Backpropagating through Structured Argmax using a SPIGOT

Hao Peng, Sam Thomson, Noah A. Smith







@ACL July 17, 2018









Aim

• Structured prediction as a layer.

Motivation

- Structures help. Ji and Smith, 2017; Oepen et al., 2017
- Linguistic structures may not be universally optimal.
 Williams, 2017



Aim

• Structured prediction as a layer.

Motivation

- Structures help. Ji and Smith, 2017; Oepen et al., 2017
- Linguistic structures may not be universally optimal.
 Williams, 2017

Challenges

argmax is non-differentiable.



Aim

• Structured prediction as a layer.

Motivation

- Structures help. Ji and Smith, 2017; Oepen et al., 2017
- Linguistic structures may not be universally optimal.
 Williams, 2017

Challenges

• argmax is non-differentiable.



Outline

- * Background: structured prediction as linear programs
- Method: SPIGOT algorithm
- Experiments

Structured Prediction Reviewed

Input

Shareholders took their money

Output



Structured Prediction Reviewed

Input

Shareholders took their money

Score S_{θ} (Shareholders took their money) || $\sum_{\text{arcs}} s_{\theta}$ (head mod)

Structured Prediction Reviewed

Input

Shareholders took their money



Linear Programming Formulation



Roth and Yih, 2004; Martins et al., 2009

Linear Programming Formulation



Roth and Yih, 2004; Martins et al., 2009

Outline

- Background: structured prediction as linear programs
- * Method: SPIGOT algorithm
- Experiments









We have: $\nabla_{\hat{\mathbf{z}}} \mathcal{L}$ We need: $\nabla_{\mathbf{s}} \mathcal{L}$

We have: $\nabla_{\hat{\mathbf{z}}} \mathcal{L}$ We need: $\nabla_{\mathbf{s}} \mathcal{L}$

Leibniz, 1676



We have: $\nabla_{\hat{\mathbf{z}}} \mathcal{L}$ We need: $\nabla_{\mathbf{s}} \mathcal{L}$

Leibniz, 1676

 $\nabla_{\mathbf{s}} \mathcal{L} = \mathbf{J} \nabla_{\hat{\mathbf{z}}} \mathcal{L} \quad \mathbf{J}$



We have: $\nabla_{\hat{\mathbf{z}}} \mathcal{L}$ We need: $\nabla_{\mathbf{s}} \mathcal{L}$

Leibniz, 1676

 $\nabla_{\mathbf{s}} \mathcal{L} = \mathbf{J} \nabla_{\hat{\mathbf{z}}} \mathcal{L}$

Straight-through Estimator (STE)

Hinton, 2012; Bengio et al., 2013

 $\nabla_{\mathbf{s}} \mathcal{L} \triangleq \nabla_{\hat{\mathbf{z}}} \mathcal{L}$

Straight-through Estimator (STE): $\nabla_{s} \mathcal{L} \triangleq \nabla_{\hat{z}} \mathcal{L}$



Straight-through Estimator (STE): $\nabla_{s} \mathcal{L} \triangleq \nabla_{\hat{z}} \mathcal{L}$



Straight-through Estimator (STE): $\nabla_{\mathbf{s}} \mathcal{L} \triangleq \nabla_{\hat{\mathbf{z}}} \mathcal{L}$





















Connections to Related Work





STE

	Pipeline	STE	Structured Att.	SPIGOT
Hard decision on $\hat{\mathbf{Z}}$				
Backprop		\checkmark		
Marginal				
Projection				

Structured Attention: Kim et al., 2017

Connections to Related Work





$$\hat{\mathbf{z}} = \operatorname{softmax}(\dots)$$

	Pipeline	STE	Structured Att.	SPIGOT
Hard decision on $\hat{\mathbf{Z}}$		\checkmark		
Backprop				
Marginal				
Projection				






Outline

- Background: structured prediction as linear programs
- Method: SPIGOT algorithm
- * Experiments

Experiments: Syntactic-then-semantic Parsing



Experiments: Syntactic-then-semantic Parsing



Experiments: Syntactic-then-semantic Parsing











Neurbo: Peng et al., 2017

Semantic Parsing for Sentiment Classification



Semantic Parsing for Sentiment Classification



Stanford Sentiment Treebank accuracy



Conclusion



Conclusion



Conclusion



Thank you!