

Judging Grammaticality with Tree Substitution Grammar Derivations

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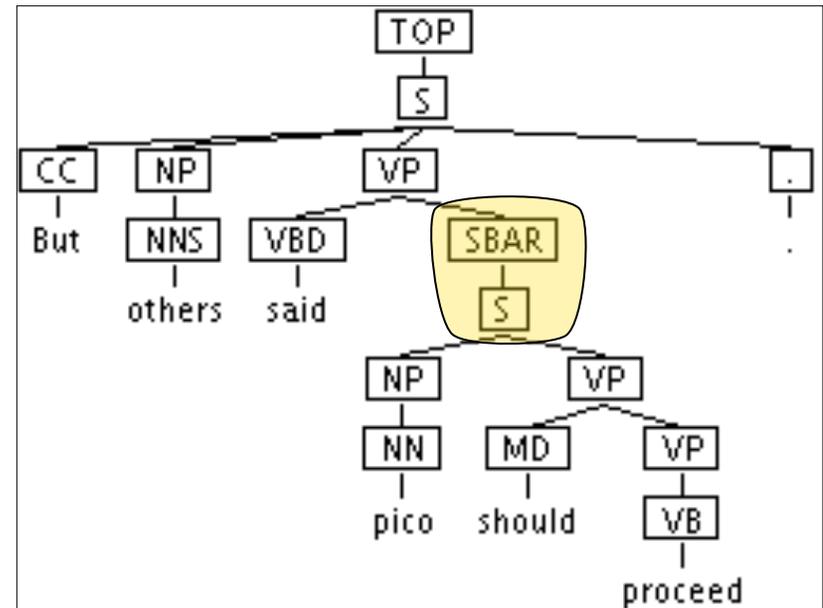
Overview

- Features from TSG derivations are helpful in distinguishing real text from “pseudo-negative” text



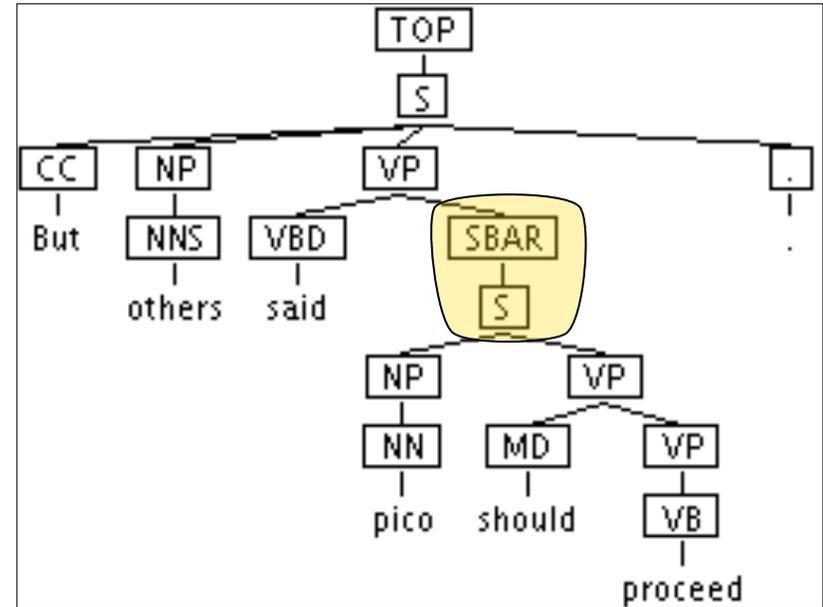
Standard CFG

Nonterminals rewrite as
sequence of child nonterminals



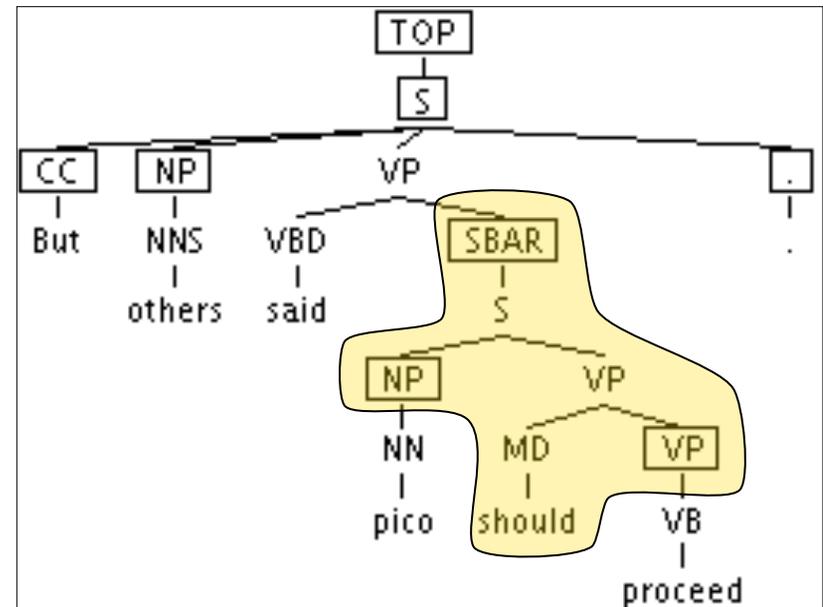
Standard CFG

Nonterminals rewrite as sequence of child nonterminals



TSG

Nonterminals rewrite as sequence of child fragments of arbitrary size



Motivation

- Language modeling is hard, and syntax isn't as easy a solution as one might hope:
 - parser scores aren't very helpful (Och et al., 2004)
 - but can be improved with task-specific training (Cherry & Quirk, 2008)



Motivation

- TSGs are an interesting formalism, but haven't been that helpful in parsing (though see Bansal & Klein, 2010)
- Tree features are helpful in related tasks, like parse reranking (Charniak & Johnson, 2005)
- Perhaps they can improve language modeling

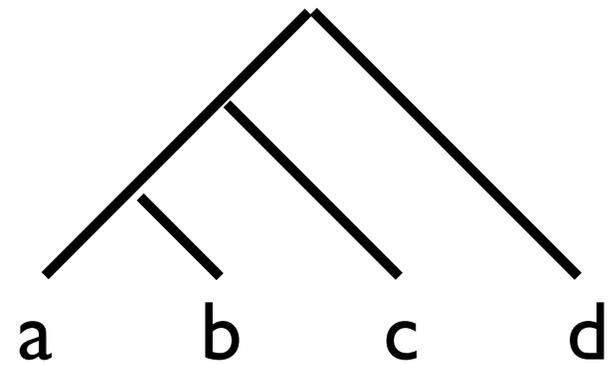


Grammatical?

a b c d

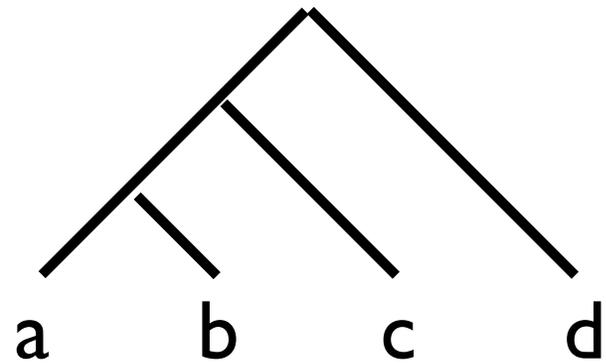


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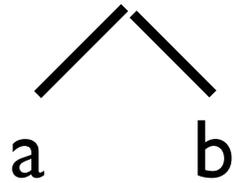


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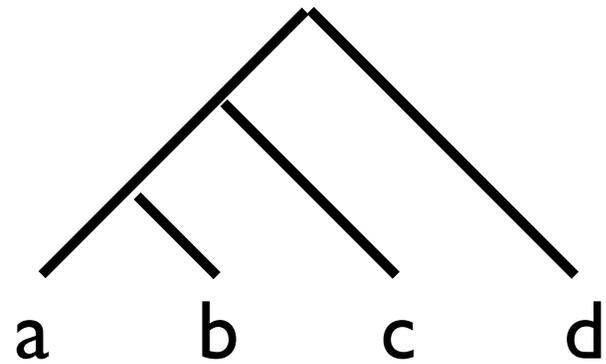
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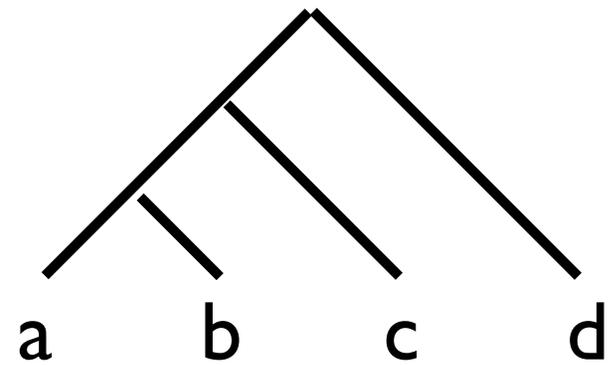
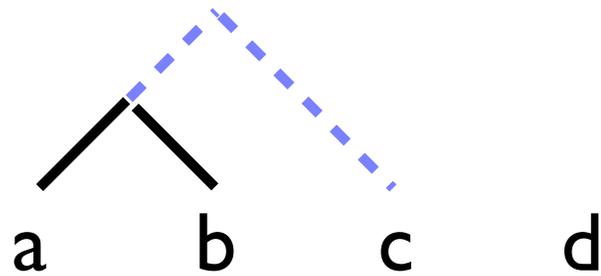
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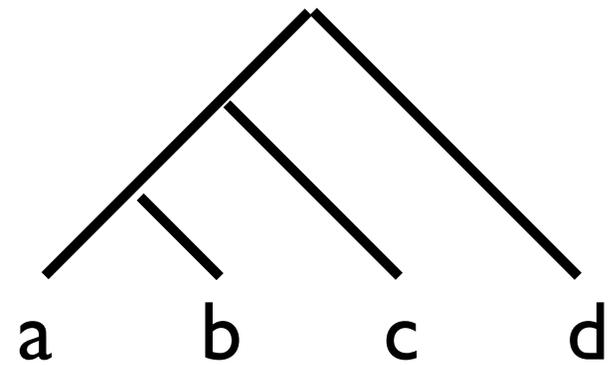
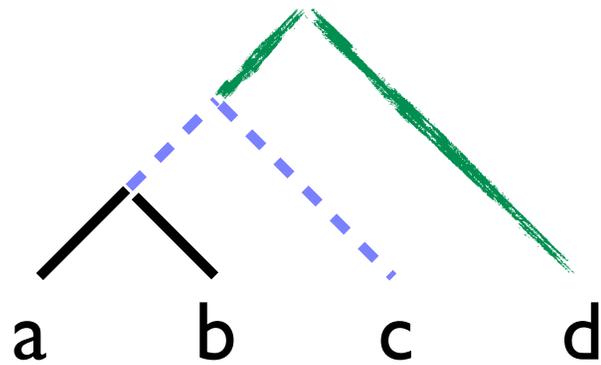
c d



Grammatical?



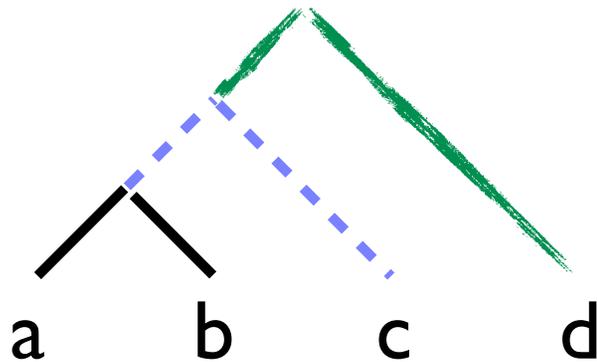
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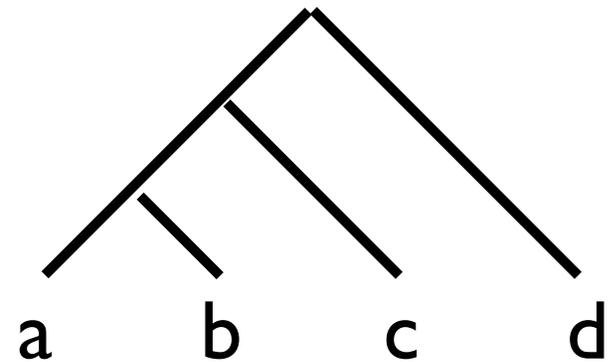
Grammatical?

increased likelihood of grammaticality →

many little fragments

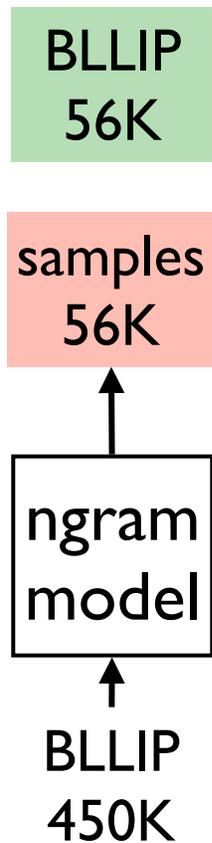


single large fragment



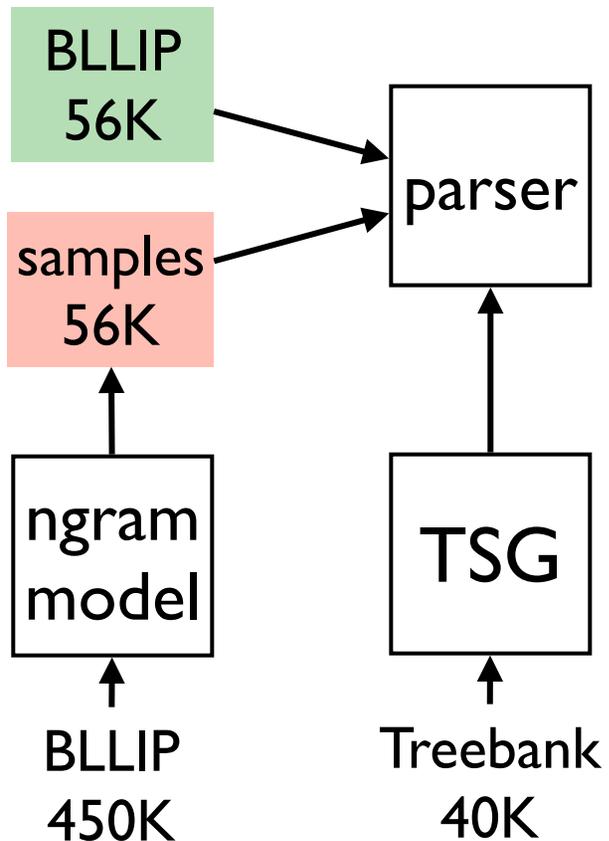
Experiments

Cherry & Quirk (2008)



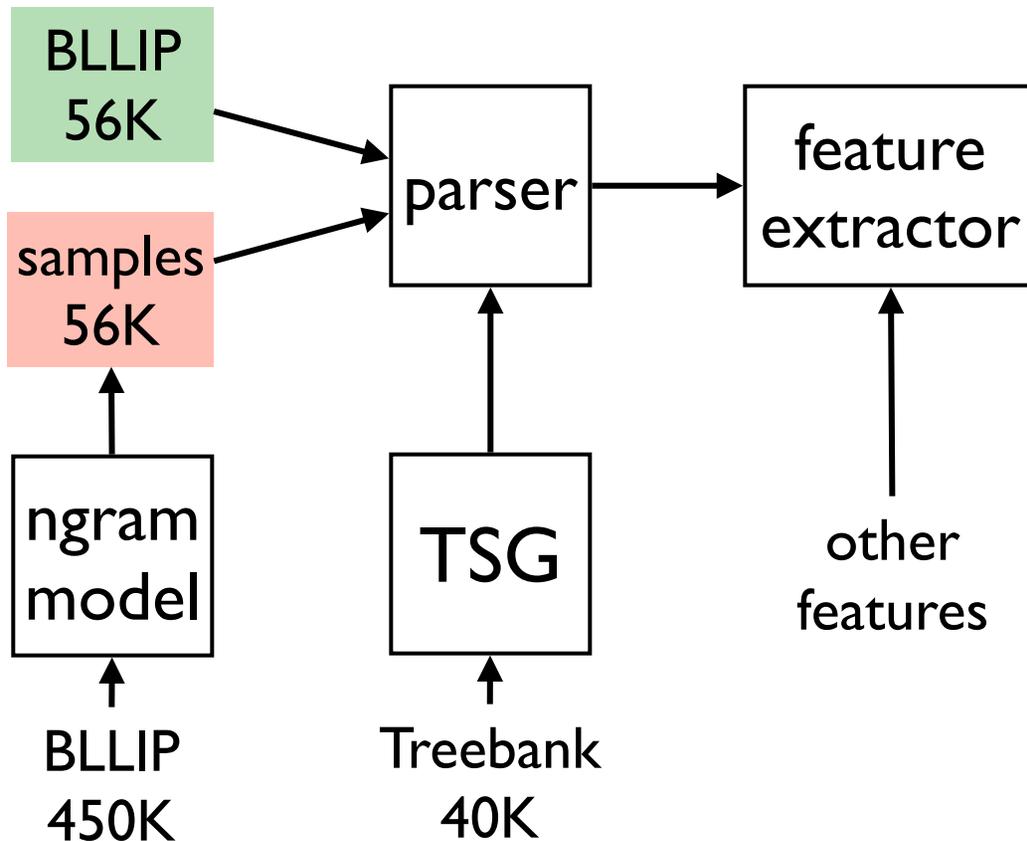
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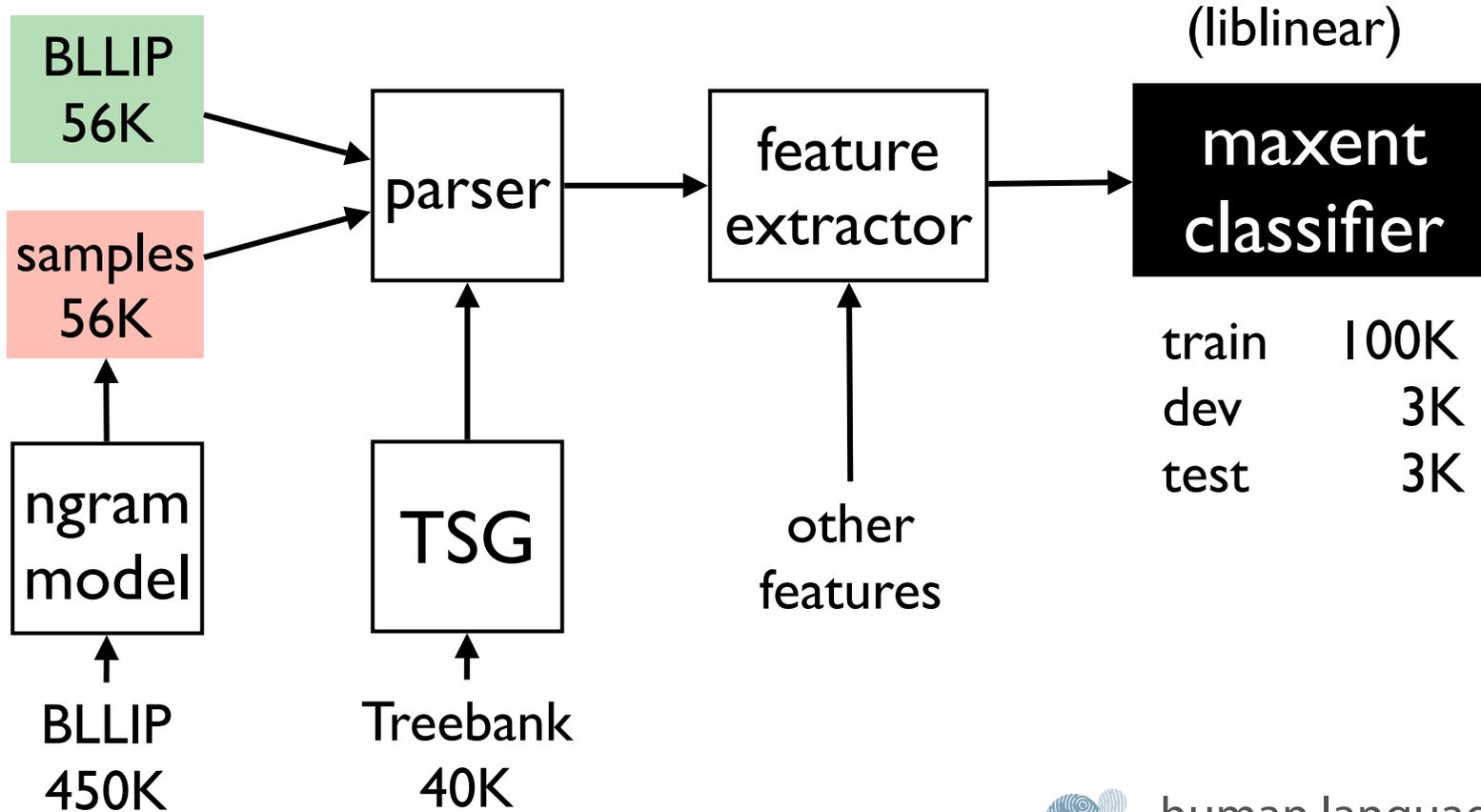
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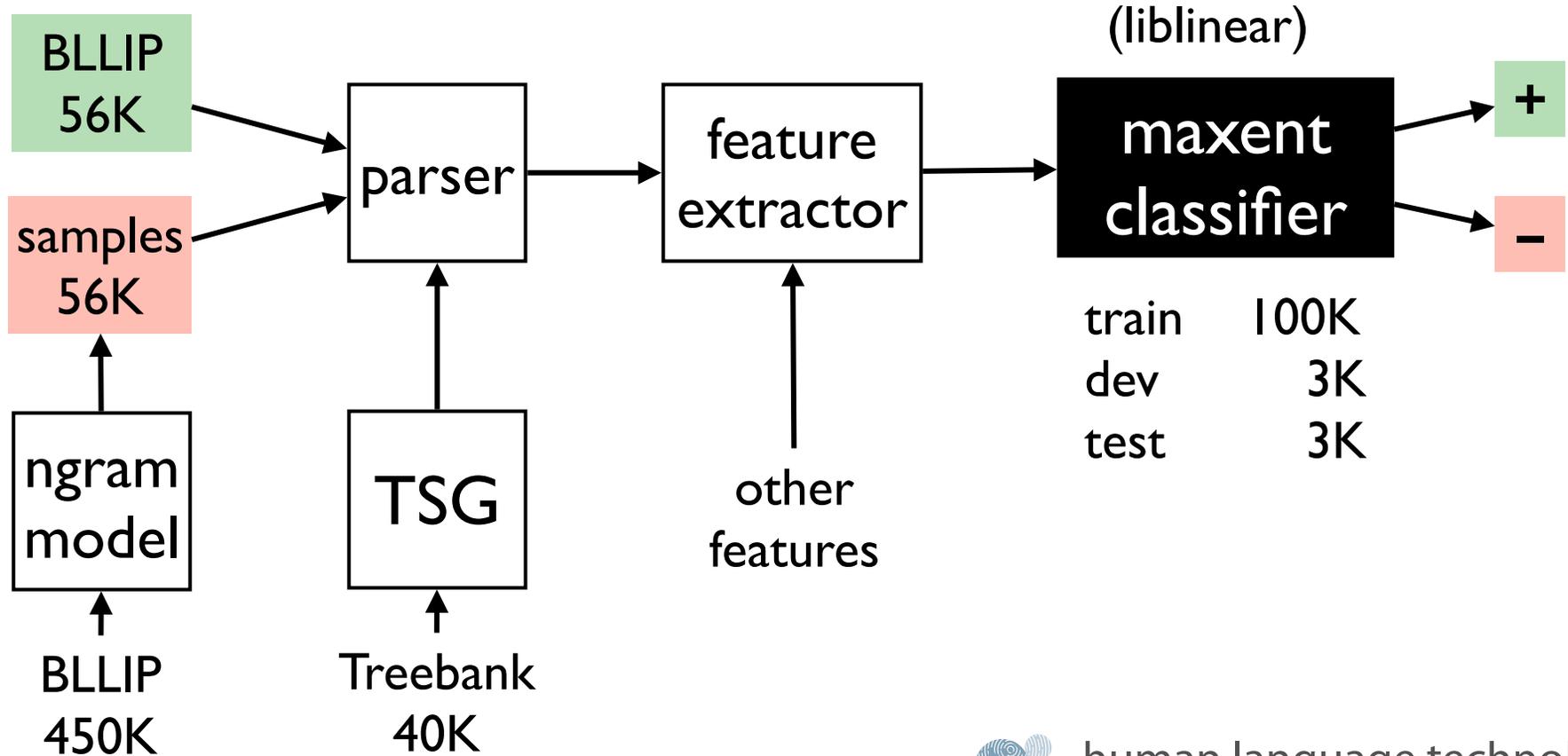
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Feature sets

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length	<i>17</i>

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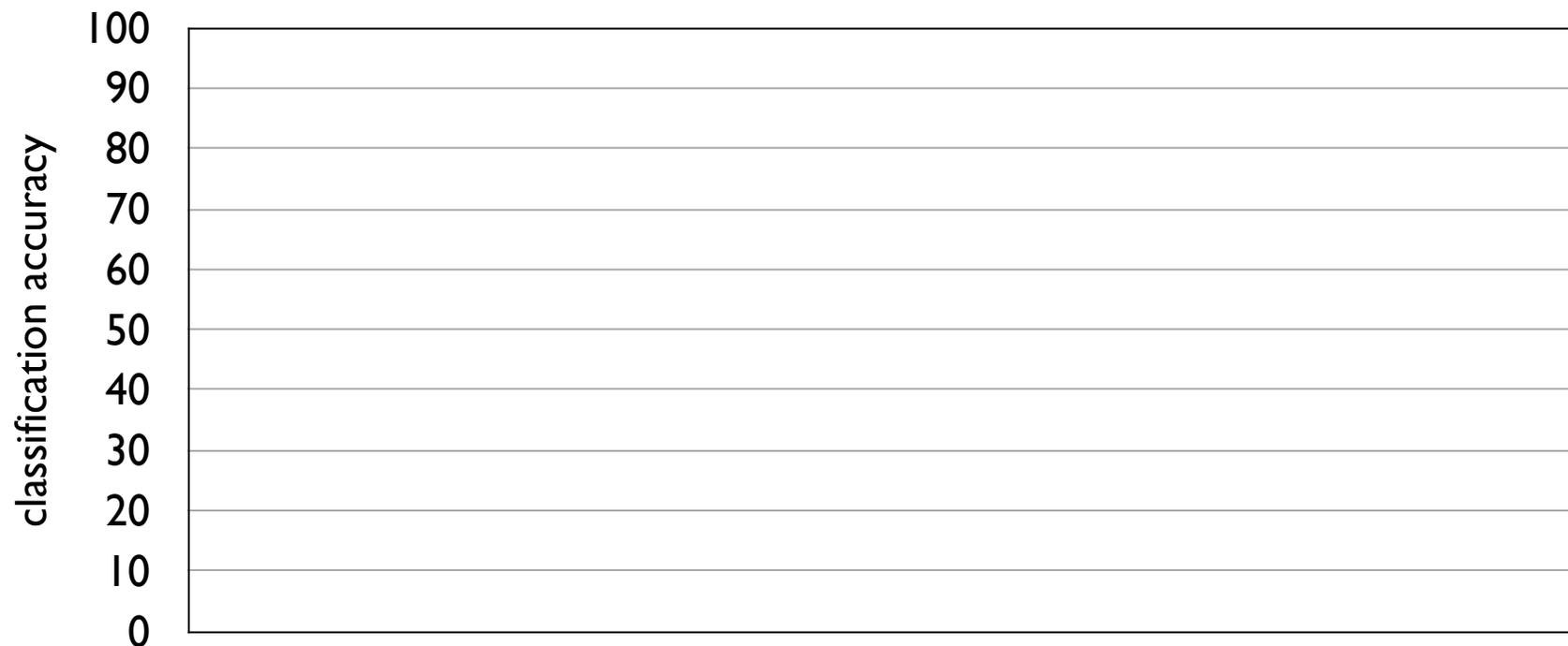
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TSG fragments	(TOP (S NP (VP VBD said) NP SBAR) .)

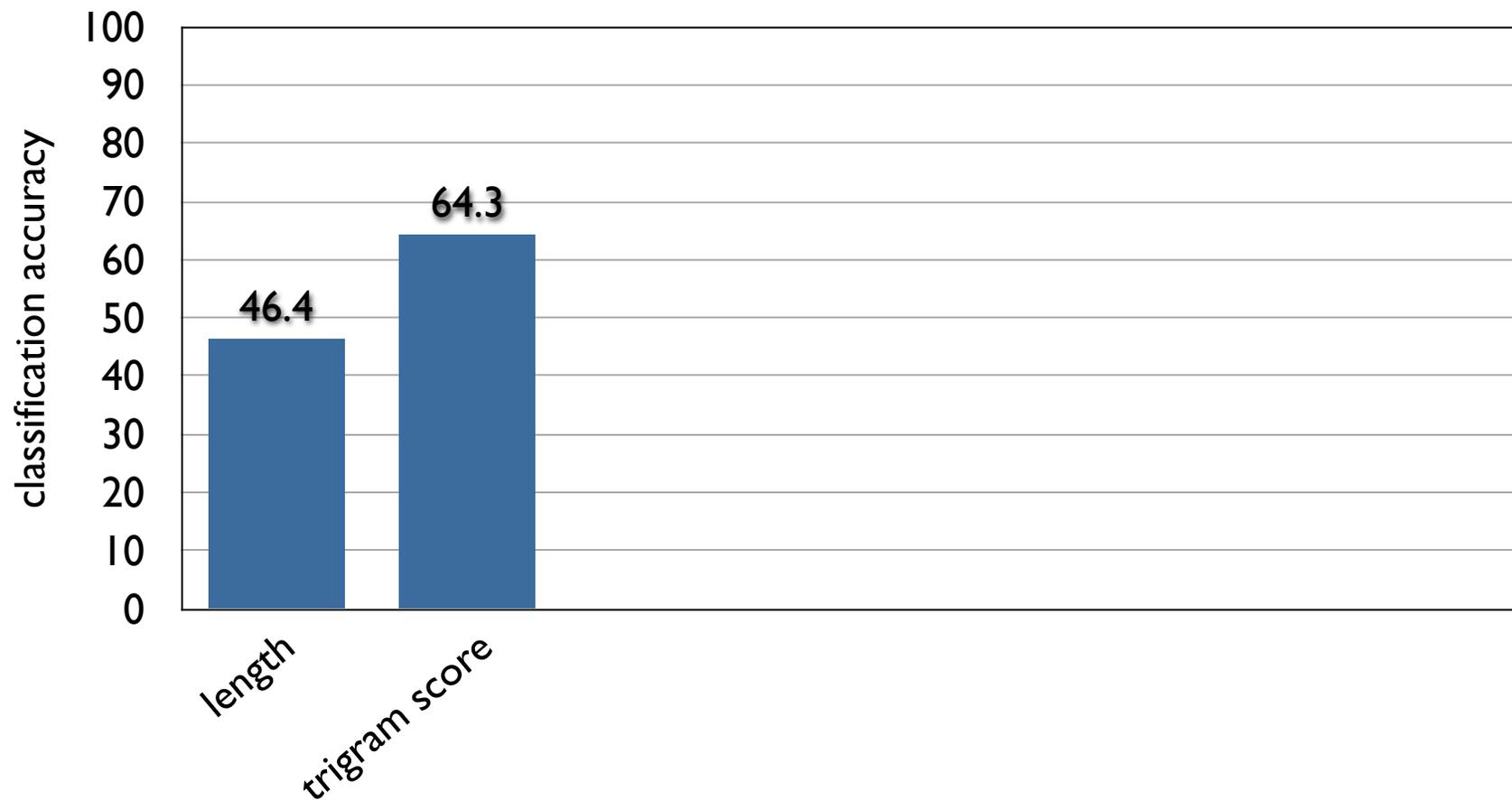
Classification results



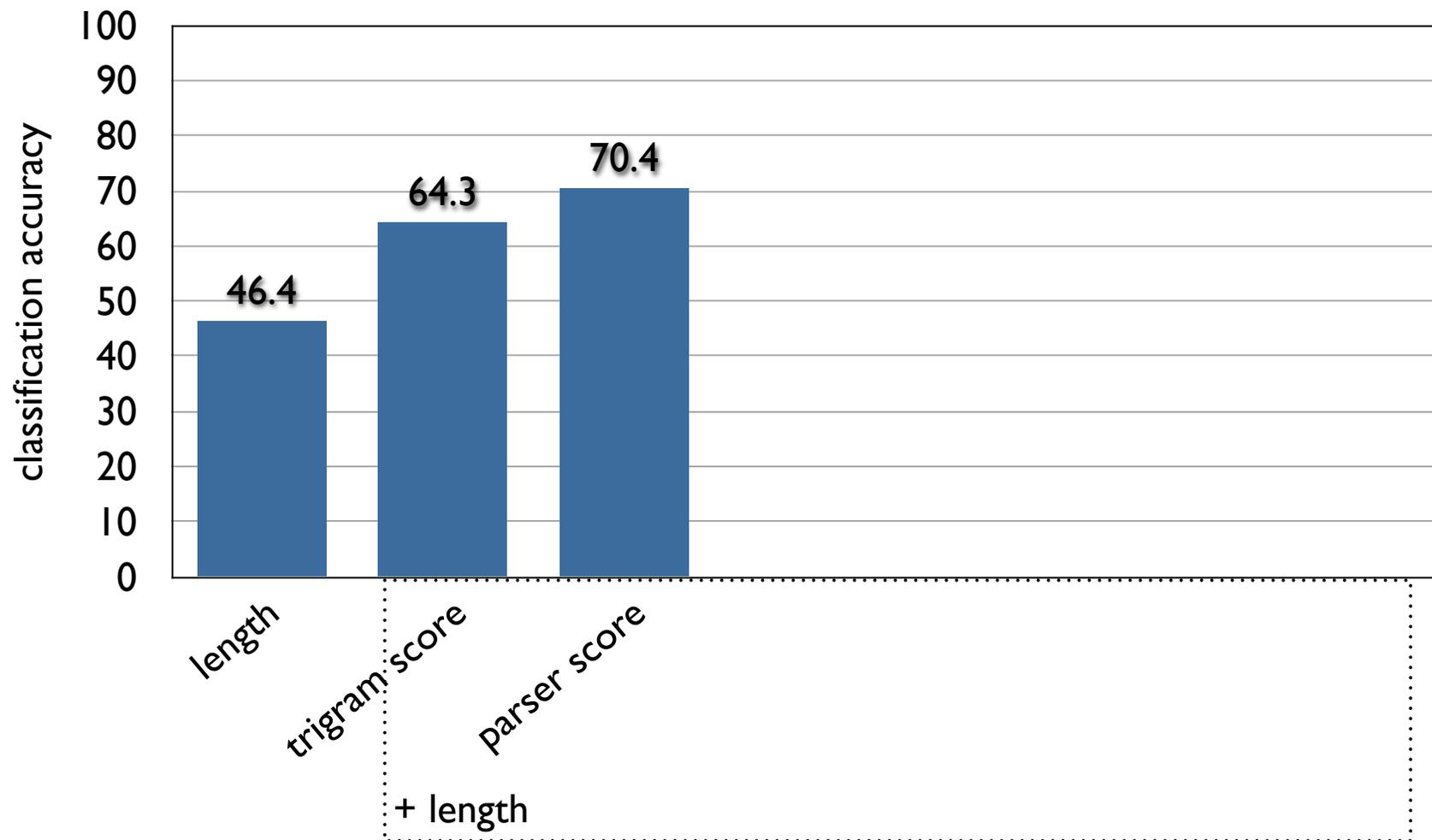
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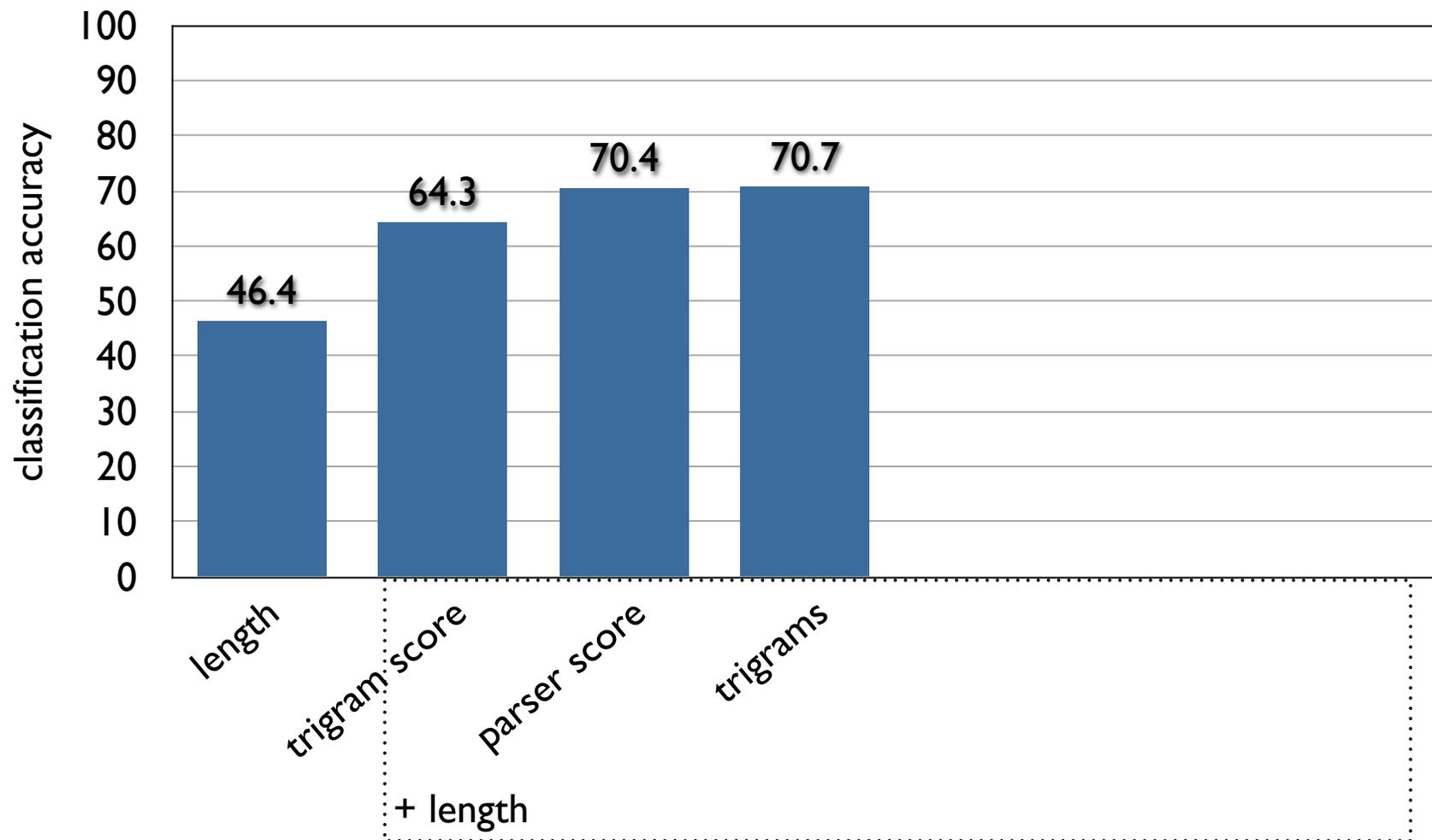
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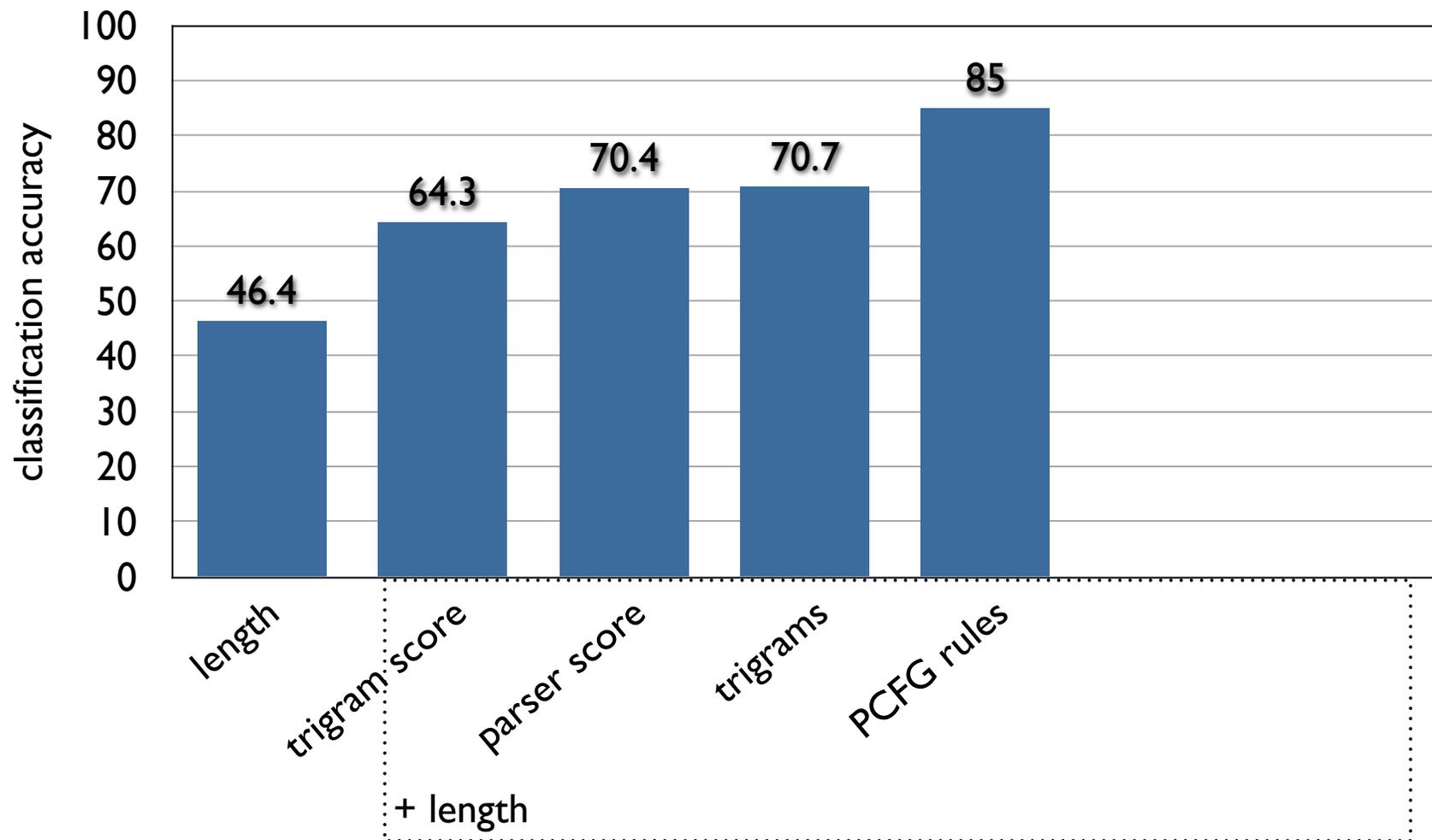
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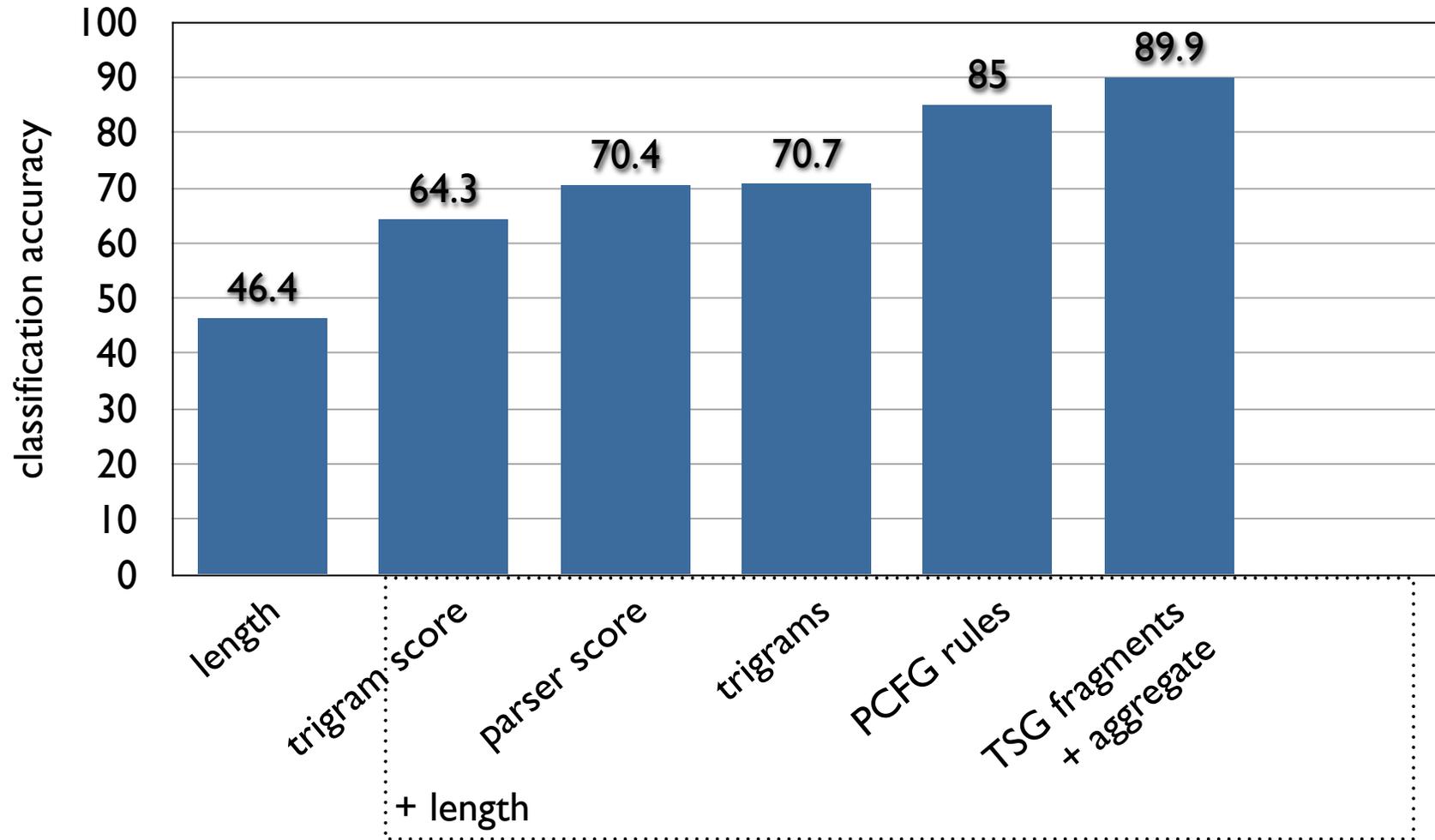
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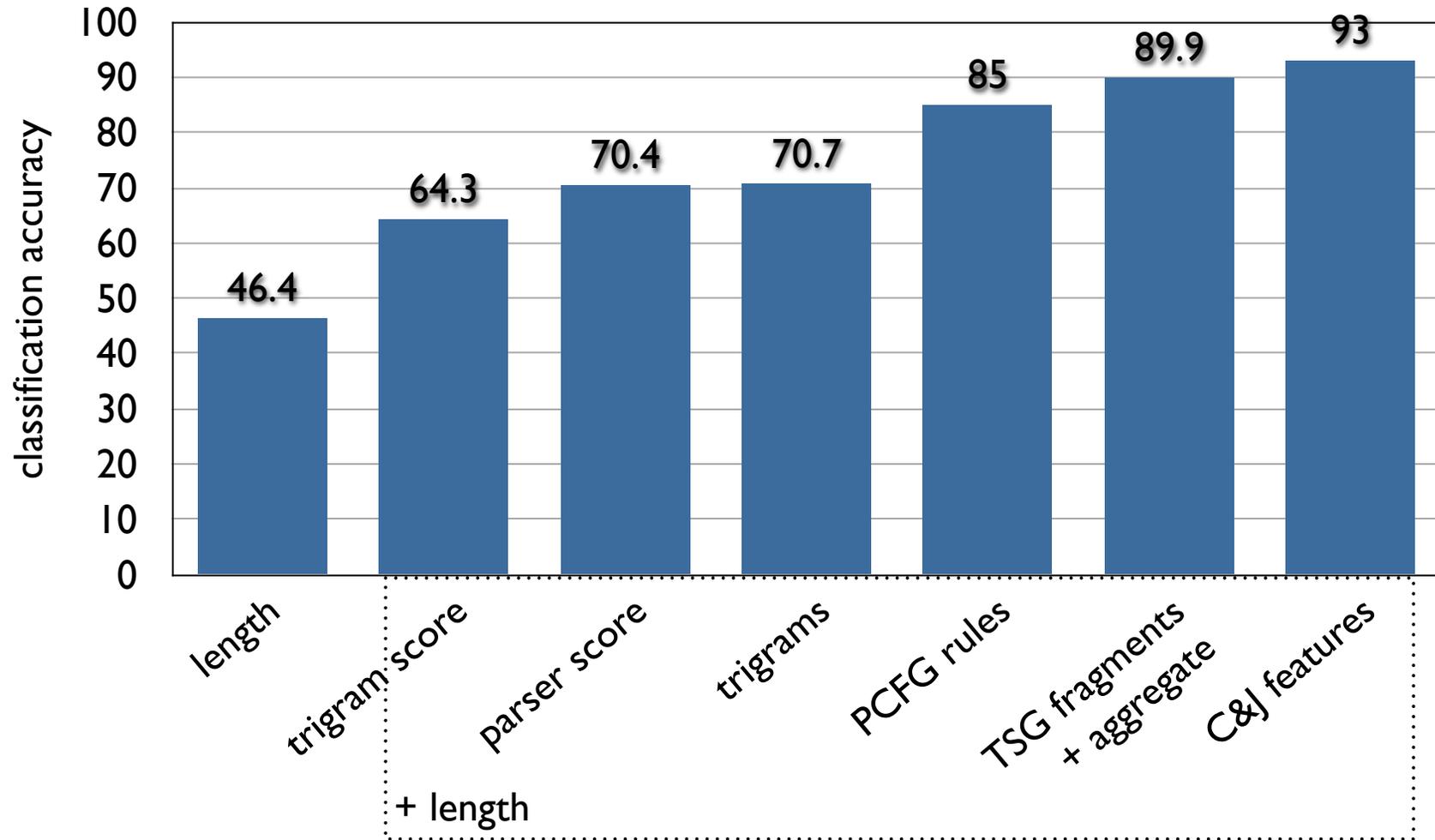
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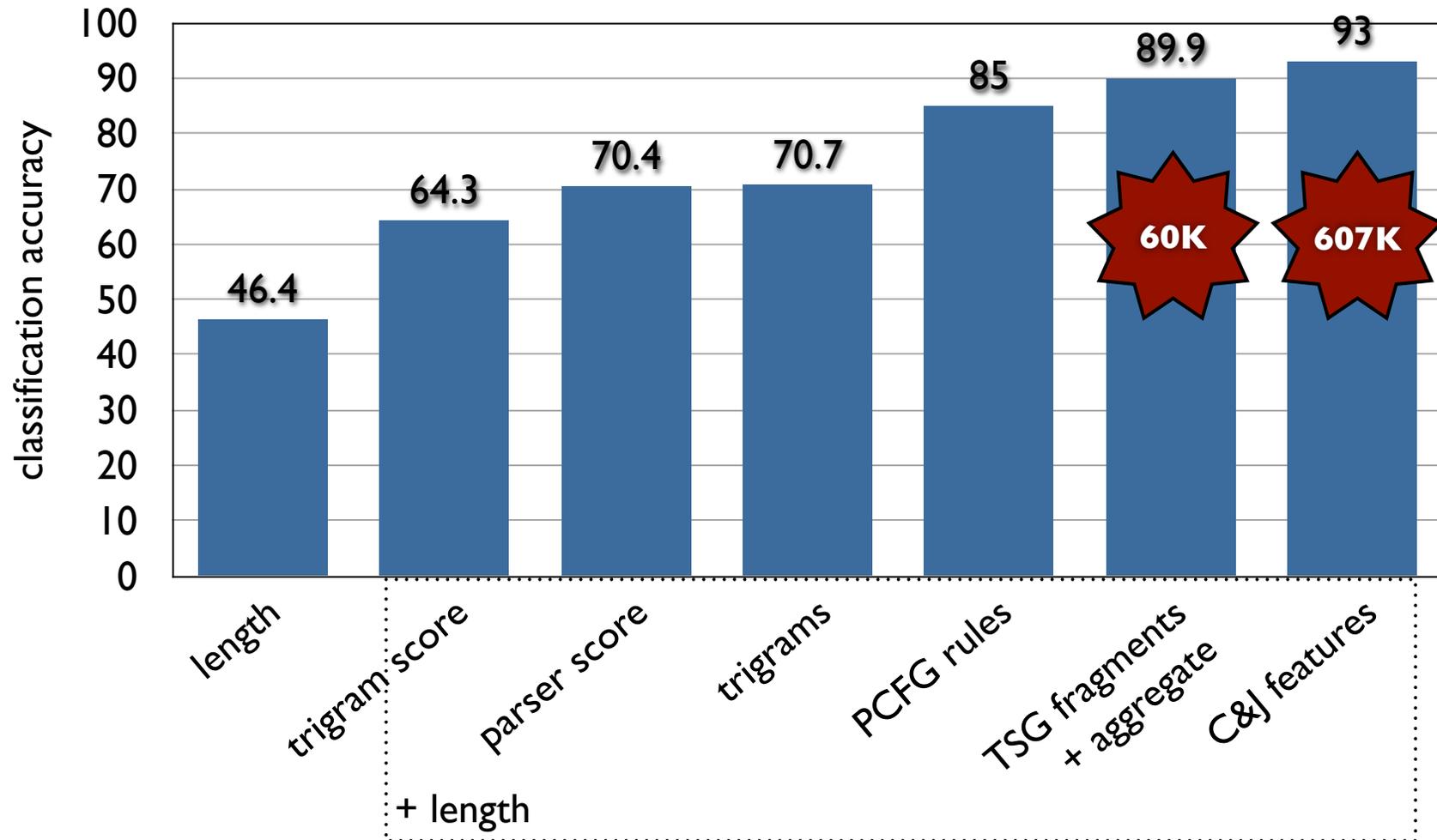
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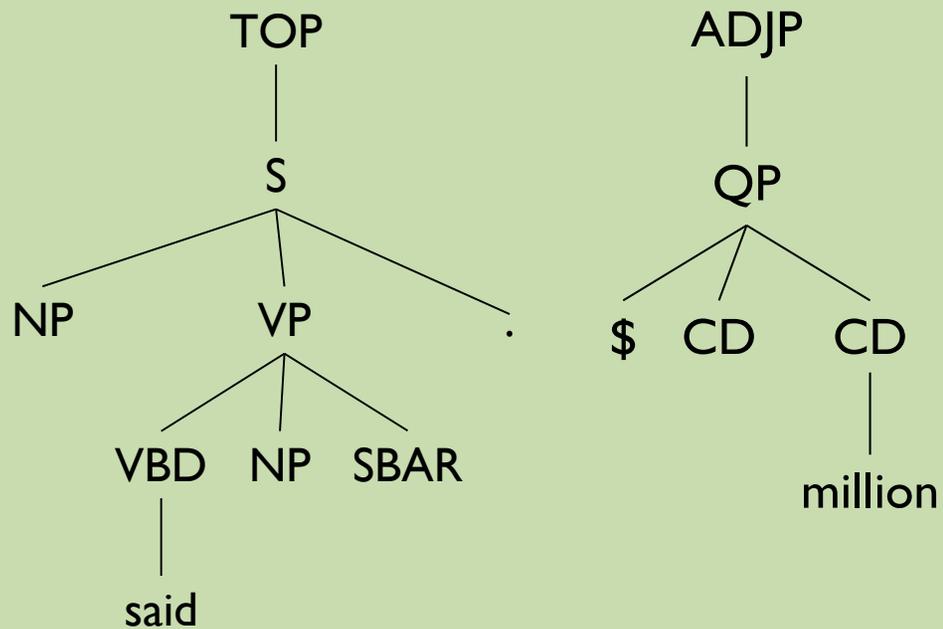
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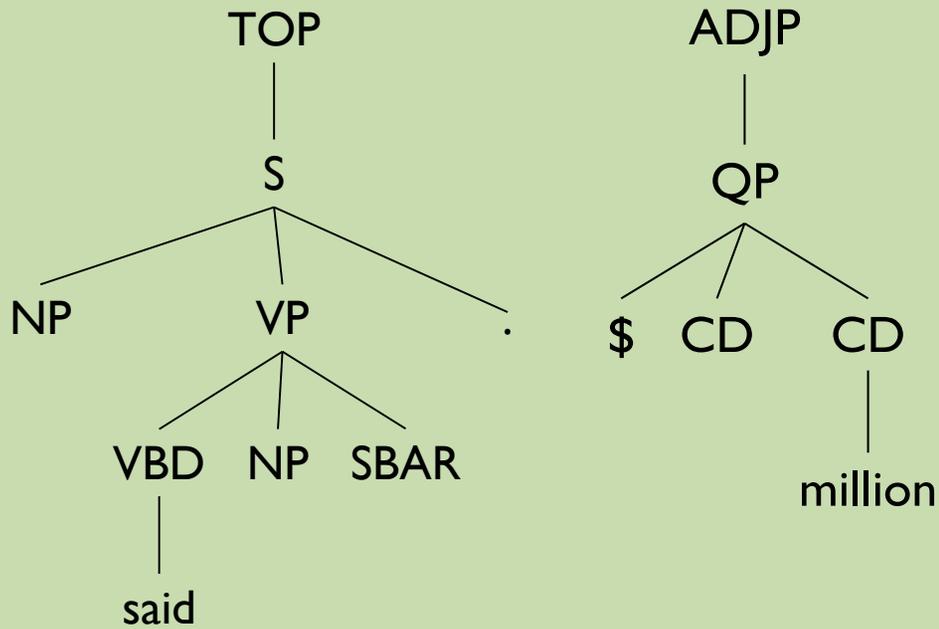
Features



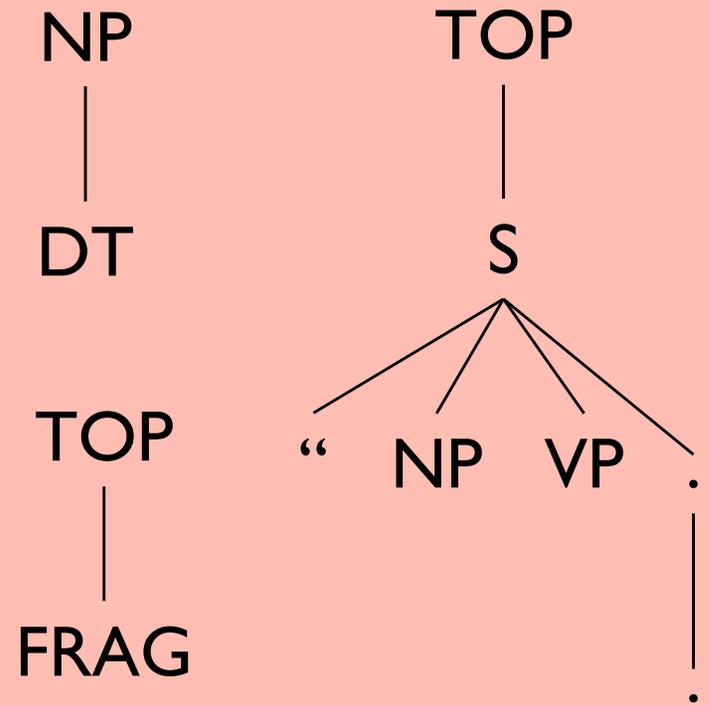
- fragments with two words
- fragments with a rank of 8



Features



- fragments with two words
- fragments with a rank of 8



- unlexicalized fragments



Conclusions

- TSGs performed well despite relatively poor parsing accuracy
- Shallow rules + coarse nonterminals => license for ungrammaticality
- Well below C&J features, but 1/10 the size, automatically discovered, and are a biproduct of the derivation



Future work

- Further baselines (implicit DOP representations, heuristic TSGs)
- Other discriminative tasks
 - more realistic negatives (Wagner et al., 2009; Tetreault et al., 2010)
 - MT classification and reranking
- Base measure changes



Thanks

Latent-variable TSGs

