QALF: CONVERSATIONALLY BUILDING A RELATION EXTRACTOR IN 10 MINUTES VIA QUESTION ANSWERING

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HIGHER BANDWIDTH SUPERVISION

Traditional Supervision



One bit of information per input... (a binary label)

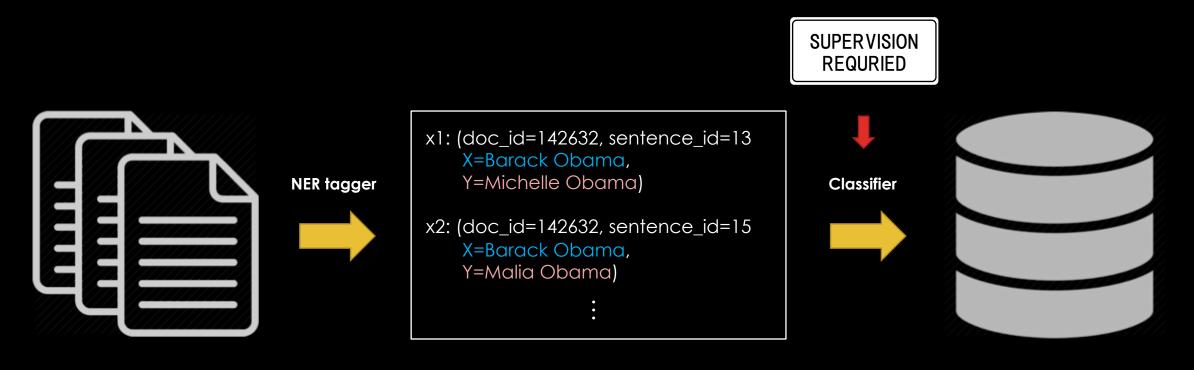
Weak Supervision



Many bits of information per input... (very many weak labels)

EXAMPLE APPLICATION: KBC

GOAL: Build a knowledge base of spouses mentioned in the news.



Documents

Candidates

Knowledge Base

LABELING FUNCTIONS (LFs)

Candidate (x1)

"Barack Obama and his wife Michelle attended their daughter's graduation hand-in-hand."

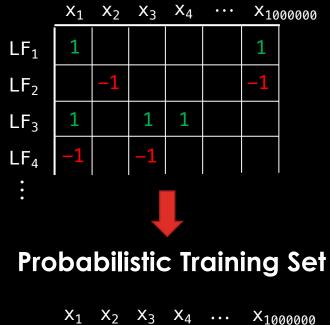
Labeling Functions

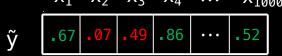
```
def LF1(x):
    return 1 if (re.search("his wife"), x.between)
    else 0
```

```
def LF2(x):
    return -1 if last_name(x.X) != last_name(x.Y)
    else 0
```

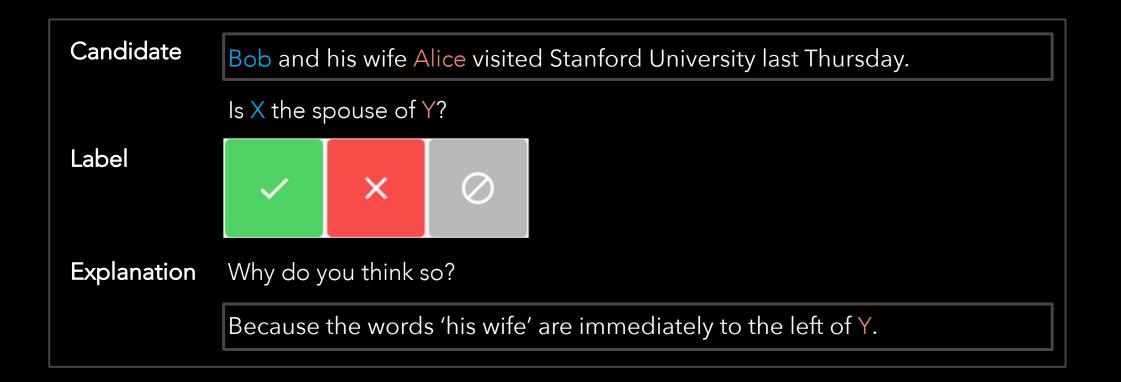
```
def LF3(x):
    return 1 if (x.X, x.Y) in set(known_spouses)
    else 0
```

Noisy Labels

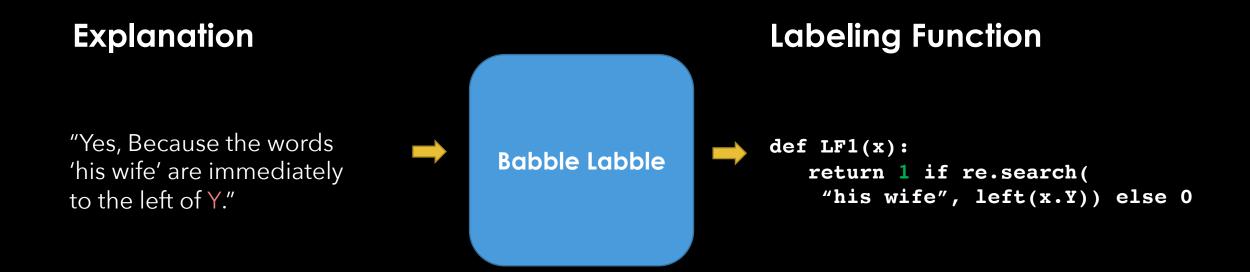




PRIOR WORK: BABBLE LABBLE



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Result

We achieved the same F1 score with classifiers trained using up to $100 \times$ fewer inputs when users provide explanations instead of labels!

"On Bob and Alice's honeymoon, it rained all week."



Prefers Semantics (e.g., "it says so") Yes, because it says they went on a honeymoon together.

Do you mean "because the word 'honeymoon' occurs within 10 characters to the right of person1 or person 2?"

Do you think X is the spouse of Y? Why?

Prefers Syntax (e.g., keywords, word distance, capitalization, etc.)

RAISING THE LEVEL OF ABSTRACTION

QALF

Supervise at the level of *semantics* with language

Babble Labble

Snorkel

Traditional

Supervise at the level of syntax with language

Supervise at the level of syntax with code

Supervise at the level of **examples** with **labels**

QA MODELS

Document

Santa Cruz (Spanish: Holy Cross) is the county seat and largest city of Santa Cruz County, California. As of 2013 the U.S. Census Bureau estimated Santa Cruz's population at 62,864.

Santa Cruz is known for its moderate climate, natural environment, coastline, redwood forests, alternative community lifestyles, and socially liberal leanings. It is also home to the University of California, Santa Cruz, a premier research institution and educational hub, as well as the Santa Cruz Beach Boardwalk, an oceanfront amusement park operating continuously since 1907.

Questions

What state is Santa Cruz in?

What university is located in Santa Cruz?

Answers

California (char. 92-101)

University of California, Santa Cruz (char. 366-401)

When did the Santa Cruz Beach Boardwalk first open? 1907 (char. 557-561)

Strengths

- Implicit typing (Who -> people, Where -> places, etc.)
- Improved robustness to paraphrases via word embeddings
- Generic relation recognizer ("X is the R of Y" \rightarrow (X, Y) \in R)

SQuAD				
The Stanford Question Answering Dataset				
Rank	Model	EM	F1	
	Human Performance Stanford University (Rajpurkar et al. '16)	82.304	91.221	
1 Jan 22, 2018	Hybrid AoA Reader (ensemble) Joint Laboratory of HIT and iFLYTEK Research	82.482	89.281	
1 Mar 06, 2018	QANet (ensemble) Google Brain & CMU	82.744	89.045	
1 Feb 19, 2018	Reinforced Mnemonic Reader + A2D (ensemble model) Microsoft Research Asia & NUDT	82.849	88.764	
2 Feb 02, 2018	Reinforced Mnemonic Reader (ensemble model) NUDT and Fudan University https://arxiv.org/abs/1705.02798	82.283	88.533	
2 Jan 03, 2018	r-net+ (ensemble) Microsoft Research Asia	82.650	88.493	
2 Jan 05, 2018	SLQA+ (ensemble) Alibaba iDST NLP	82.440	88.607	
3 Dec 17, 2017	r-net (ensemble) Microsoft Research Asia http://aka.ms/rnet	82.136	88.126	
3 Dec 22, 2017	AttentionReader+ (ensemble) Tencent DPDAC NLP	81.790	88.163	
4 Feb 27, 2018	QANet (single model) Google Brain & CMU	80.929	87.773	
4 Nov 17, 2017	BiDAF + Self Attention + ELMo (ensemble) Allen Institute for Artificial Intelligence	81.003	87.432	

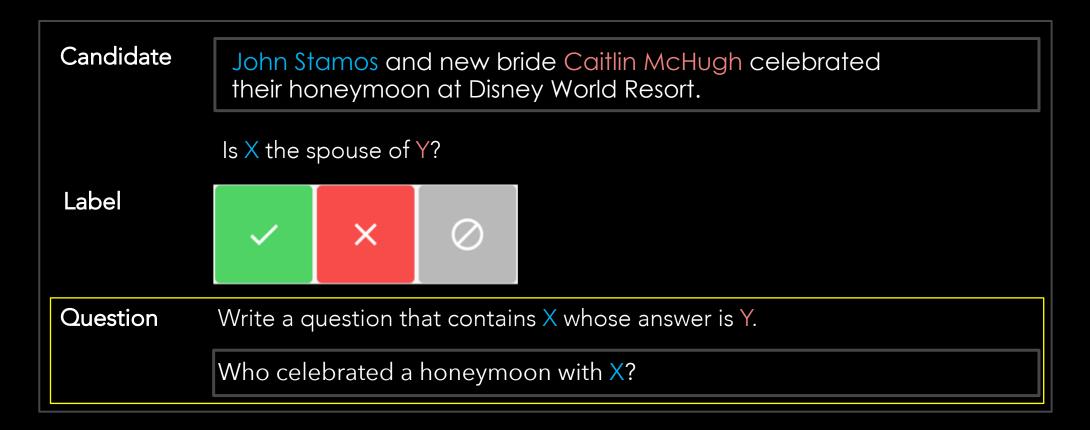
THE QA ARMS RACE

SQuAD Leaderboard

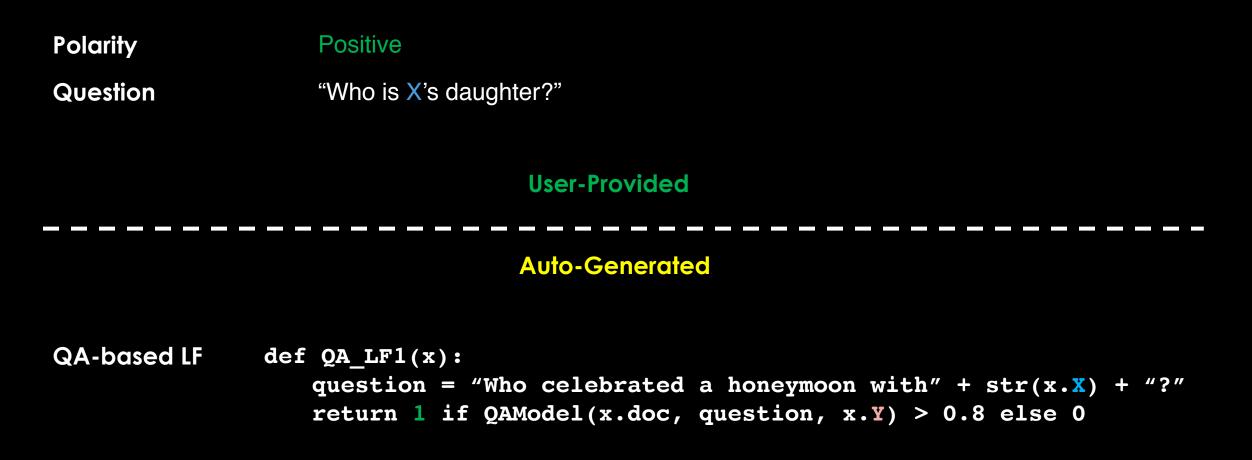
- 17 of the top 20 scores are from 2018
- Leaderboard populated with top companies and universities around the world



QALF INTERFACE



MAKING QA-LFs



APPLYING QA-LFs

Intuition

If the QA Model is very confident that Y is the answer to the question based on X, return a label of the user-provided polarity; otherwise, abstain.

QA-based LF	<pre>def QA_LF1(x):</pre>
	question = "Who celebrated a honeymoon with" + $str(x.X)$ + "?"
	return 1 if QAModel(x.doc, question, $x.Y$) > 0.8 else 0

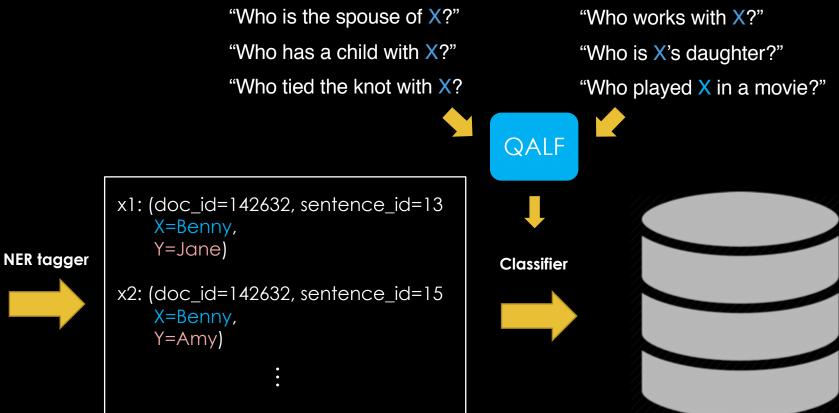
Candidates

x1 (Aaron, Alice) QAModel(x1.doc, "Who celebrated a honeymoon with Aaron?", Alice)
x2 (Billy, Betty) QAModel(x2.doc, "Who celebrated a honeymoon with Billy?", Betty)
x3 (Chris, Carol) QAModel(x3.doc, "Who celebrated a honeymoon with Chris?", Carol)

MAKING QA-LFs

(-) Negative Questions

(+) Positive Questions



= Possible X

"Benny promised Jane he wouldn't miss their daughter Amy's recital this year."

"Molly Brown and Bob Johnson went over their notes on the way to the meeting."

"Jenny and Dave, her boyfriend of 10 years, finally tied the knot this weekend."



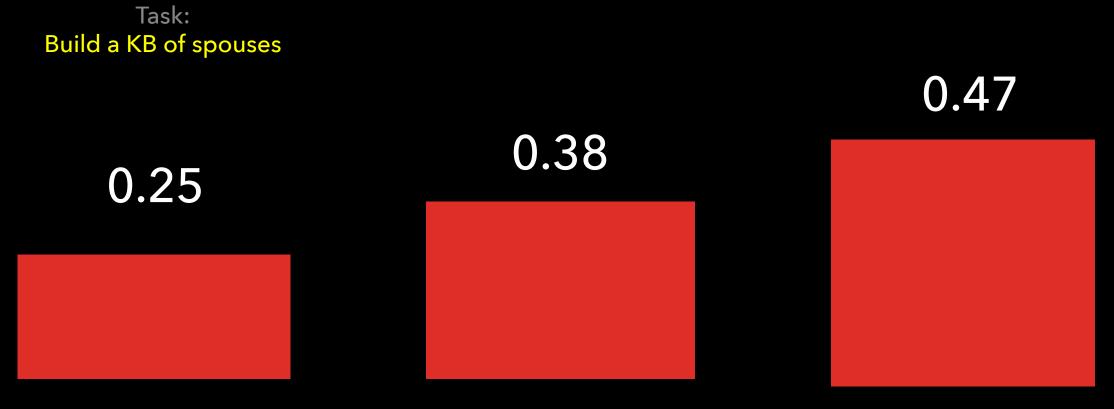
Documents

Candidates

Knowledge Base

EXPERIMENTAL RESULTS

(F1 scores)



1 Question

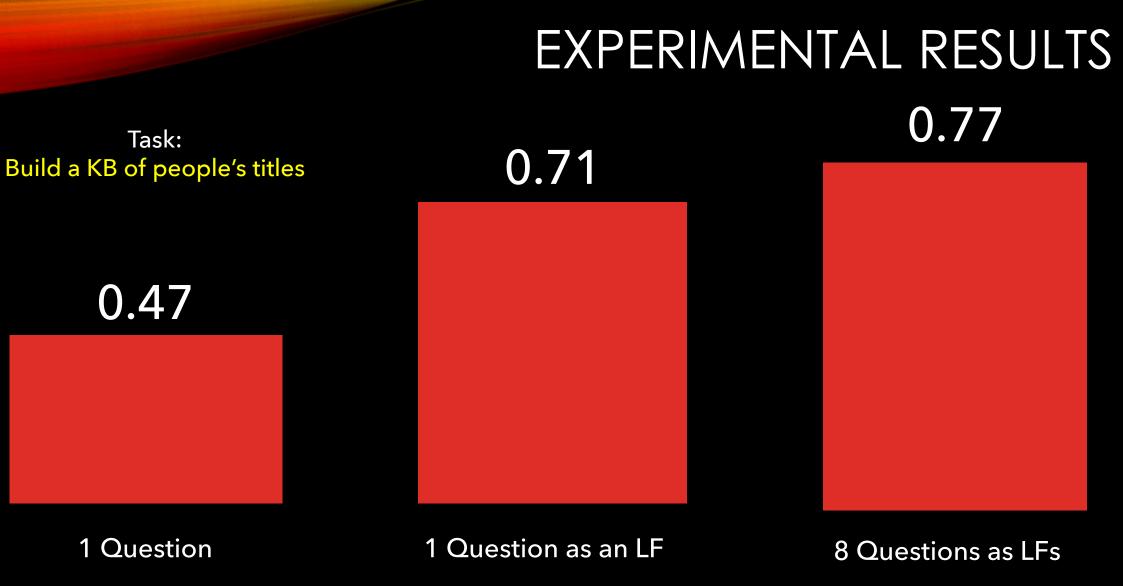
Ask the question on the test set "Who is the spouse of X?"

1 Question as an LF

Use the question to make a training set from unlabeled data

12 Questions as LFs

Aggregate the noisy labels from multiple LFs



Ask the question on the test set "What is X's title?" Use the question to make a training set from unlabeled data

Aggregate the noisy labels from multiple LFs

CONCLUSION

• LFs allow higher bandwidth supervision

• QA LFs allow semantic-level supervision

 Initial results suggest big wins (up to 30 F1 points) over direct QA model use



