Iterative Search for Weakly Supervised Semantic Parsing











Ed Hovy

Pradeep Dasigi

Matt Gardner

Shikhar Murty

Luke Zettlemoyer









This talk in one slide

- Training semantic parsing with denotation-only supervision is challenging because of **spuriousness**: incorrect logical forms can yield correct denotations.
- Two solutions:
 - Iterative training: Online search with initialization \Rightarrow MML over offline search output
 - Coverage during online search
- State-of-the-art single model performances:
 - WikiTableQuestions with comparable supervision
 - NLVR semantic parsing with significantly less supervision



Semantic Parsing for Question Answering

Athlete	Nation	Olympics	Medals	
Gillis Grafström	Sweden (SWE)	1920–1932	4	
Kim Soo-Nyung	South Korea (KOR)	1988-2000	6	
Evgeni Plushenko	Russia (RUS)	2002–2014	4	
Kim Yu-na	South Korea (KOR)	2010-2014	2	
Patrick Chan	Canada (CAN)	2014	2	

Question: Which athlete was from South Korea after the year 2010?

Answer: Kim Yu-Na

Reasoning:

- 1) Get rows where Nation is South Korea
- 2) Filter rows where value in *Olympics > 2010*.
- 3) Get value from *Athlete* column

Program:

```
(select_string
  (filter
      (filter, all_rows olympics 2010)
      south_korea)
  athlete)
```



Weakly Supervised Semantic Parsing

 $\mathbf{x_i}$: Which athlete was from South Korea after 2010?

y; (select_string (filter_ (filter_ all_rows olympics 2010) south korea) athlete)

 $\mathbf{z}_{i^{\bullet}}$ Kim Yu-Na

MethleteNationOlympicsMedalsWith Yu-naSouth Korea2010–20142Tenley
AlbrightUnited
States1952-19562

Train on
$$D=\{x_i,w_i,z_i\}_{i=1}^N$$

Test: Given x_{N+k}, w_{N+k} find y_{N+k} such that $[\![y_{N+k}]\!]^{w_{N+k}} = z_{N+k}$



Challenge: Spurious logical forms

Athlete	Nation	Olympics	Medals	
Gillis Grafström	Sweden (SWE)	1920–1932	4	
Kim Soo-Nyung	South Korea (KOR)	1988-2000	6	
Evgeni Plushenko	Russia (RUS)	2002–2014	4	
Kim Yu-na	South Korea (KOR)	2010-2014	2	
Patrick Chan	Canada (CAN)	2014	2	

Which athletes are from South Korea after 2010? Kim Yu-Na

Logical forms that lead to answer:

Athlete from South Korea after 2010

Athlete from South Korea with 2 medals

First athlete in the table with 2 medals

Athlete in row 4

Challenge: Spurious logical forms

There is exactly one square touching the bottom of a box True



Due to binary denotations, 50% of logical forms give correct answer!

lead to answer:

touching bottom of boxes

Count of yellow squares is 1

There exists a yellow triangle

There exists an object

Cornell Natural Language Visual Reasoning, Suhr et al., 2017

Training Objectives

Maximum Marginal Likelihood

Reward/Cost -based approaches

search to get stuck in the exponential search

7. 2018).

Eg : Li Proposal: Alternate between the two objectives while gradually Krishr increasing the search space! Maxin set of N $\min_{\theta} \sum \mathbb{E}_{p(y_i|x_i;\theta)} \mathcal{C}(x_i, y_i, w_i, d_i)$ $p(y_i|x_i; heta)$ max θ $x_i, w_i, z_i \in D \ y_i \in Y | \llbracket y_i \rrbracket^{w_i} = z_i$... but random initialization can cause the

space

... but we need a good set of approximate logical forms



Step 0: Get seed set of logical forms till depth **k**





 $D = \{x_i, w_i, z_i\}_{i=1}^N$ Limited depth exhaustive search Max logical form $D^{0} = \{x_{j}, Y_{j}\}_{j=1}^{M} \\ \forall_{y_{j} \in Y_{j}} \mathcal{C}(x_{j}, y_{j}, w_{j}, d_{j}) = 0$ depth = k LSTM LSTM LSTM LSTM Maximum Marginal Likelihood

Step 0: Get seed set of logical forms till depth **k**

Step 1: Train model using MML on seed set





Step 0: Get seed set of logical forms till depth **k**

Step 1: Train model using MML on seed set

Step 2: Train using MBR on all data
till a greater depth k + s





Step 0: Get seed set of logical forms till depth **k**

Step 1: Train model using MML on seed set

Step 2: Train using MBR on all data
till a greater depth k + s

Step 3: Replace offline search with trained MBR and update seed set







Spuriousness Solution 2: Coverage guidance

There is exactly one square touching the bottom of a box. (count_equals (square (touch_bottom all_objects)) 1)

- **Insight:** There is a significant amount of trivial overlap
- Solution: Use overlap as a measure guide search



Spuri	Example		ce
_	Sentence: There is exactly one square touching the bottom of Triggered target symbols: {count_equals, square, 1, touch_bottom} Coverage costs of candidate logical forms:	top → bottom → above → below → square → circle	
1 square touch_b	Logical form	Coverage	$\begin{array}{c} le \rightarrow triangle \\ \rightarrow yellow \end{array}$
Coverage cos symbols that form	<pre>(count_equals (square (touch_bottom all_objects)) 1)</pre>	0	→ black blue big → small → medium
	(count_equals (square all_objects) 1)	1	
	(object_exists all_objects)	4	



Training with Coverage Guidance

• Augment the reward-based objective:

$$\min_{\theta} \sum_{I=1}^{N} \mathbb{E}_{p(y_i|x_i;\theta)} \mathcal{C}(x_i, y_i, w_i, d_i)$$

now \mathcal{C} is defined a linear combination of **coverage** and **denotation** costs

$$\mathcal{C}(x_i, y_i, w_i, d_i) = \lambda \mathcal{S}(y_i, x_i) + (1 - \lambda) \mathcal{T}(y_i, w_i, d_i)$$



Results of training with iterative search on NLVR*



* using structured representations

Iterations



Results of training with iterative search on WikiTableQuestions



Iterations



Results of using coverage guided training on NLVR*



when model initialized from an MML model trained on a seed set of offline searched paths



when trained from scratch

AZ

Comparison with previous approaches on NLVR*



- MaxEnt, BiAttPonter are not semantic parsers
- Abs. supervision + Rerank uses manually labeled abstractions of utterance - logical form pairs to get training data for a supervised system, and reranking
- Our work outperforms Goldman et al., 2018 with fewer resources



* using structured representations

Comparison with previous approaches on WikiTableQuestions



Non-neural models

Reinforcement Learning models

Non-RL Neural Models



Summary

- Spuriousness is a challenge in training semantic parsers with weak supervision
- Two solutions:
 - Iterative training: Online search with initialization \Rightarrow MML over offline search output
 - Coverage during online search
- SOTA single model performances:
 - WikiTableQuestions: **44.3%**
 - NLVR semantic parsing: 82.9%

Thank you! Questions?

