

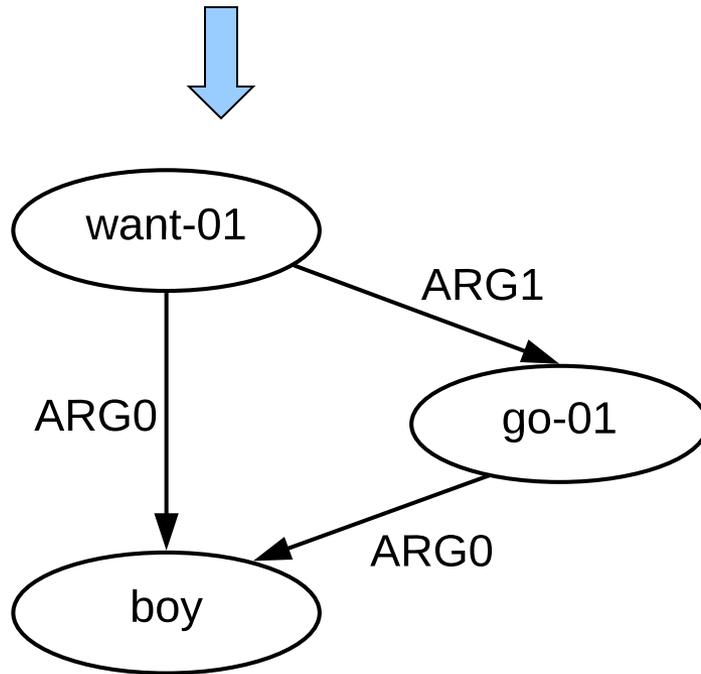
# Sequence-to-sequence Models for Cache Transition Systems

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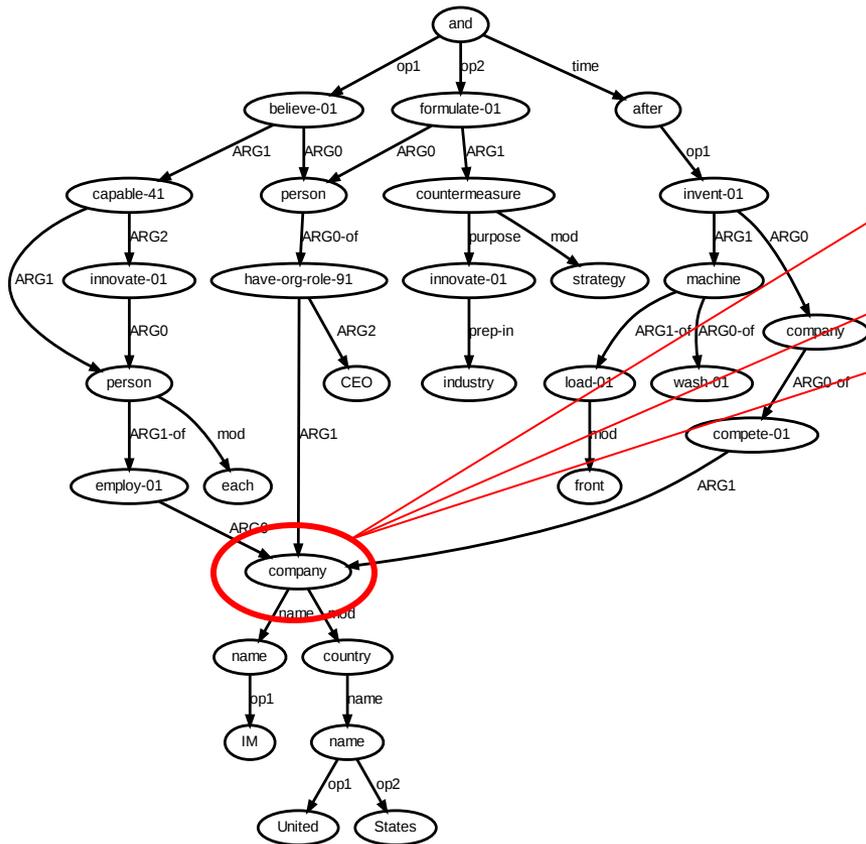


# AMR

- “John wants to go”



# AMR



After **its** competitor invented the front loading washing machine, the CEO of the American **IM company** believed that each of **its** employees had the ability for innovation, and formulated strategic countermeasures for innovation in the industry.

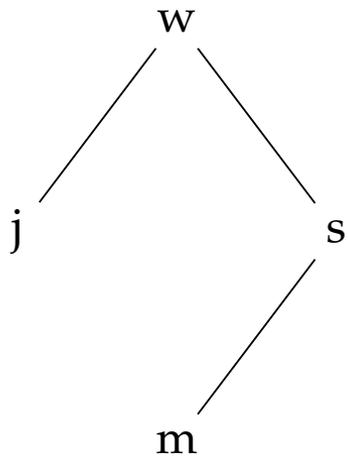


# Transition-based AMR parsing

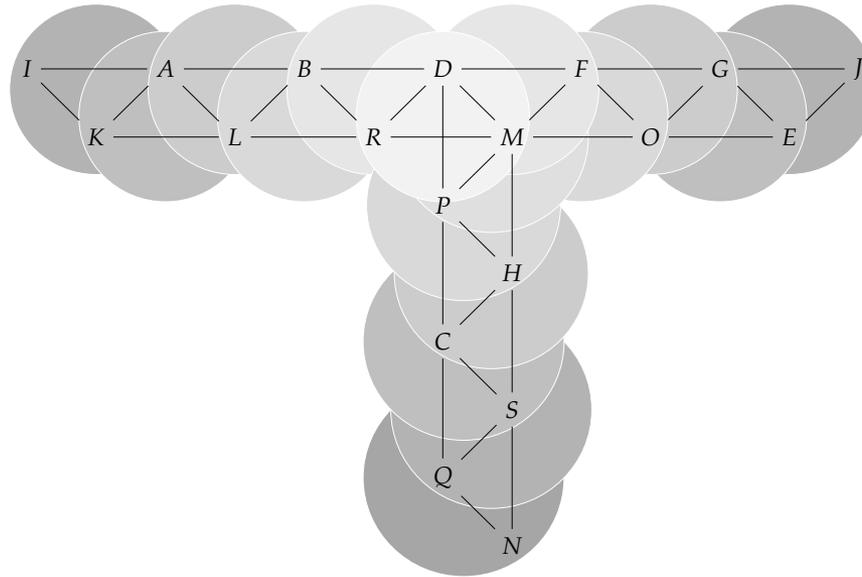
- There has been previous work (Sagae and Tsujii; Damonte et al.; Zhou et al.; Ribeyre et al.; Wang et al.) on transition-based graph parsing.
- Our work introduces a new data structure “cache” for generating graphs of certain *treewidth*.



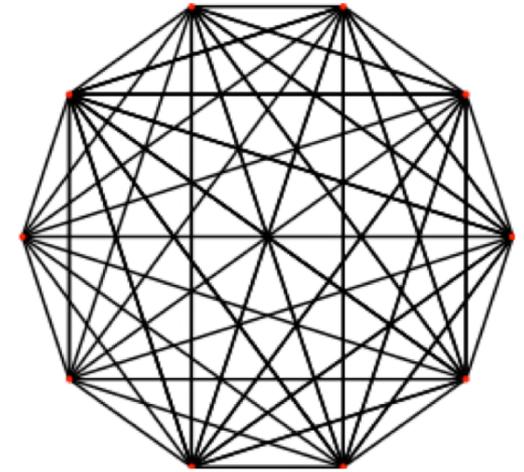
# Introduction to treewidth



A tree: treewidth 1



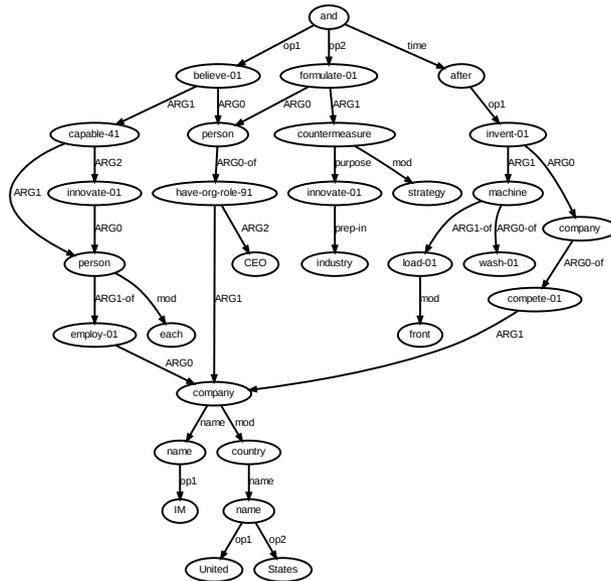
treewidth 2



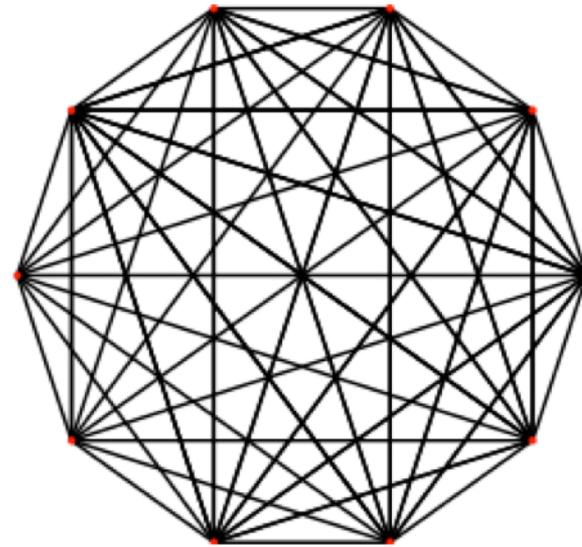
Complete graph of  $N$  nodes: treewidth  $N-1$



# Introduction to treewidth



small tree width  
~ 2.8 on average



large tree width





# Cache transition system

- Configuration  $c = (\sigma, \eta, \beta, E)$ 
  - Stack  $\sigma$ : place for temporarily storing concepts
  - **Cache  $\eta$ : working zone for making edges, fixed size corresponding to the treewidth.**
  - Buffer  $\beta$ : unprocessed concepts
  - $E$ : set of already-built edges



# Cache transition system

- Actions
  - SHIFT PUSH(i): shift one concept from buffer to right-most position of cache, then select one concept (index i) from cache to stack.

stack

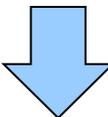


cache

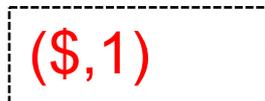


buffer



SHIFT PUSH(1) 

stack



cache



buffer



# Cache transition system

- Actions
  - POP: pop the top from stack and put back to cache, then drop the right-most item from cache.

stack

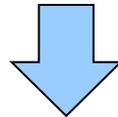
(\$,1)

cache

\$ \$ PER

buffer

want-01 go-01



stack

cache

\$ \$ \$

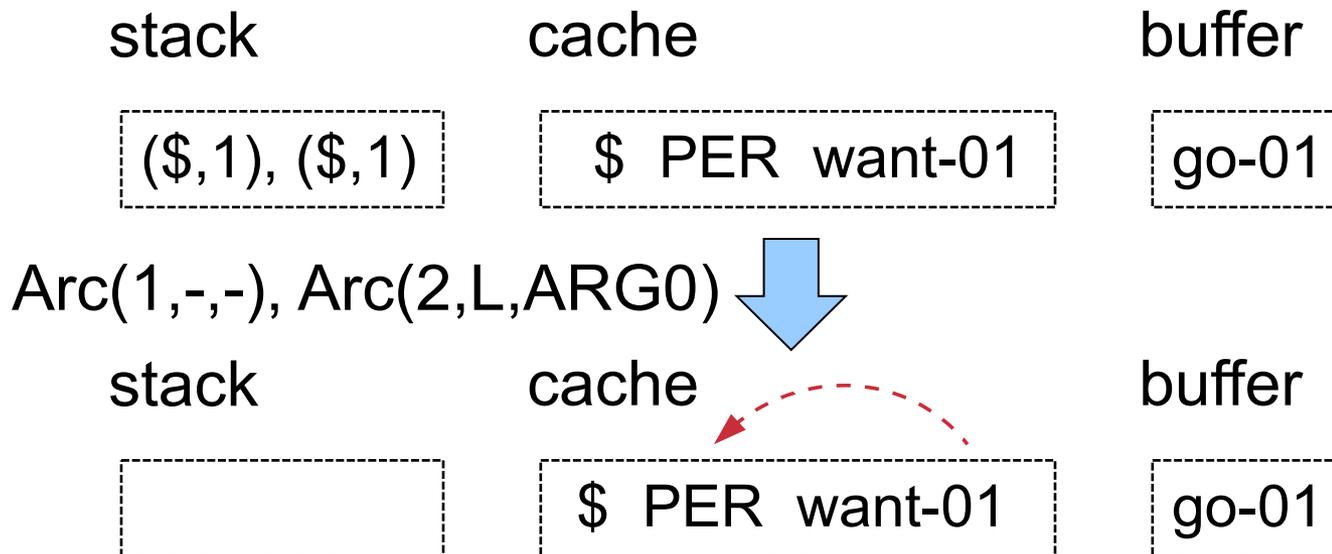
buffer

want-01 go-01



# Cache transition system

- Actions
  - $\text{Arc}(i, l, d)$ : make an arc (with direction  $d$ , label  $l$ ) between the right-most node to node  $i$ .  $\text{Arc}(i,-,-)$  represents no edge between them.



# Example of cache transition

Action taken: Initialization

stack



cache



buffer



# Example of cache transition

Action taken: SHIFT, PUSH(1)

stack

(1, \$)

cache

\$ \$ PER

buffer

want-01 go-01

Hypothesis: PER



# Example of cache transition

Action taken: Arc(1, -, -), Arc(2, -, -)

stack

(1, \$)

cache

\$ \$ PER

buffer

want-01 go-01

Hypothesis: PER



# Example of cache transition

Action taken: SHIFT, PUSH(1)

stack

cache

buffer

(1, \$) (1, \$)

\$ PER want-01

go-01

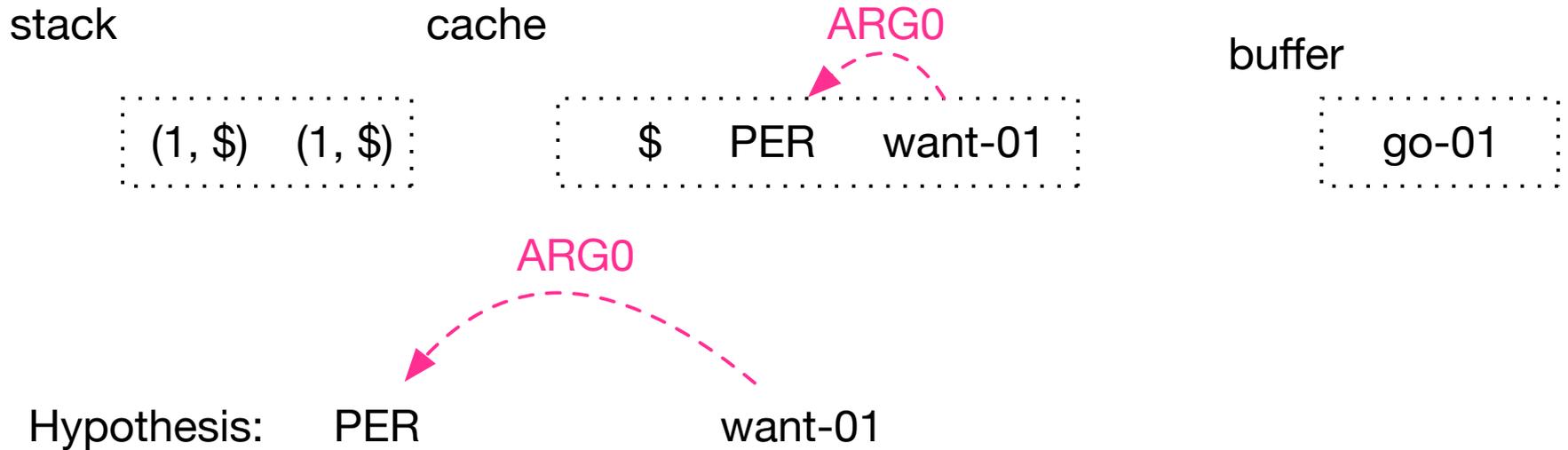
Hypothesis: PER

want-01



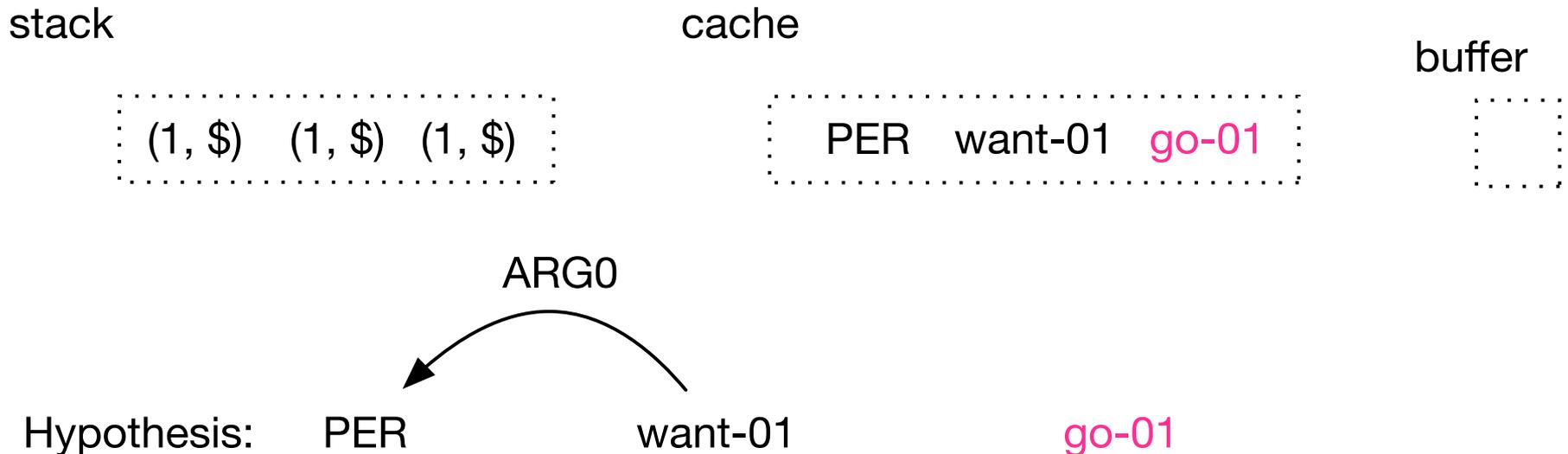
# Example of cache transition

Action taken: Arc(1, -, -), Arc(2, L, ARG0)



# Example of cache transition

Action taken: SHIFT, PUSH(1)



# Example of cache transition

Action taken: Arc(1, L, ARG0), Arc(2, R, ARG1)

stack

(1, \$) (1, \$) (1, \$)

cache

PER want-01 go-01

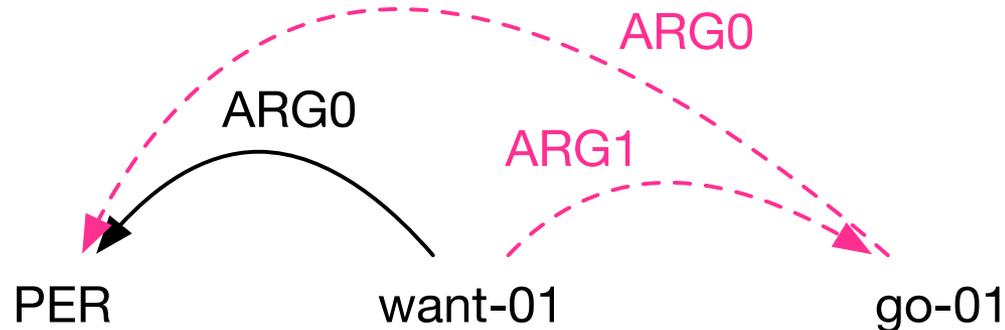


The cache contains the words PER, want-01, and go-01. A dashed pink arc labeled ARG0 connects PER to go-01. A dashed pink arc labeled ARG1 connects want-01 to go-01.

buffer

Hypothesis:

PER want-01 go-01

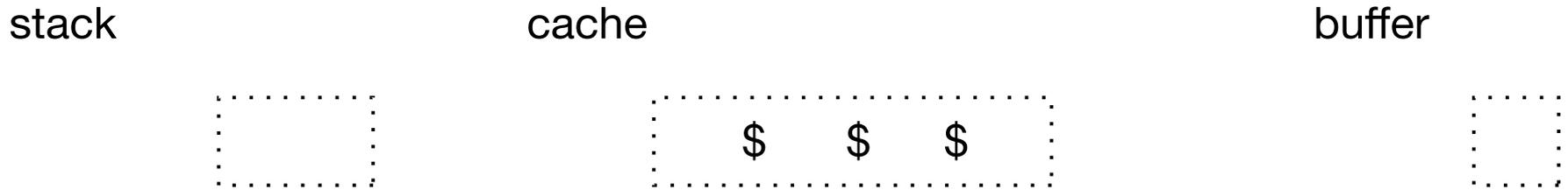


The hypothesis shows the words PER, want-01, and go-01. A solid black arc labeled ARG0 connects PER to want-01. A dashed pink arc labeled ARG1 connects want-01 to go-01. A dashed pink arc labeled ARG0 connects PER to go-01.

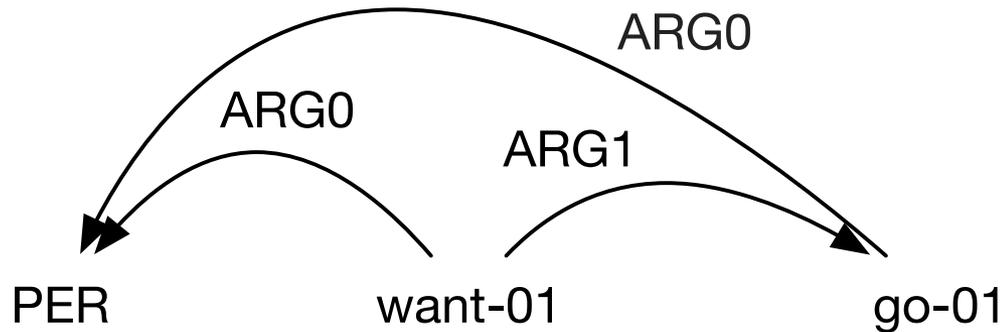


# Example of cache transition

Action taken: POP POP POP



Hypothesis:

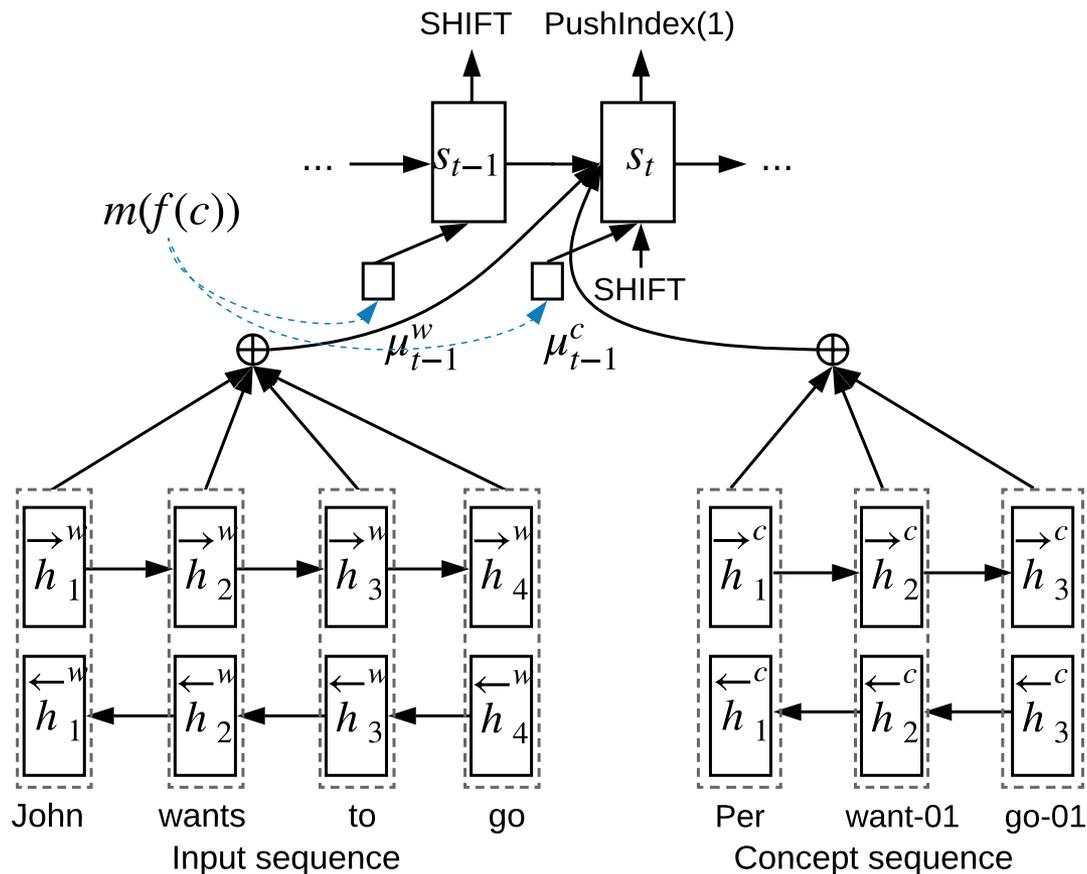


# Sequence to sequence models for cache transition system

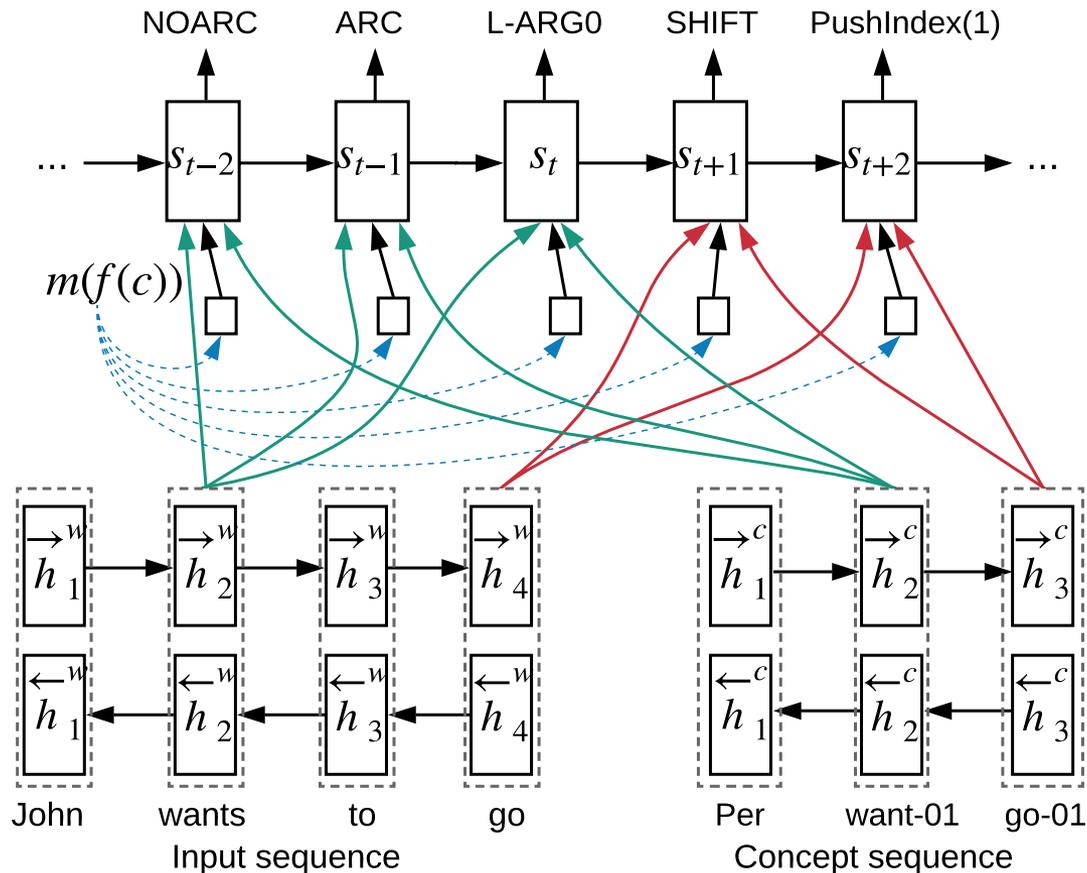
- Concepts are generated from input sentences by another classifier in the preprocessing step.
- Separate encoders are adopted for input sentences and sequences of concepts, respectively.
- One decoder for generating transition actions.



# Seq2seq (soft-attention+features)



# Seq2seq (hard-attention+features)

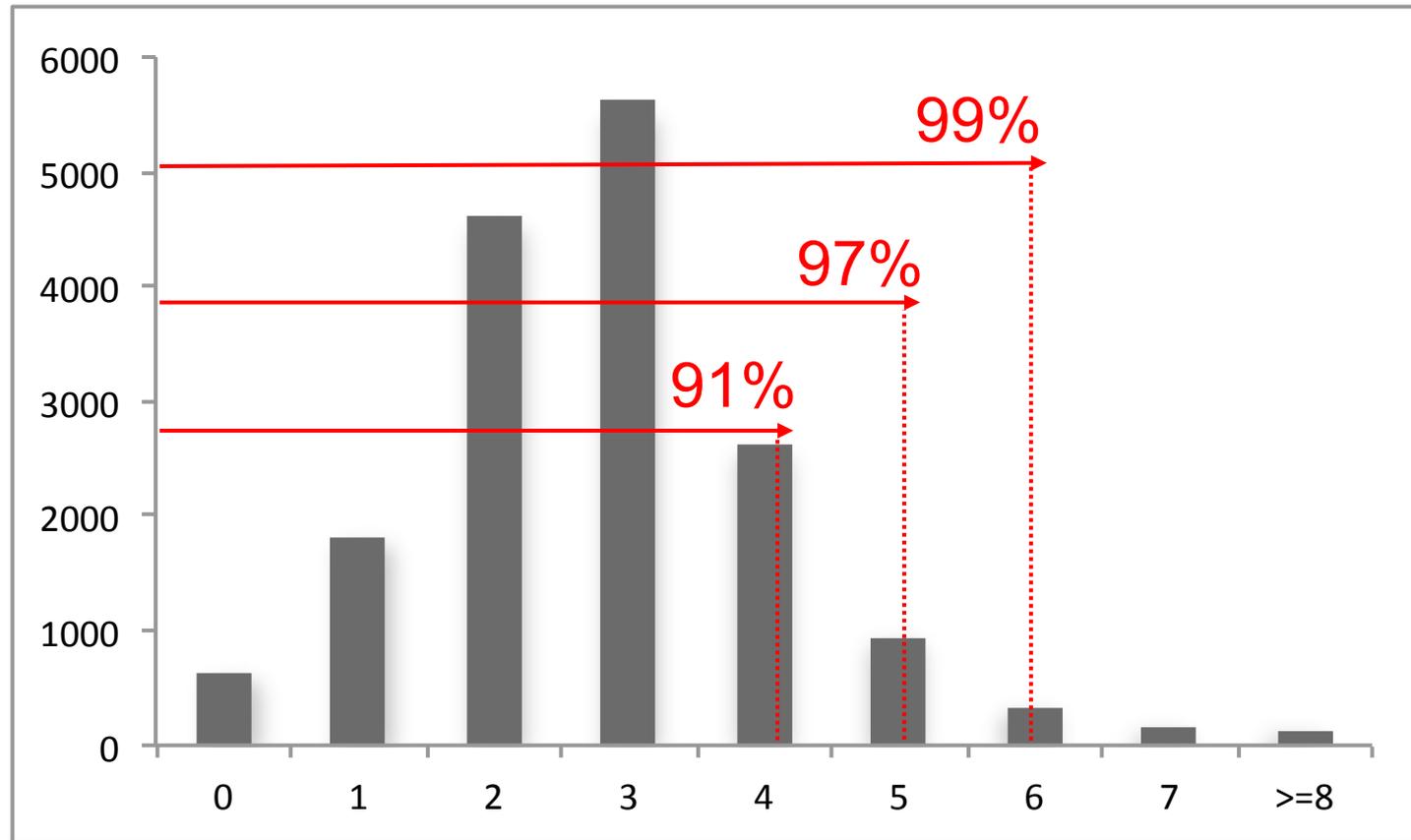


# Experiments

- Dataset: LDC2015E86
  - 16,833(train)/1,368(dev)/1,371(test)
- Evaluation: Smatch (Cai et al., 2013)



# AMR Coverage with different cache sizes



# Development results

| Model      | P    | R    | F    | cache size | P    | R    | F    |
|------------|------|------|------|------------|------|------|------|
| Soft       | 0.55 | 0.51 | 0.53 | 4          | 0.69 | 0.63 | 0.66 |
| Soft+feats | 0.69 | 0.63 | 0.66 | 5          | 0.70 | 0.64 | 0.67 |
| Hard+feats | 0.70 | 0.64 | 0.67 | 6          | 0.69 | 0.64 | 0.66 |

Impact of various components

Impact of cache size



# Main results

| Model                             | P    | R    | F    |
|-----------------------------------|------|------|------|
| Buyss and Blunsom (2017)          | --   | --   | 0.60 |
| Konstas et al. (2017)             | 0.60 | 0.65 | 0.62 |
| Ballesteros and Al-Onaizan (2017) | --   | --   | 0.64 |
| Damonte et al. (2016)             | --   | --   | 0.64 |
| Wang et al. (2015a)               | 0.70 | 0.63 | 0.66 |
| Flanigan et al. (2016)            | 0.70 | 0.65 | 0.67 |
| Wang and Xue (2017)               | 0.72 | 0.65 | 0.68 |
| Lyu and Titov (2018)              | --   | --   | 0.74 |
| Soft+feats                        | 0.68 | 0.63 | 0.65 |
| Hard+feats                        | 0.69 | 0.64 | 0.66 |



# Accuracy on reentrancies

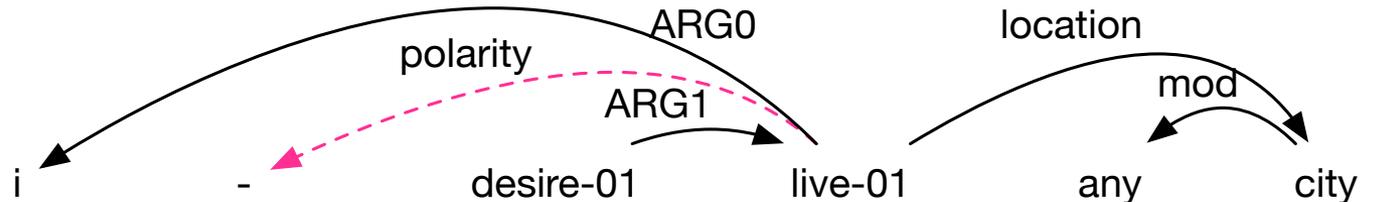
| Model                  | P           | R           | F           |
|------------------------|-------------|-------------|-------------|
| Peng et al., (2018)    | 0.44        | 0.28        | 0.34        |
| Damonte et al., (2017) | --          | --          | 0.41        |
| JAMR                   | 0.47        | <b>0.38</b> | 0.42        |
| Hard+feats (ours)      | <b>0.58</b> | 0.34        | <b>0.43</b> |



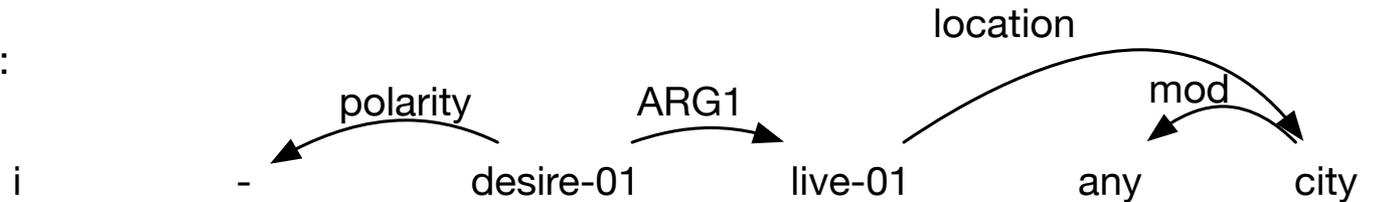
# Reentrancy example

Sentence: I have no desire to live in any city .

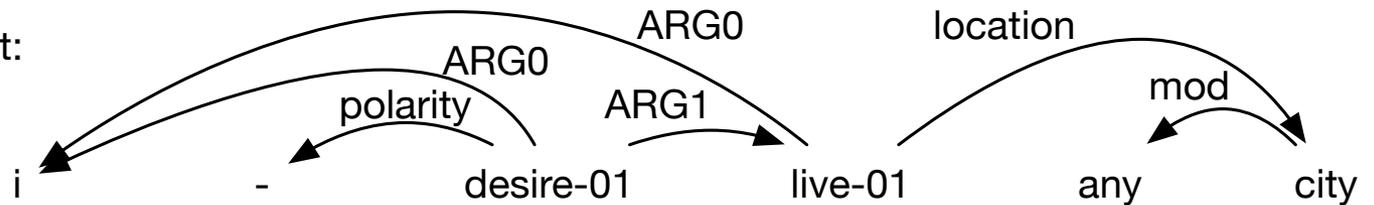
JAMR output:



Peng et al. (2018) output:



Our hard attention output:



# Conclusion

- Cache transition system based on a mathematical sound formalism for parsing to graphs.
- The cache transition process can be well-modeled by sequence-to-sequence models.
  - Features from transition states.
  - Monotonic hard attention.



Thank you for listening!  
Questions

