Semantic Parsing to Probabilistic Programs for Situated Question Answering

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Situated Question Answering

Which animal is both predator and prey?



(A) hawk(B) mouse(C) cricket(D) owl



Food Web Questions



How many organisms consume the mouse? (A) 1 (B) 2 (C) 3 (D) 4

Which animal is both predator and prey? (A) hawk (B) mouse (C) cricket (D) owl

If all the rabbits die off, what will happen to the mountain lion population? (A) it will increase (B) it will decrease (C) it will remain the same



Examples of Situated Question Answering

Visual QA [Antol et al., 2015]



Q: what is her mustache made of? A: bananas

Robot Commands

[Kollar et al., 2010]



Q: go down the hallway then right

A:



(path through environment)

Text QA [Berant et al., 2014]



Q: What can the splitting of water lead to? A: Transfer of ions



Challenges

- 1. Noisy computer vision
- 2. Compositionality in language "predator and prey"
- 3. Interactions between question and diagram interpretations
- Background knowledge Secondary consumer = an animal that eats animals that eat plants





Prior Work



Prior Work



Outline

- Introduction
- Parsing to Probabilistic Programs [EMNLP 2016]
 - Model
 - Experiments
- Probabilistic Neural Programs

[NAMPI 2016]



Parsing to Probabilistic Programs (P³)

Question How many animals consume the mouse?





Semantic Parsing

Question	How many animals consume the mouse?		
	Semantic Parsing		
Logical Form	<pre>(count (lambda (x) (eats x mouse)))</pre>		

[Zelle and Mooney, 1996] [Zettlemoyer and Collins 2005] [Liang et al., 2011] [Kwiatkowski et al., 2013] [Artzi et al., 2013]



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Semantic Parsing

Question	How many a consume the	nimals e mouse?	Database	eats weasel mouse	mouse shrubs	organism mouse weasel shrubs
	Semantic Parsing					
Logical Form	(count (la (mbda (x) eats x mouse	•)))			
	Execution	· · · · · ·			+	
		2				

[Zelle and Mooney, 1996] [Zettlemoyer and Collins 2005] [Liang et al., 2011] [Kwiatkowski et al., 2013] [Artzi et al., 2013]



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Food Web Domain Theory

• Roles:

organism, animal, plant, sun, decomposer, predator, prey, herbivore, carnivore, omnivore, primary-consumer, secondary-consumer, tertiary-consumer, top-predator

- Change Directions: increase, decrease, unchanged, affect
- Relations:

eats, cause, has-role

Miscellaneous:
 count, and, or, not, exists



CCG Semantic Parsing

S : dec	S: decrease (frog)		$(S S): \lambda e. \lambda f. cause (e, f(raccoon))$			
frog	λx.decrease(x)	raccoon	$\lambda x.\lambda e.\lambda f.cause(e, f(x))$			
NP :	S\NP :	NP :	(S\S)\NP :			
If all the frogs	died the population of raccoon		would _ ?			

 $S: \lambda f. cause (decrease (frog), f(raccoon))$

[Zelle and Mooney, 1996] [Zettlemoyer and Collins 2005] [Liang et al., 2011] [Kwiatkowski et al., 2013] [Artzi et al., 2013] [Krishnamurthy 2016]



Semantic Parsing

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Parsing to Probabilistic Programs (P³)





Parsing to Probabilistic Programs (P³)



for ARTIFICIAL INTELLIGENCE

;; A normal program

(+ 1 1)

→ 2



;; choose represents a nondeterministic choice (choose 1 2)





- ;; choices can be nested
- (+ (choose 1 2) (choose 3 4))





- ;; choices can be nested
- (+ (choose 1 2) (choose 3 4))





Probabilistic Programming for Visual Uncertainty

```
(define (eats x y)
  ; first-cut approximation
  (choose true false)
)
; Helper functions
(define (plant x)
  (eat x sun))
(define (count f)
  (length (filter entities f)))
(define entities
  (list weasel shrubs mouse ...))
...
```

(eats weasel mouse)

Food web domain theory

Possible logical form executions



Probabilistic Programming for Visual Uncertainty

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Food web domain theory

Possible logical form executions



Probabilistic Programming for Visual Uncertainty



Food web domain theory

Possible logical form executions



Parsing to Probabilistic Programs (P³)



P³ as Loglinear Model





P³ as Loglinear Model

$$\begin{array}{ccc} & & & \text{Semantic} \\ & & \text{Model} & & \text{Parser} \\ & & & & \\ P(e,\ell,t|v,q;\theta) = & \frac{1}{Z_{q,v}} f_{\text{ex}}(e,\ell|v;\theta_{\text{ex}}) f_{\text{prs}}(\ell,t|q;\theta_{\text{prs}}) \end{array}$$



Loglinear CCG parser

$$\begin{aligned} \text{parameters} & \text{features} \\ f(\ell,t|s;\theta) = \exp\{\theta^T \phi(\ell,t,s))\} \\ & & & \\ & &$$



Loglinear Execution Model





Visual Preprocessing

























Parsing to Probabilistic Programs (P³)



"if all the frogs died what will happen to the raccoons?"	(lambda x (eats x raccoon))
Parser Beam Search	Execution Beam Search [Goodman and Stuhlmuller, 2014]
<pre>(lambda x (eats x raccoon)) (lambda x (cause (increase x) (decrease raccoon)))) (lambda x (plant x))</pre>	-> {frog} -> {} -> {frog, snake} -> {rabbit} ->



Training



Optimize Loglikelihood:

 $O(\theta) = \sum_{i=1}^{n} \log \sum_{e,l,t} c^{i}(e) P(e,\ell,t|q^{i},v^{i};\theta)$



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FoodWebs Data Set



Large roach Large roach Steelhead Steelhead Roach fry Predatory insects (lestids) Tuft-weaving chironomids Cladophora, epiphytic diatoms, Nostoe



How will it most likely affect the ecosystem if the human population increases?

(A) sharks will increase

(B) fish take over

(C) everything will decrease (and be destroyed..)

(D) seagulls will eat sharks

Which of the organisms has more prey?

- (A) stickleback fry
- (B) epiphytic diatoms

(C) roach fry

(D) steelhead

What is a producer? (A) fox (B) hen (C) plant (D) python

5000 questions, 500 diagrams, 3/1/1 train/validation/test split





Accuracy















FoodWebs Accuracy with Unseen Organisms





Scene Dataset

A blue colored coffee mug is placed very near to the computer on the table.





15 images, 284 natural language descriptions



Scene Experimental Results



ALLEN INSTITUTE

[Krishnamurthy and Kollar, 2013]

Accuracy

Scene Experimental Results



[Krishnamurthy and Kollar, 2013]

Accuracy

Future Directions

Visual QA [Antol et al., 2015]



Q: what is her mustache made of? A: bananas

Chart Questions



Q: By how much did model X improve accuracy over the best baseline? A: 5.5 points



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Rich Parameterization of Probabilistic Programs

(count (lambda (x)
 (eats x mouse)))





Rich Parameterization of Probabilistic Programs

(count (lambda (x)
 (eats x mouse)))







Computation Graphs for Neural Networks

- W = parameter(pW)
- b = parameter(pb)
- V = parameter(pV)
- a = parameter(pa)

```
x = vecInput(2)
y = scalarInput(0)
h = tanh((W*x) + b)
y_pred = logistic((V*h) + a)
```





Computation Graphs for Structured Prediction





Computation Graphs for Structured Prediction



Not "free" with computation graphs:

- Globally Normalized Models
- Cost-Sensitive Training
 - \circ $\,$ e.g., dynamic oracles for dependency parsing
- Reinforcement Learning









State Space of Discrete Choices

Computation Graphs for Neural Networks



val flip: Pp[Boolean] = choose(Seq(true, false), Seq(0.75, 0.25))





```
def mlp(featureVector: Tensor): Pp[CompGraphNode] = for {
```

```
params <- param("params")</pre>
```

```
bias <- param("bias")</pre>
```

```
hidden = ((params * featureVector) + bias).tanh
```

```
params2 <- param("params2")</pre>
```

```
bias2 <- param("bias2")</pre>
```

```
dist = (params2 * hidden) + bias2
```

```
\} \ \textbf{yield} \ \{
```

```
dist
```













Inference

val answerPnp: Pp[String] = answerQ("A thermometer ...")
val answerDist = answerPnp.beamSearch(100, nnParams)

<u>answerDist</u>	
temperature	0.7
fever	0.2





Training

```
val trainer = LoglikelihoodTrainer(...)
```

val trainedParams = trainer.train(data, initialParams, ...)

- Maximum likelihood w/ approximate inference
- Learning to search
- Reinforcement learning





What happens to the snake population if the field mouse population decreases? (A) it will increase (B) it will decrease (C) it will remain the same





λf.cause(decrease(mouse), f(snake))





Experiments





Experiments





Conclusion





Which of the organisms has more prey? (A) stickleback fry (B) epiphytic diatoms

- (D) epipinylic dialoi
- (C) roach fry
- (D) steelhead



Code and data http://allenai.org/paper-appendix/emnlp2016-p3/ http://github.com/allenai/pnp/

