

## Unsupervised Question Decomposition

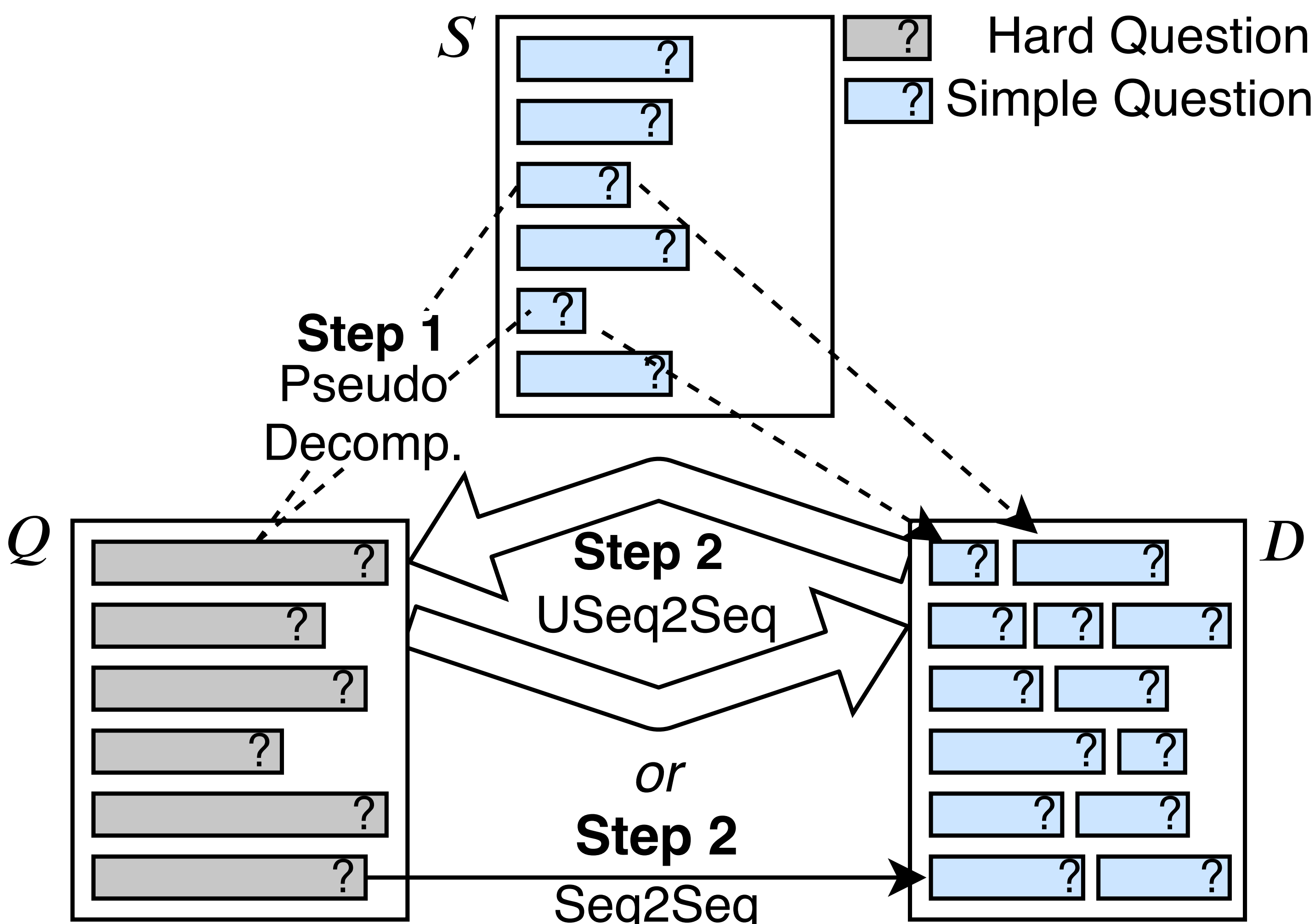


Figure 1

**Goal:** Improve QA by decomposing hard questions into easier sub-questions that existing QA systems can answer.

**Problem:** Prior work learns to decompose questions by relying on human annotation and extractive heuristics.

**Solution:** Decompose questions with *unsupervised methods*, using 2 stages (Figure 1):

- (1) Construct a noisy, “pseudo-decomposition” for each hard question by retrieving relevant sub-question candidates.
- (2) Train neural text generation models on that data with standard or unsupervised sequence-to-sequence learning.

**Finding:** We greatly improve multi-hop QA on HotpotQA with unsupervised decompositions, using a 3-stage method (Figure 2):

- (1) Generate single-hop sub-questions for a multi-hop question.
- (2) Answer sub-questions with a single-hop QA model.
- (3) Add sub-questions and their answers as additional input for a multi-hop QA model.

## Unsupervised Decomposition

### Creating Pseudo-Decompositions

For each  $q$  in a corpus  $Q$  of hard questions, we construct a pseudo-decomposition  $d' = [s_1; s_2; \dots s_N]$  by retrieving  $s$  from a corpus  $S$  of simple questions. We want  $s$  that are (1) similar to  $q$  w.r.t. a metric  $f$  (e.g., cos distance) and (2) maximally diverse:

$$d'^* = \operatorname{argmax}_{d' \subset S} \sum_{s_i \in d'} f(q, s_i) - \sum_{\substack{s_i, s_j \in d' \\ s_i \neq s_j}} f(s_i, s_j)$$

We embed  $q$  and  $s$  via sum-of-FastText word vectors. We also test random pseudo-decompositions where  $s_i \sim S$ .

### Training Models on Pseudo-Decompositions

We train models on pseudo-decompositions via:

- **No Learning:** Use  $d' = [s_1; s_2; \dots s_N]$  as sub-questions
- **Seq2Seq:** maximize  $\log p(d'|q)$
- **Unsup. Seq2Seq:** learn a  $q \rightarrow d$  mapping without training on noisy  $(q, d')$  pairs, similar to unsupervised translation

### QA Results

We greatly improve the baseline by adding sub-questions and answers.

We are competitive with DecompRC, SAE, and HGN which use strong supervision.

Decomp. Method	Pseudo-Decomps.	HOTPOTQA F1		
		Orig	MultiHop	OOD
$\times$	$\times$ (1hop)	66.7	63.7	66.5
$\times$	$\times$ (Baseline)	77.0 $\pm$ .2	65.2 $\pm$ .2	67.1 $\pm$ .5
No Learn	Random	78.4 $\pm$ .2	70.9 $\pm$ .2	70.7 $\pm$ .4
	FastText	78.9 $\pm$ .2	72.4 $\pm$ .1	72.0 $\pm$ .1
Seq2Seq	Random	77.7 $\pm$ .2	69.4 $\pm$ .3	70.0 $\pm$ .7
	FastText	78.9 $\pm$ .2	73.1 $\pm$ .2	73.0 $\pm$ .3
USeq2Seq	Random	79.8 $\pm$ .1	76.0 $\pm$ .2	76.5 $\pm$ .2
	FastText	<b>80.1<math>\pm</math>.2</b>	<b>76.2<math>\pm</math>.1</b>	<b>77.1<math>\pm</math>.1</b>
DecompRC*		79.8 $\pm$ .2	76.3 $\pm$ .4	77.7 $\pm$ .2
SAE (Tu et al., 2020) †		80.2	61.1	62.6
HGN (Fang et al., 2019) †		82.2	78.9 $\ddagger$	76.1 $\ddagger$
		Ours	SAE†	HGN†
Test (EM/F1)		66.33/79.34	66.92/79.62	69.22/82.19

## Using Decompositions in Question Answering (QA)

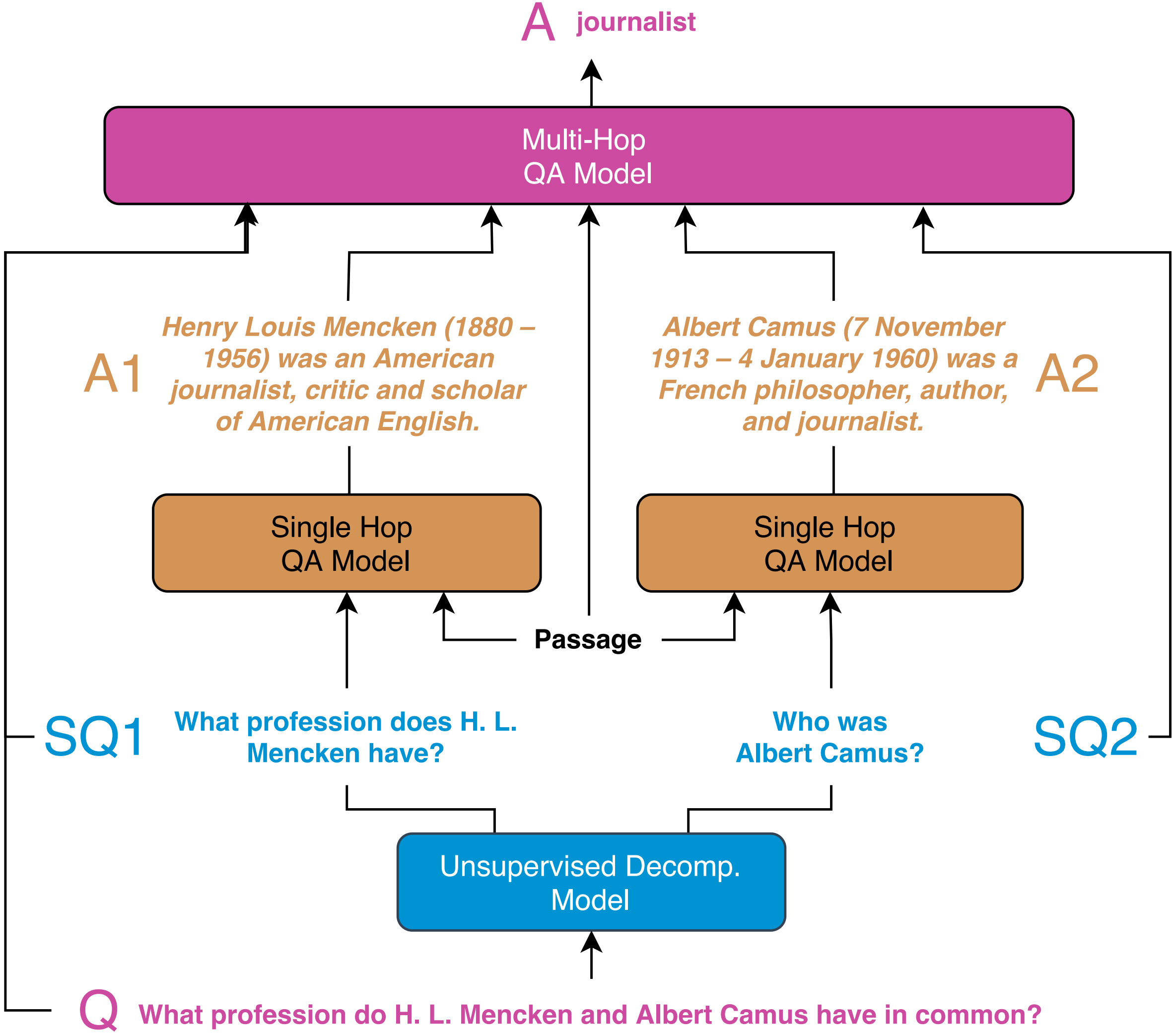


Figure 2

### Examples

Generated sub-questions are single-hop and question-relevant.

Add interpretability to black-box QA models.

Automatically learned to decompose many kinds of questions. Improved QA across all 4 question categories (Table 1).

Sub-questions are fluent, especially w.r.t. supervised decompositions (Table 2).

<b>Q1:</b> Are both Coldplay and Pierre Bouvier from the same country?
<b>SQ<sub>1</sub>:</b> Where are Coldplay and Coldplay from?
└ Coldplay are a <u>British</u> rock band formed in 1996 by lead vocalist and keyboardist Chris Martin and lead guitarist Jonny Buckland at University College London (UCL).
<b>SQ<sub>2</sub>:</b> What country is Pierre Bouvier from?
└ Pierre Charles Bouvier (born 9 May 1979) is a <u>Canadian</u> singer, songwriter, musician, composer and actor who is best known as the lead singer and guitarist of the rock band Simple Plan.
<b>Â:</b> No
<b>Q2:</b> How many copies of Roald Dahl’s variation on a popular anecdote sold?
<b>SQ<sub>1</sub>:</b> How many copies of Roald Dahl’s?
└ His books have sold more than <u>250 million</u> copies worldwide.
<b>SQ<sub>2</sub>:</b> What is the name of the variation on a popular anecdote?
└ “Mrs. Bixby and the Colonel’s Coat” is a short story by Roald Dahl that first appeared in the 1959 issue of Nugget.
<b>Â:</b> more than 250 million
<b>Q3:</b> Who is older, Annie Morton or Terry Richardson?
<b>SQ<sub>1</sub>:</b> Who is Annie Morton?
└ Annie Morton (born October 8, 1970) is an <u>American</u> model born in Pennsylvania.
<b>SQ<sub>2</sub>:</b> When was Terry Richardson born?
└ Kenton Terry Richardson (born 26 July 1999) is an English professional footballer who plays as a defender for League Two side Hartlepool United.
<b>Â:</b> Annie Morton

Decomps.	Bridge	Comp.	Intersec.	Single-hop	Decomp. Method	GPT2 NLL	% Well-Formed
$\times$	80.7 $\pm$ .2	73.8 $\pm$ .4	78.1 $\pm$ .6	73.8 $\pm$ .6	USeq2Seq	5.56	60.9
$\checkmark$	<b>82.3<math>\pm</math>.4</b>	<b>80.1<math>\pm</math>.3</b>	<b>81.2<math>\pm</math>.4</b>	<b>76.7<math>\pm</math>.6</b>	DecompRC	6.04	32.6

Table 1: QA F1 with and without Decompositions

Table 2: Decompositions from USeq2Seq (ours) vs. DecompRC

### Analysis

Including sub-answers is crucial. Returning sentences with sub-answer spans is better than just returning sub-answer spans.

Multi-hop QA improves when the single-hop QA model answers with gold, question-relevant “supporting fact” sentences. We find supporting facts without strong supervision.

Multi-hop QA improves when the single-hop QA model is more confident of its answers to sub-questions. Low confidence sub-answers may be more likely to be incorrect/hurt multi-hop QA.

