursera

https://www.coursera.org/.../machine-learning/what-is-machine-learning-Uj... ▼ 翻译此页 Video created by Stanford University for the course "Machine Learning". Welcome to Machine Learning! In this module, we introduce the core idea of teaching a ...

#### Search



#### **Extract**

Machine learning (ML) is a field artificial intelligence that uses statistical techniques to give computer systems the ability to "Yearn" (e.g., progressively improve performance on a specific task from data, without being explicitly programmed.[2]

The name machine learning was coined in 1959 by Arthur Samuel. [1] Machine learning explores learn from and make predictions on data[3] - such the study and construction of algorithms that algorithms overcome following strictly static ram instructions by making data-driven predictions

Machine learning (ML) is a field of artificial intelligence that uses statistical techniques to give computer systems the ability to "learn" data, without being explicitly from programmed.

#### **Knowledge Base**



#### Machine learning

From Wikipedia, the free encyclopedia



**Problems** Supervised learning [show] (classification · regression) Clustering Dimensionality reduction Structured prediction [show] **Anomaly detection** [show] Artificial neural networks Reinforcement learning [show] Machine-learning venues [show] Glossary of artificial intelligence Related articles Machine learning portal

Machine learning intelligence that use computer systems t progressively impro task) from data, witl programmed.[2]

The name machine Arthur Samuel.[1] M study and construct from and make pred algorithms overcom instructions by mak decisions.[4]:2 throu inputs. Machine lea computing tasks wh explicit algorithms v or infeasible; exami filtering, detection o

Machine learning is overlaps with) comp focuses on prediction computers. It has st optimization, which

## I Outline

Open-domain Question Answering

Formulation & HAS-QA

Experiments

Conclusion & Going Forward

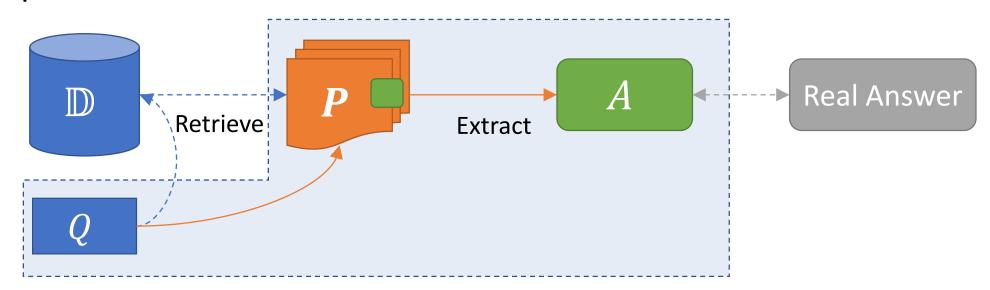
## I OpenQA - Open-domain Question Answering

### o Input:

- Q: real world question.
- D: huge *unstructured* text dataset, such as **web page corpus** or **Wikipedia.**

### Output:

• A: answer text from text dataset, which can answer the input question



## I Challenges

Sparsity of paragraphs contain answers **Uncertainty** of answer spans **Conflict** of span locations

## I Challenges - Sparsity

#### 新版五元人民币是什么花





百度一下

#### 第五套人民币上面的花卉是些什么花? 百度知道



7个回答 - 回答时间: 2018年9月27日 - 40人觉得有用

最佳答案: 1、1元纸币——兰花 象征意义 风姿绰约 幽香远溢 ——四君子之一,四季名花之司春使者,在...

更多关于新版五元人民币是什么花的问题>>

https://zhidao.baidu.com/quest... ▼ - 百度快照

#### 最新版人民币的正面分别是什么花?背面分别是哪些风景? 百人 道

2个回答 - 回答时间: 2016年6月23日 - 39人觉得有用

最佳答案: 最新版的 100:人民大会堂 50:布达拉宫 20:桂林山水 10:长江三峡 5:泰山 地西湖 十景之三潭印月 参考 第四套.即新版人民币,正面图案一是体现我国...

更多关于新版五元人民币是什么花的问题>>

https://zhidao.baidu.com/quest... ▼ - 百度快照

#### 人民币从一元钱到一百上面的图案都有花请问这些都是什么花?

1个回答 - 回答时间: 2008年04月26日

最佳答案: 第五套<mark>人民币</mark> 1元:水印和正面花团中的<mark>花都是</mark>兰花; 5元:水印和王面花团中的

是水仙; 10元:水印和正面花团中的花都是月季; 20元:水印和正面花团中 《是

https://iask.sina.com.cn/b/124... ▼ - 百度快照

#### 新版5元人民币背面是什么图??? 爱问知识人

举报 新版5元人民币 背面是什么图??? 自-浮生未 分享: 全部答案 ...安迪出生的地 城,安迪的妈妈是当地有名的花癫,因为疯了以后喜欢穿...

https://iask.sina.com.cn/b/iR2... ▼ - 百度快照

#### 这张5元人民币,一定要记清楚,千万不要随便花!



2018年5月22日 - 2002年11月18日中国人民银行发行了9° 币,虽然并不是1999年发行的,但是确定是属于99版人F 中... baijiahao.baidu.com/s?... ▼ - 百度快照

#### 五元人民币上的水印花是什么花?\_百度知道

1个回答 - 回答时间: 2016年6月16日

更多关于新版五元人民币是什么花的问题>>

https://zhidao.baidu.com/quest... ▼ - 百度快照

#### 5元人民币有两个版本,出现这种号码,一定不要花掉!



2018年8月9日 - 第五套人民币是在1999年开始发行的,而面的 元的的人民币也是在这个时期发行的,但是在第五套人民币中 https://baijiahao.baidu.com/s?... ▼ - 百度快照

## I Challenges - Uncertainty

#### 新版五元人民币是什么花





百度一下



## I Challenges - Conflict

5元纸币 — 水仙 象征意义 <u>叶姿</u>秀美; 亭亭玉立,雅号"<u>凌波仙子</u>",深

受国人喜爱。

每到农历春节,家家户户的厅堂中都要摆上一盆水仙花,黄蕊白被如金

银的水仙总是会捎给人们新一年的喜气与财运。

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## I Probabilistic Model – Target

Target:

Answer Probability:  $P(A|Q,\mathbf{P})$  Question Retrieved Paragraphs

### o Interpret:

The probability of the text A is the answer text, given the question
 Q and retrieved paragraphs P.

Implement: HAS-QA Model

## I Probabilistic Model – Challenge 1

### ○ Challenge 1:

Many paragraphs without the answer span are included in the data collection.

Answer Probability: 
$$P(A|Q,\mathbf{P}) = \sum_{i=1}^K P(P_i|Q,\mathbf{P})P(A|Q,P_i);$$
 Conditional Answer Probability

Paragraph Probability

### o Interpret:

- Paragraph Probability: The probability of the paragraph  $P_i$  that contains answer text, given the question Q and retrieved paragraphs P.
- Conditional Answer Probability: The probability of the text A is the answer text, given the question Q and paragraph  $P_i$ .

Implement: Paragraph Quality Estimator

## I Probabilistic Model – Challenge 2

### ○ Challenge 2:

Multiple answer spans may exist within one given paragraph.

```
Conditional Answer P(A|Q,P_i):=\mathcal{F}(\{P(L_j(A)|Q,P_i)\}_j), Probability: j\in[1,|\mathcal{L}(A,P_i)|]; Span Probability
```

### o Interpret:

- Span Probability: The probability of the text span  $L_j(A)$  is the answer span, given the question Q and paragraph  $P_i$ .
- ullet  ${\mathcal F}$  is an aggregation function, such as MAX, SUM.
- There exist  $|\mathcal{L}(A, P_i)|$  answer spans in paragraph  $P_i$ .

Implement: Multiple Span Aggregator

## I Probabilistic Model – Challenge 3

### Challenge 3:

The end position of an answer span is dependent with the start position.

Span Probability: 
$$P(L_j(A)|Q,P_i) = P(L_j^s(A)|Q,P_i) \longrightarrow$$
 Location Start Probability  $\cdot P(L_j^e(A)|Q,P_i,L_j^s(A)).$ 

#### o Interpret:

- Location Start Probability: The probability of the text span start location  $L_i^s(A)$  is the answer span location.
- Location End Probability: The probability of the text span end location  $L_j^e(A)$  is the answer span location, given start location  $L_j^s(A)$ .

Location End Probability

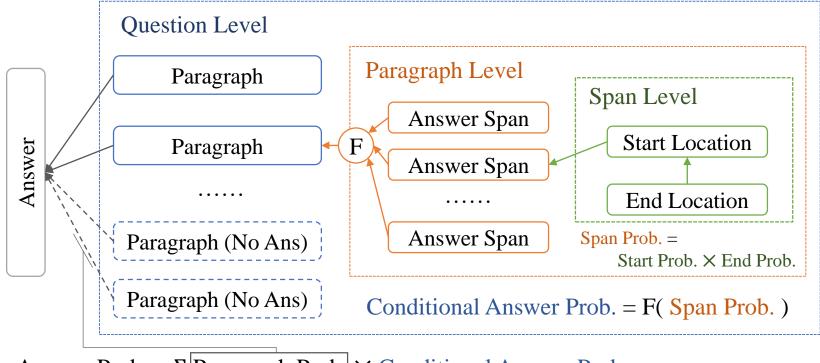
Implement: Conditional Span Predictor

## I HAS-QA Model – OpenQA Task

### Target:

Answer Probability:  $P(A|Q,\mathbf{P})$ 

### ∘ Model:



Answer Prob. =  $\Sigma$  Paragraph Prob.  $\times$  Conditional Answer Prob.

## I HAS-QA Model – RC Task

### For a Reading Comprehension Task

$$\circ$$
 Object:  $P(A|Q,P^+)$ 

o Model: 
$$P(A|Q,P^+) := P(L(A)|Q,P^+);$$
 
$$P(L(A)|Q,P^+) = P(L^s(A)|Q,P^+) \cdot P(L^e(A)|Q,P^+).$$

### **RC** ⊆ OpenQA

- 1. One paragraph
- 2. One answer span
- 3. Independence start and end location

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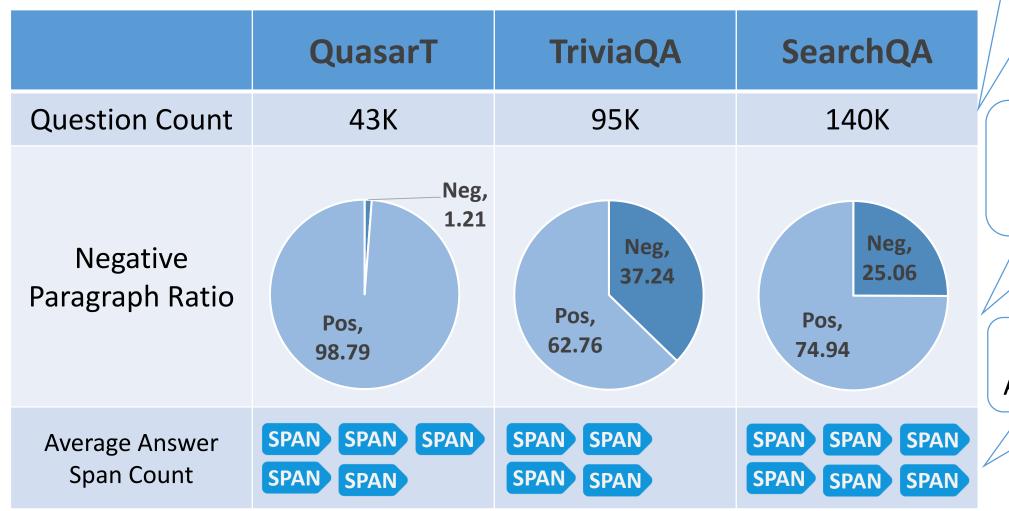
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### Dataset

### Three OpenQA Dataset

Large Dataset



Noisy Paragraphs

Multiple Answer Spans

### Experiments

	QuasarT		TriviaQA		SearchQA	
Model	EM	F1	EM	F1	EM	F1
GA (Dhingra et al., 2017a)	0.264	0.264	-	-	-	-
BiDAF (Seo et al., 2016)	0.259	0.285	0.411	0.474	0.286	0.346
AQA (Buck et al., 2017)	-	-	-	-	0.387	0.456
DrQA (Chen et al., 2017)	0.377	0.445	0.323	0.383	0.419	0.487
R <sup>3</sup> (Wang et al., 2017a)	0.353	0.417	0.473	0.537	0.490	0.553
Shared-Norm (Clark and Gardner, 2017)	0.386	0.454	0.613	0.672	0.598	0.671
HAS-QA (MAX Ans. Span)	0.432	0.489	0.636	0.689	0.627	0.687

- 1) HAS-QA **outperforms** traditional RC baselines with a large gap, such as GA, BiDAF, AQA listed in the first part.
- 2) HAS-QA **outperforms** recent OpenQA baselines, such as DrQA, R3 and Shared-Norm listed in the second part.

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## I Conclusion

 A new probabilistic formulation of OpenQA, based on a three-level hierarchical structure, i.e., the question level, the paragraph level and the answer span level.

#### HAS-QA Model

- 1) a paragraph quality estimator makes it robust for the paragraphs without answer spans
- 2) a multiple span aggregator points out that it is necessary to combine the contributions of multiple answer spans in a paragraph
- 3) a conditional span predictor is proposed to model the dependence between the start and end positions of each answer span.

## I Going Forward



Single Answer Spans Extraction



Multiple Answer Spans Extraction



Well-formed Answers Generation

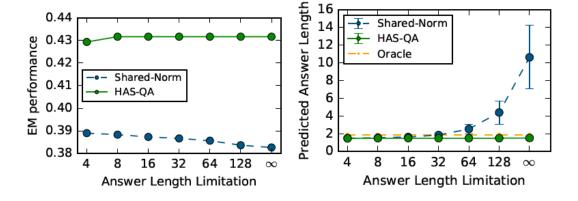


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## **Experimental Analysis**

#### Effects of Conditional Pointer Networks

- 1. The performance of Shared-Norm **decreases** when removing the answer length limitation, while the performance of HAS-QA first **increases** then becomes stable.
- The average predicted answer length increases in Shared-Norm when removing the answer length limitation. However, HASQA stably keeps the about 1.8 average words, where the oracle average answer length is about 1.9 words.



Example:
About Celebrating the contributions of Louis Braille January 5th, 2009
On the 200th anniversary of Louis Braille 's birth, people around the world are saluting a man whose tactile alphabet has provided a lifeline to people with impaired vision.

Shared-Norm HAS-QA

## **Experimental Analysis**

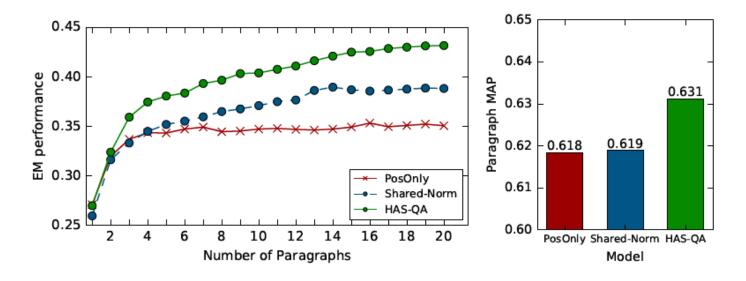
Effects of Multiple Spans Aggregation

Model	EM	F1
HAS-QA (HEAD Ans. Span)	0.372	0.425
HAS-QA (RAND Ans. Span)	0.341	0.394
HAS-QA (SUM Ans. Span)	0.423	0.484
HAS-QA (MAX Ans. Span)	0.432	0.489

- 1. SUM and MAX operations. They take advantages of using multiple answer spans for training and improve about 6% 10% in EM comparing to the HEAD operation.
- 2. The failure of **RAND operation**, mainly comes down to the conflicting training samples.

## **Experimental Analysis**

Effects of Paragraph Quality



- 1. With the increasing number of given paragraphs which ordered by the rank of search engine, EM performance of HAS-QA sustainably grows.
- 2. The Mean Average Precision (MAP) score between the predicted scores and the label whether a paragraph contains answer spans, shows that HAS-QA can rank **the high quality paragraphs** in the front of the given paragraph list.