# Greedy, Joint Syntactic-Semantic Parsing with Stack I STMs

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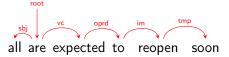
12th Aug

YM-style Syntactic dependency parsing + PropBank-style semantic role labeling

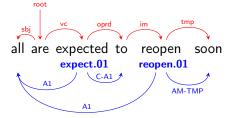
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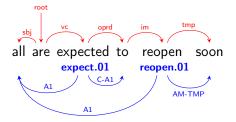
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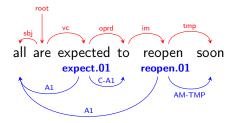


YM-style Syntactic dependency parsing + PropBank-style semantic role labeling



 Correspondence between syntactic and semantic dependencies [Levin and Hovav, 1996]

YM-style Syntactic dependency parsing + PropBank-style semantic role labeling



- Correspondence between syntactic and semantic dependencies [Levin and Hovav, 1996]
- Language understanding: QA, relation extraction, text categorization

# A little more about PropBank SRL

[Palmer et al., 2005]

Most common solution: pipeline syntax and semantics

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#### **Pipelines**

 Have access to complete syntactic information

#### Incremental, joint approach

No such access

# A little more about PropBank SRL

[Palmer et al., 2005]

- Most common solution: pipeline syntax and semantics
- ▶ Pipelines involve *expensive* feature extraction step [Johansson, 2009, He et al., 2013]
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#### **Pipelines**

- Have access to complete syntactic information
- ► Slow feature extraction step

#### Incremental, joint approach

- No such access
- Fast

#### Outline

Introduction

Incremental Algorithm

Stack LSTM Model

CoNLL 2008-09 Shared Task Results

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- $lackbox{ Parse structure} 
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- ► Transition : **shift** and **reduce** actions

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- ▶ Initialize the *stack* as empty and the *buffer* to contain the sentence

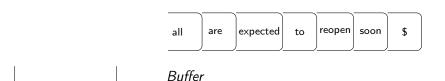
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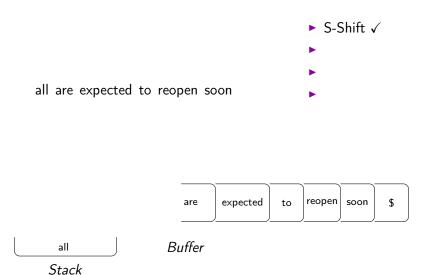
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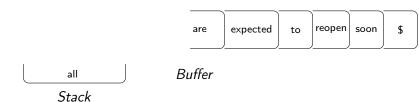
Modified arc-eager algorithm [Nivre, 2008, Henderson et al., 2008, Henderson et al., 2013, Gesmundo et al., 2009, Titov et al., 2009]



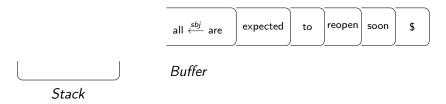






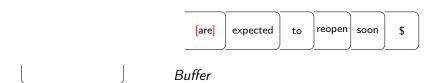






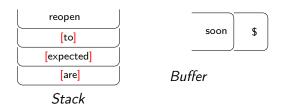
Stack





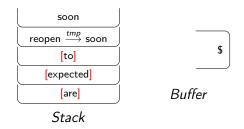


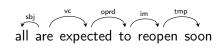
- ► S-Shift
- ► S-Left
- ► S-Right



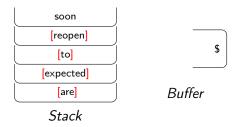


- ► S-Shift
- ► S-Left
- ► S-Right ✓



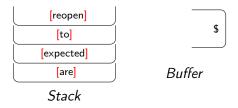


- ► S-Shift
- ► S-Left
- ► S-Right
- S-Reduce

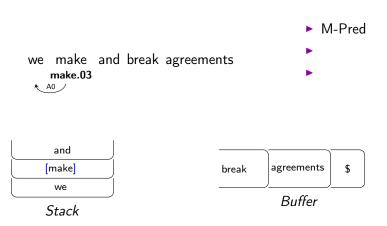


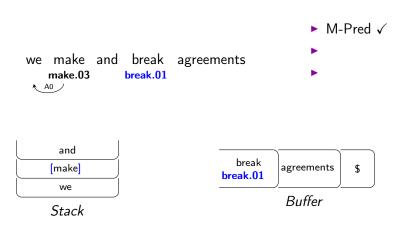


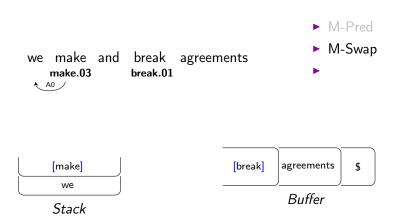
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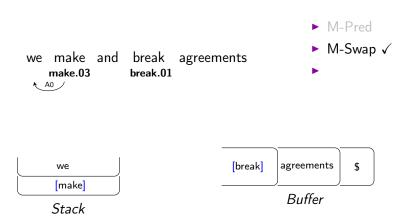


- M-Shift
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- M-Right
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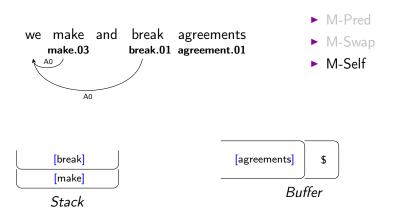




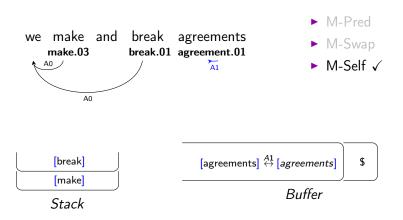




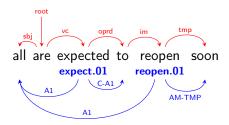
#### More transitions for semantics



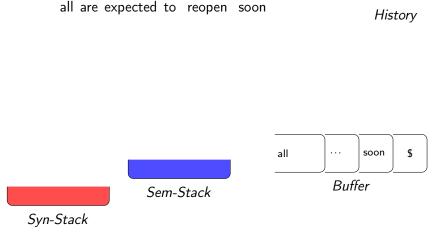
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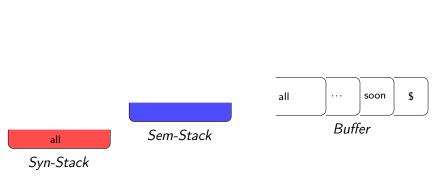
# Synchronizing syntax and semantics



- ► Two stacks: Syn-Stack and Sem-Stack
- ▶ Share the Buffer
- Syntactic transitions followed by semantic transitions for a given Buffer state [Henderson et al., 2008]

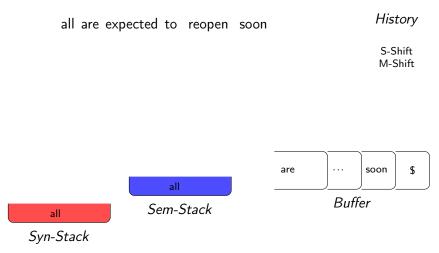


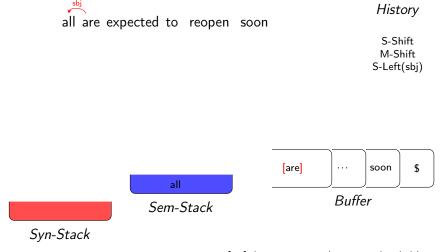
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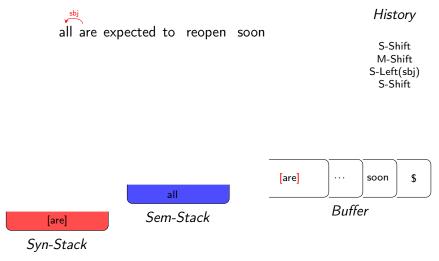


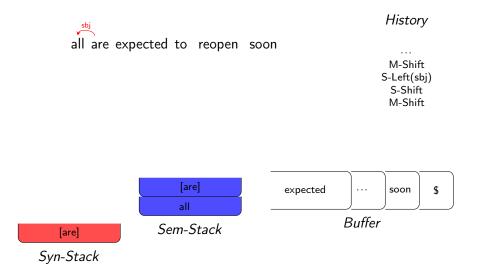
History

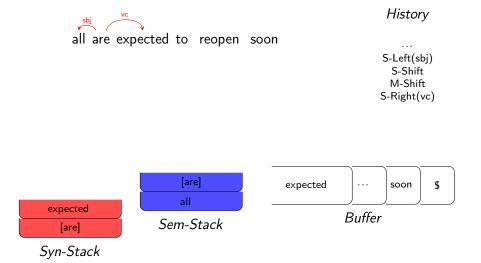
S-Shift



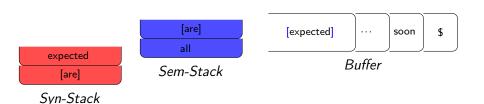


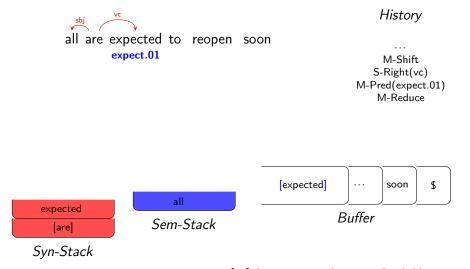


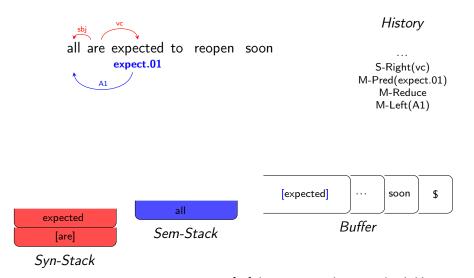




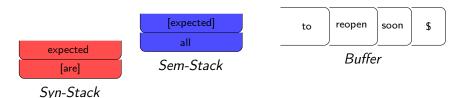








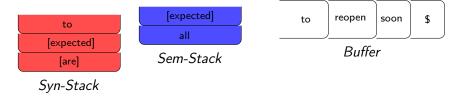




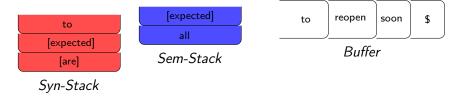




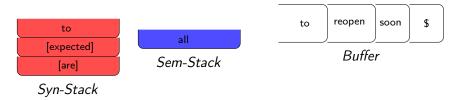




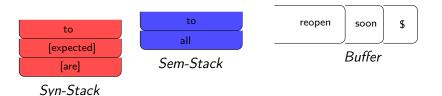




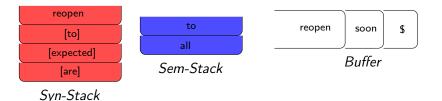




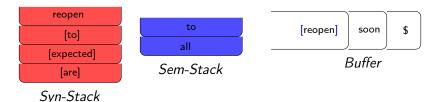




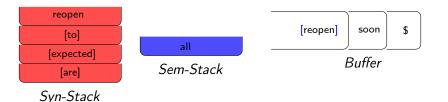


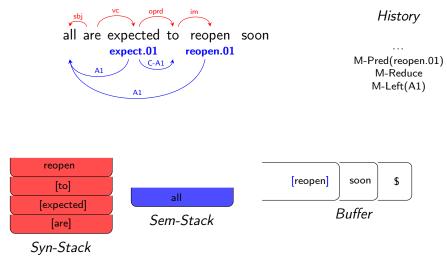




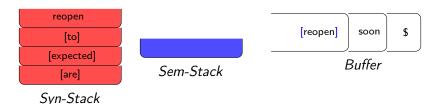


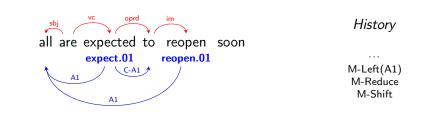


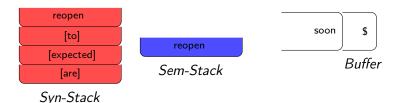


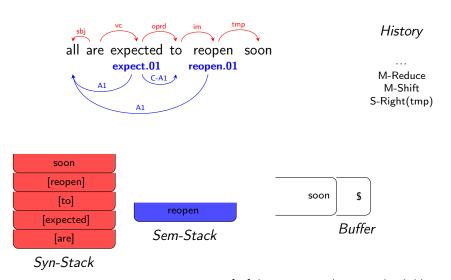


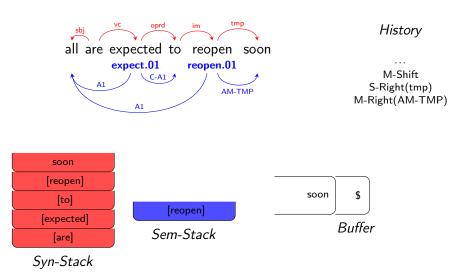


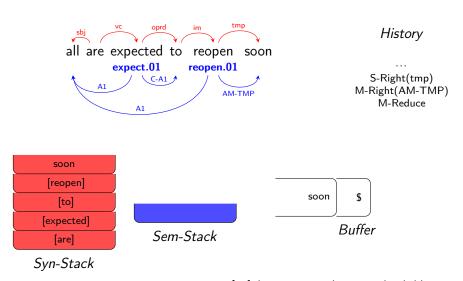


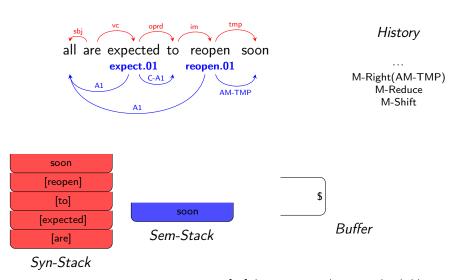


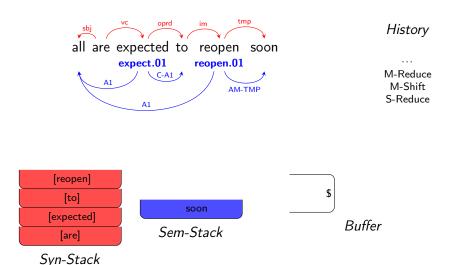


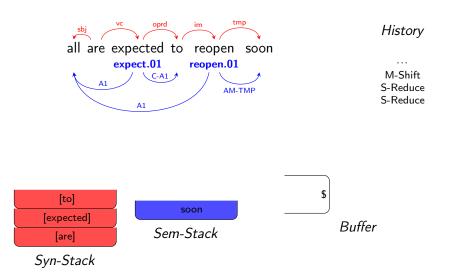


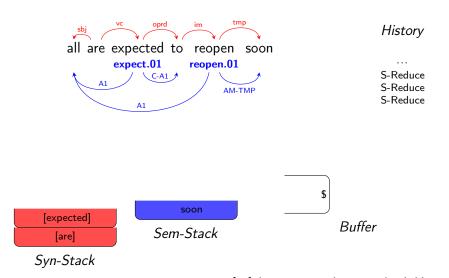


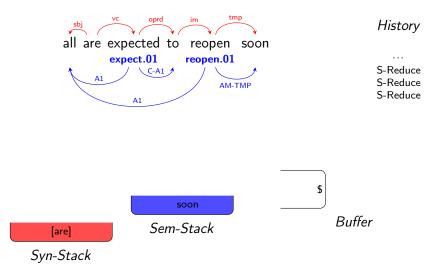


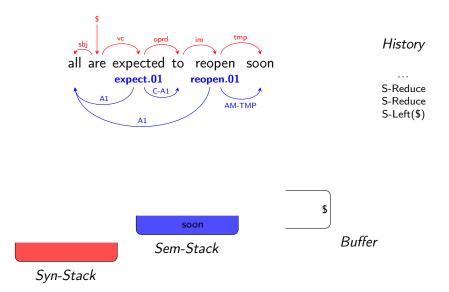


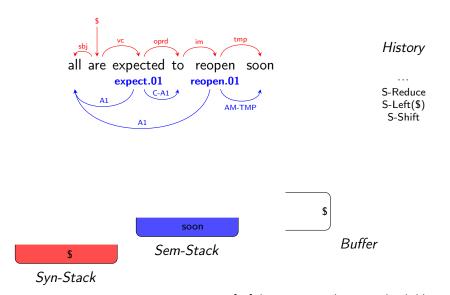


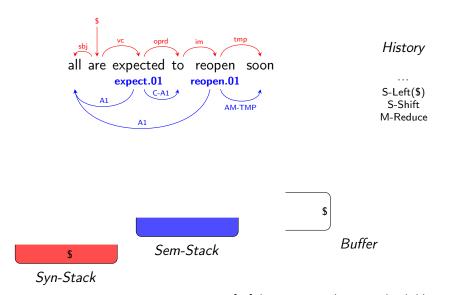




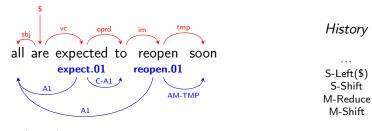




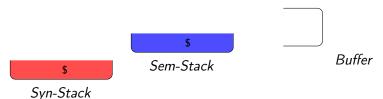




### An example transition sequence



Linear algorithm



#### Outline

Introduction

Incremental Algorithm

Stack LSTM Model

CoNLL 2008-09 Shared Task Results

#### Stack LSTM Model

▶ LSTMs: Recurrent neural networks with special memory cell [Hochreiter and Schmidhuber, 1997, Graves, 2013] to learn fixed-size representations for variable-length sequences

#### Stack LSTM Model

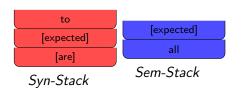
- ► LSTMs: Recurrent neural networks with special memory cell [Hochreiter and Schmidhuber, 1997, Graves, 2013] to learn fixed-size representations for variable-length sequences
- Stack LSTMs: LSTMs equipped with stack operations [Dyer et al., 2015]
  - lacktriangle summary ightarrow return a fixed-size continuous representation
  - lacktriangle push ightarrow add to the sequence
  - $lackbox{ pop} 
    ightarrow {\sf remove}$  from the sequence

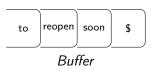
### Stack LSTM for Joint Parsing



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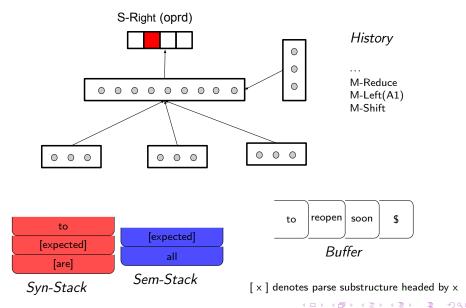
M-Reduce M-Left(A1) M-Shift



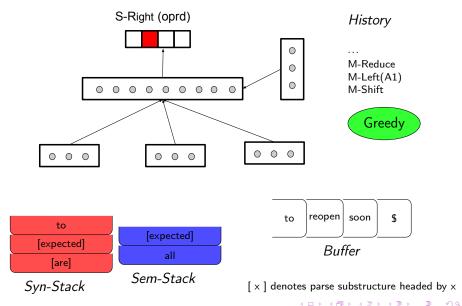


[  $\times$  ] denotes parse substructure headed by  $\times$ 

# Stack LSTM for Joint Parsing



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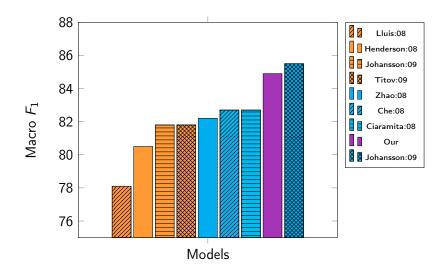
#### CoNLL Shared Tasks

- ▶ 2008: English only
- ▶ 2009: Multilingual
- Evaluation metrics:
  - Syntax: Labeled Accuracy Score (LAS)
  - SRL: Semantic F<sub>1</sub>
  - Rank systems: Macro F<sub>1</sub>

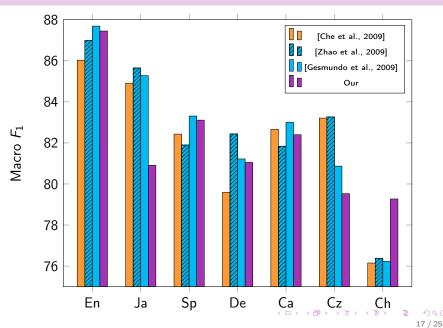
### Experimental Setup

- ▶ POS tags were used, but no other provided features
- No manually-designed features
- Dataset-specific hyperparameter tuning

# CoNLL 2008 (English only) Shared Task



# CoNLL 2009 (Multilingual) Shared Task



#### Conclusion

Take-aways!

- ► Incremental algorithm (linear) + model using stack LSTMs
- Avoid the effort involved in manual feature engineering
- Light-weight alternative to expensive pipelined systems

Code available at
https://github.com/clab/joint-lstm-parser

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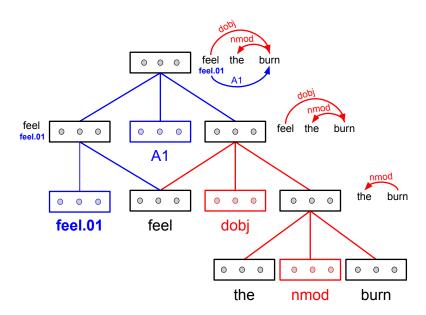
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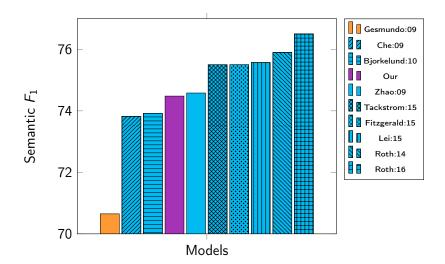
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### Extras

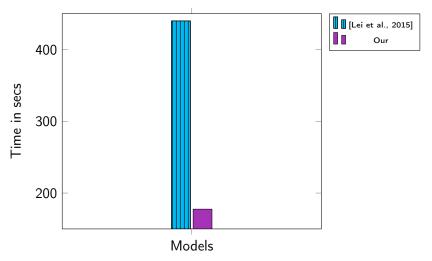
### Syntactic-semantic composition



# SRL performance on out-of-domain (Brown) data CoNLL 2009 Shared Task



# Time to decode the CoNLL 2009 English dataset



Experiments were run end to end on a single CPU