

# Weakly-Supervised Bayesian Learning of a CCG Supertagger

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# Type-Level Supervision

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- Unannotated text
- Incomplete tag dictionary: word  $\mapsto$  {tags}

# Type-Level Supervision

Used for POS tagging for 20+ years

[Kupiec, 1992]  
[Merialdo, 1994]

# Type-Level Supervision

Good POS tagger performance  
even with low supervision

[Das & Petrov 2011]

[Garrette & Baldrige 2013]

[Garrette et al. 2013]

# Combinatory Categorial Grammar (CCG)

# CCG

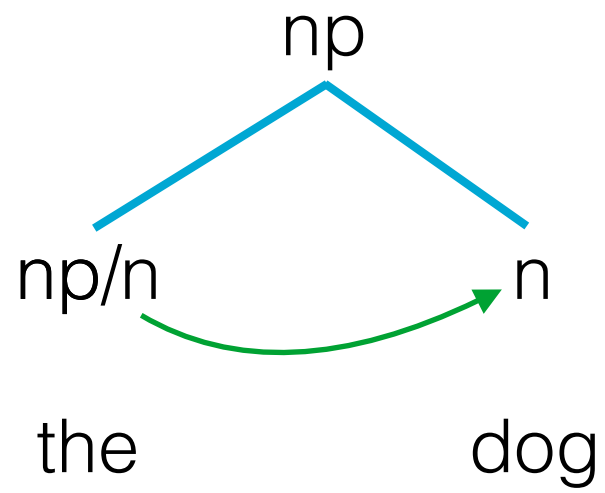
Every word token is associated with a **category**

Categories **combine** to categories of constituents

[Steedman, 2000]

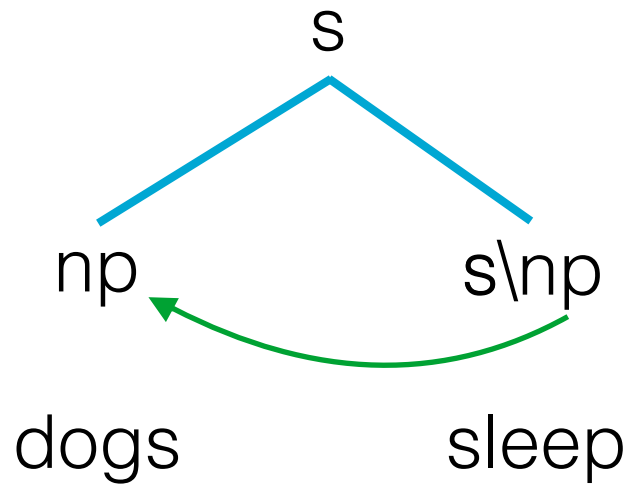
[Steedman and Baldrige, 2011]

# CCG

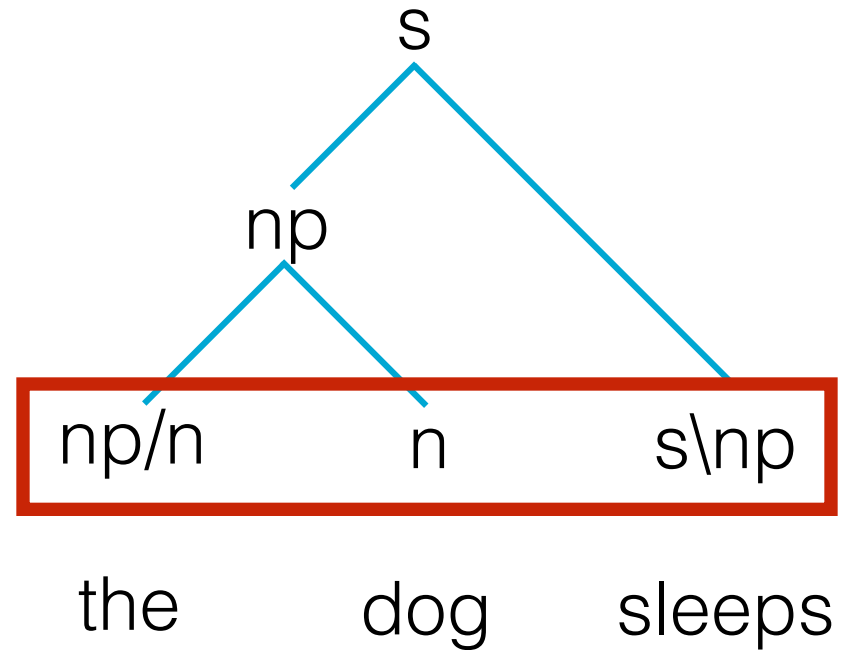
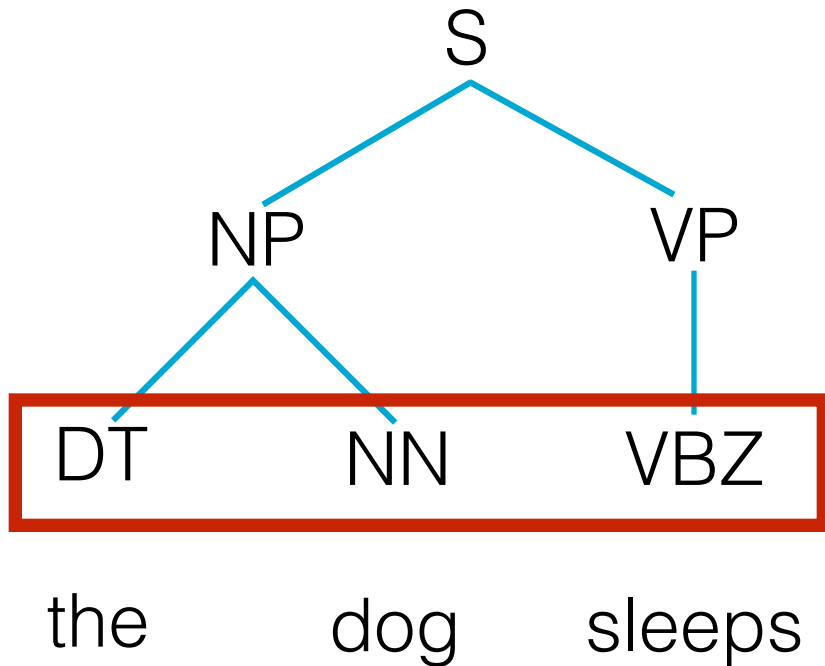




# CCG



# POS vs. Supertags



# Supertagging

Type-supervised learning for supertagging  
is much more difficult than for POS

Penn Treebank POS

**48 tags**

CCGBank Supertags

**1,239 tags**

# CCG

The grammar formalism *itself*  
can be used to guide learning

# CCG Supertagging

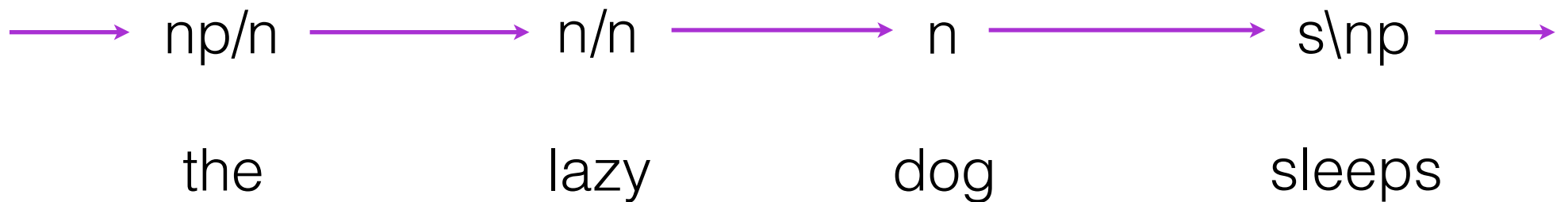
# CCG Supertagging

- Sequence tagging problem, like POS-tagging
- Building block for grammatical parsing

# Supertagging

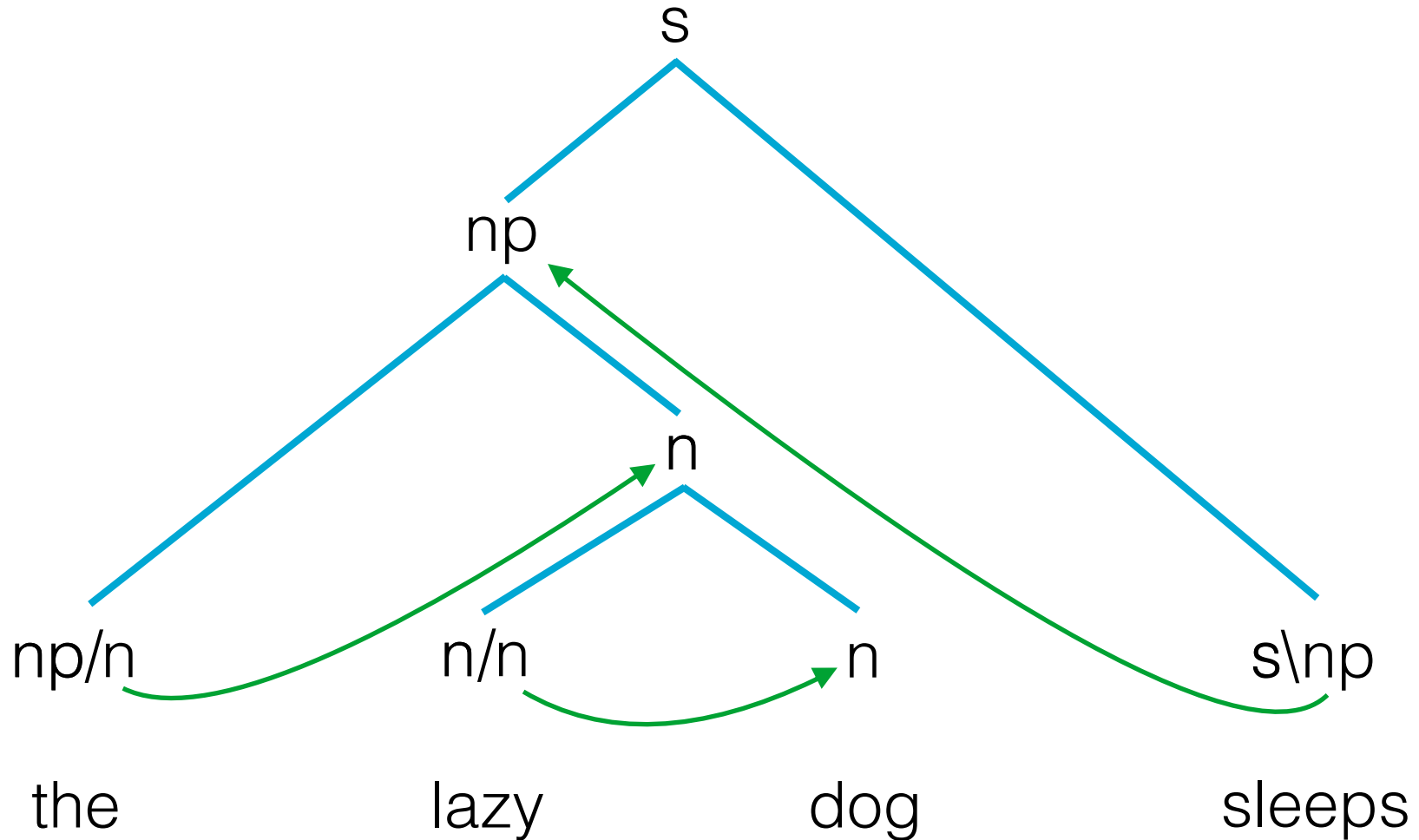
“almost parsing”

# Why Supertagging?

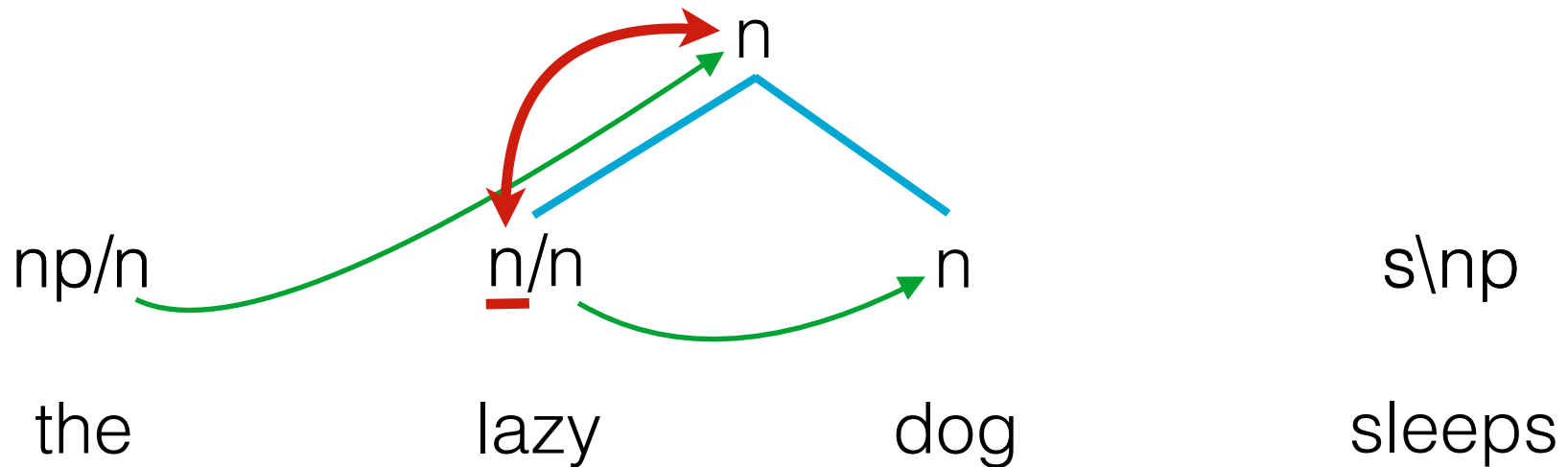




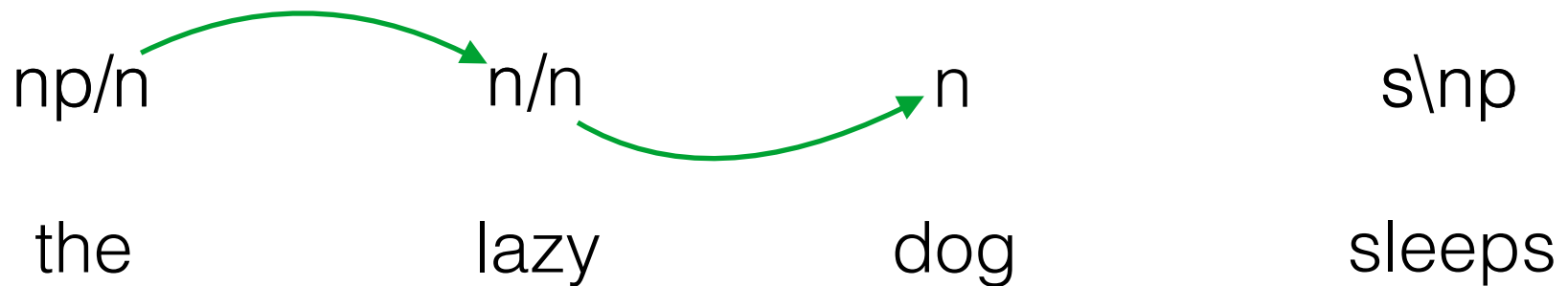
# Why Supertagging?



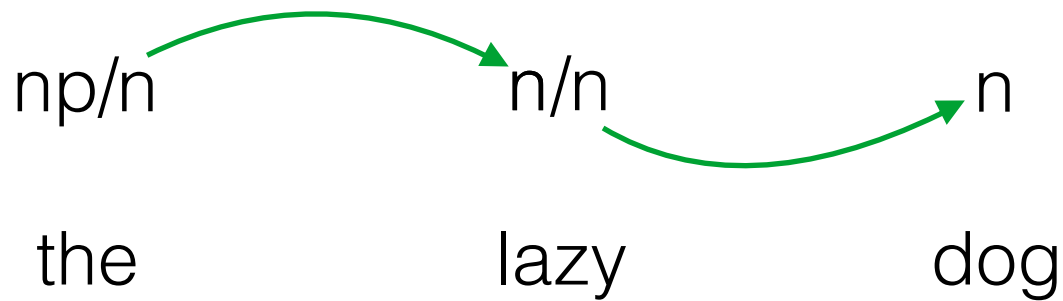
# CCG Supertagging



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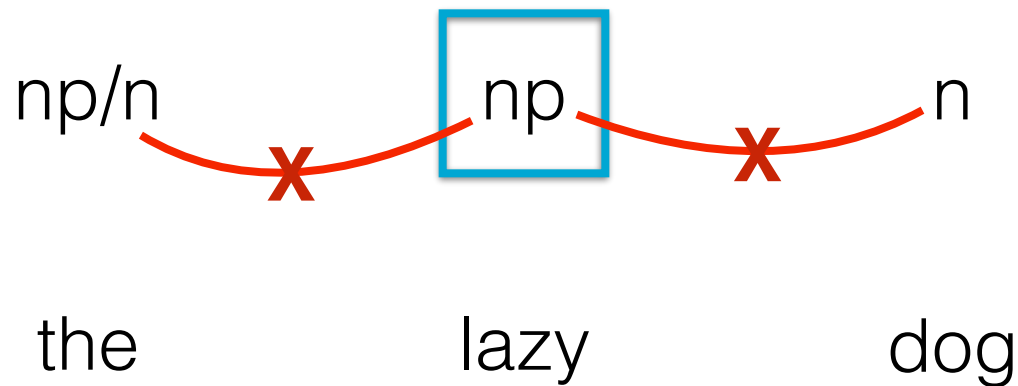


s\np  
sleeps

# CCG Supertagging

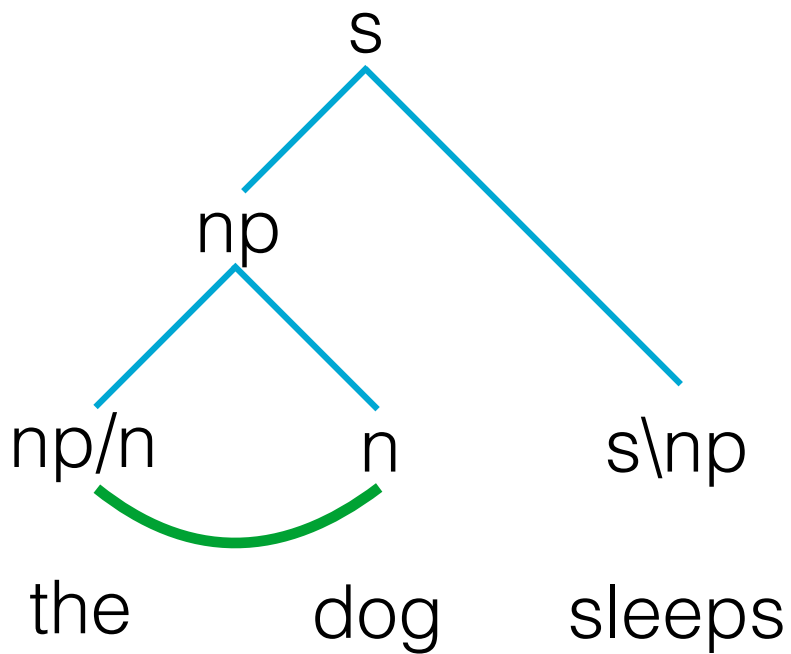


# Principle #1



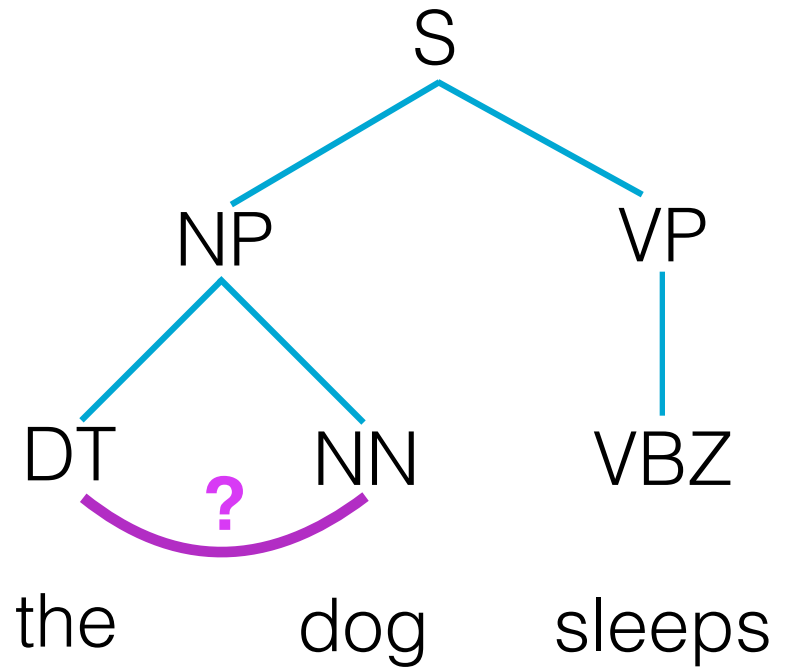
**Prefer Connections**

# Supertags vs. POS



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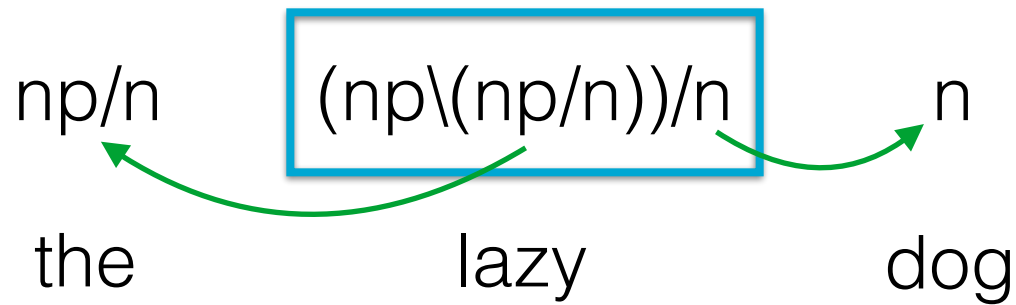
universal, intrinsic  
grammar properties



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all relationships  
must be learned

# Principle #2



**Prefer Simplicity**



# Prefer Simplicity

buy := (s<sub>b</sub>\np)/np

appears **342** times in CCGbank

e.g. “Opponents don't **buy** such arguments.”

buy := (((s<sub>b</sub>\np)/**pp**)/**pp**)/np

appears **once**

“Tele-Communications agreed to **buy** half of Showtime Networks from Viacom for \$ 225 million.”  
pp pp

# Weighted Tag Grammar

$a \longrightarrow \{s, np, n, \dots\} \quad p_{\text{atom}}(a) \times p_{\text{term}}$

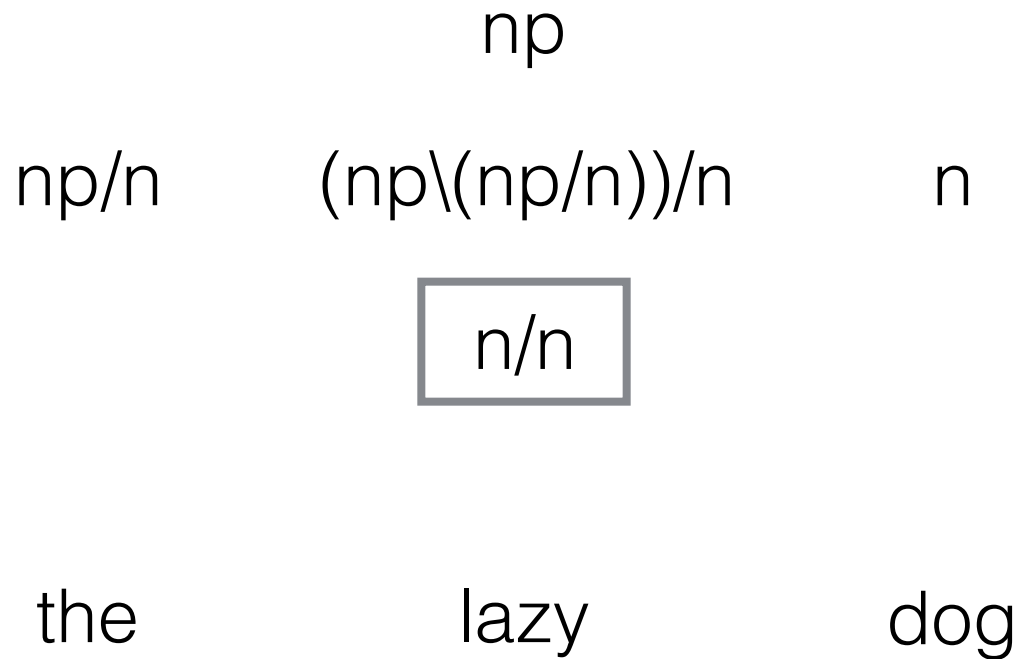
$A \longrightarrow B / B \quad \overline{p_{\text{term}}} \times p_{\text{fwd}} \times p_{\text{mod}}$

$A \longrightarrow B / C \quad \overline{p_{\text{term}}} \times p_{\text{fwd}} \times \overline{p_{\text{mod}}}$

$A \longrightarrow B \setminus B \quad \overline{p_{\text{term}}} \times \overline{p_{\text{fwd}}} \times p_{\text{mod}}$

$A \longrightarrow B \setminus C \quad \overline{p_{\text{term}}} \times \overline{p_{\text{fwd}}} \times \overline{p_{\text{mod}}}$

# CCG Supertagging



# HMM Transition Prior

$$P(\mathbf{t} \rightarrow \mathbf{u}) = \lambda \cdot \underbrace{P(\mathbf{u})}_{\text{simple is good}} + (1-\lambda) \cdot \underbrace{P(\mathbf{t} \rightarrow \mathbf{u})}_{\text{connecting is good}}$$

# Type-Supervised Learning

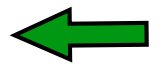
unlabeled corpus

tag dictionary



same as  
POS tagging

universal properties of the CCG formalism



Training

# Posterior Inference

Forward-Filter Backward-Sample (FFBS)

# Posterior Inference

Unlabeled Data

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Tag Dictionary

— : —, —, —  
— : —, —, —  
— : —, —, —  
— : —, —, —  
— : —, —, —

the

lazy

dogs

wander

np/n

n/n

n

n

np

np

n/n

(s\np)/np

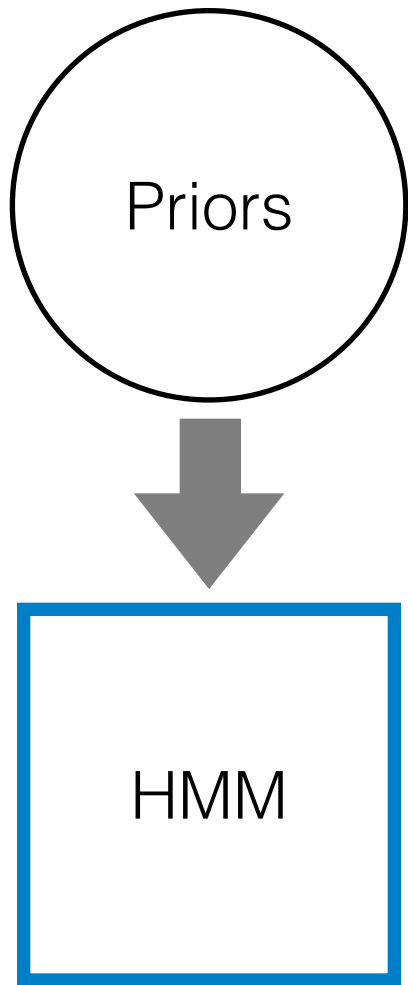
np/n

s\np

...

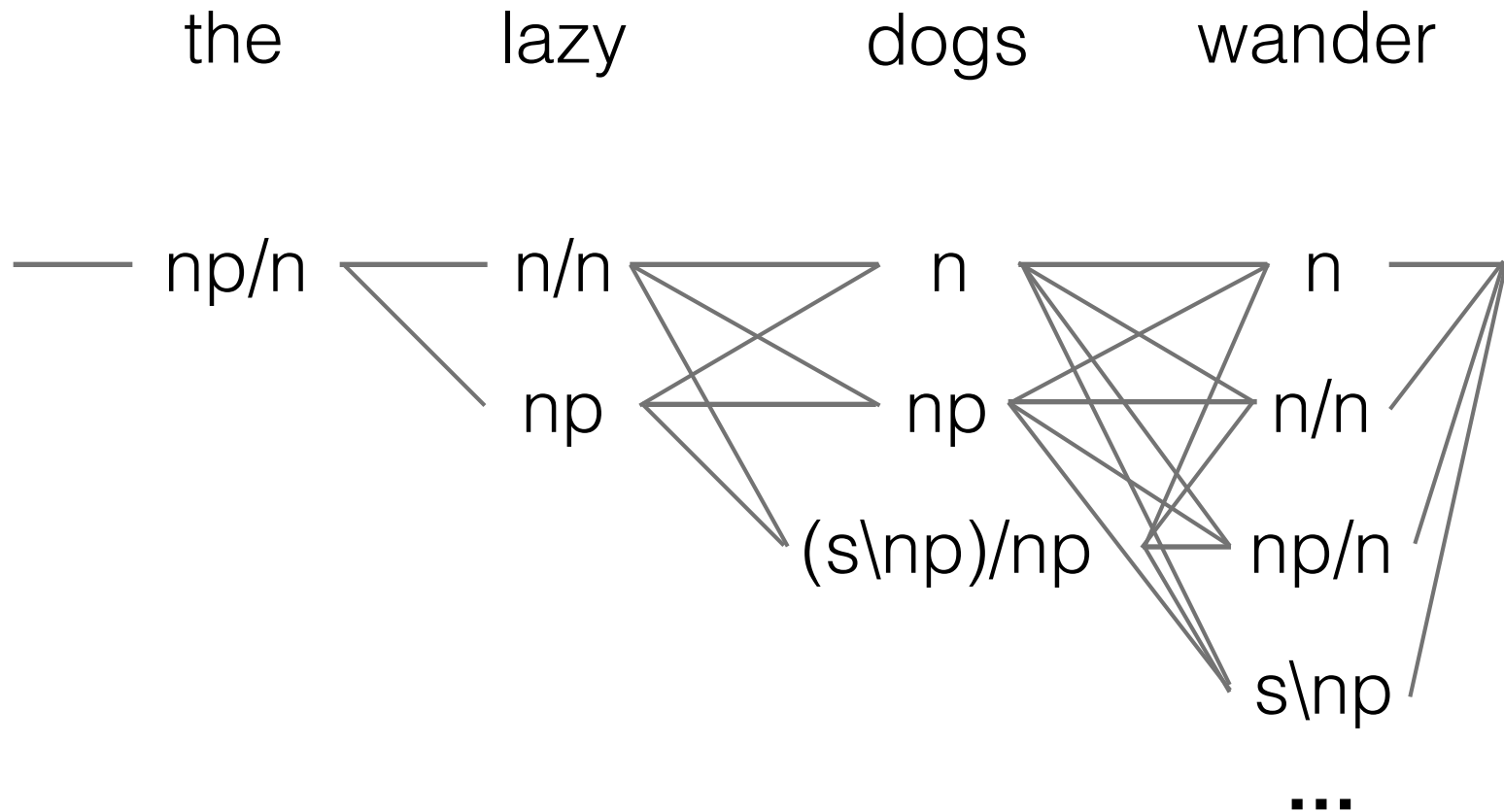
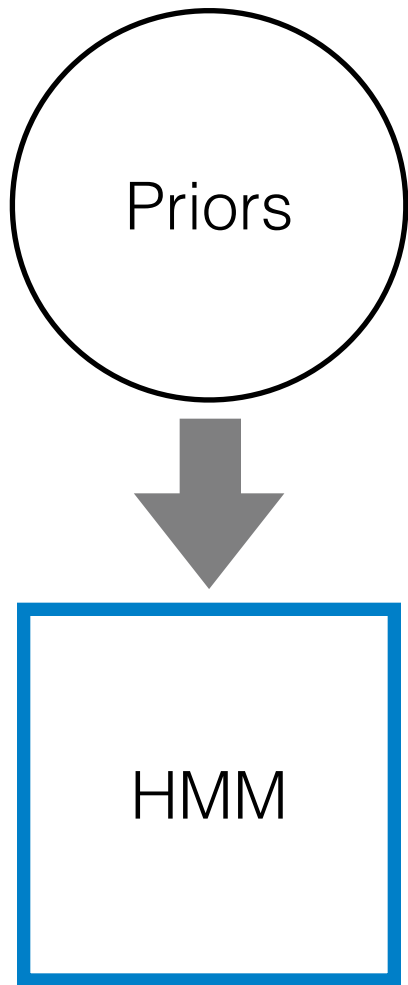


# Posterior Inference

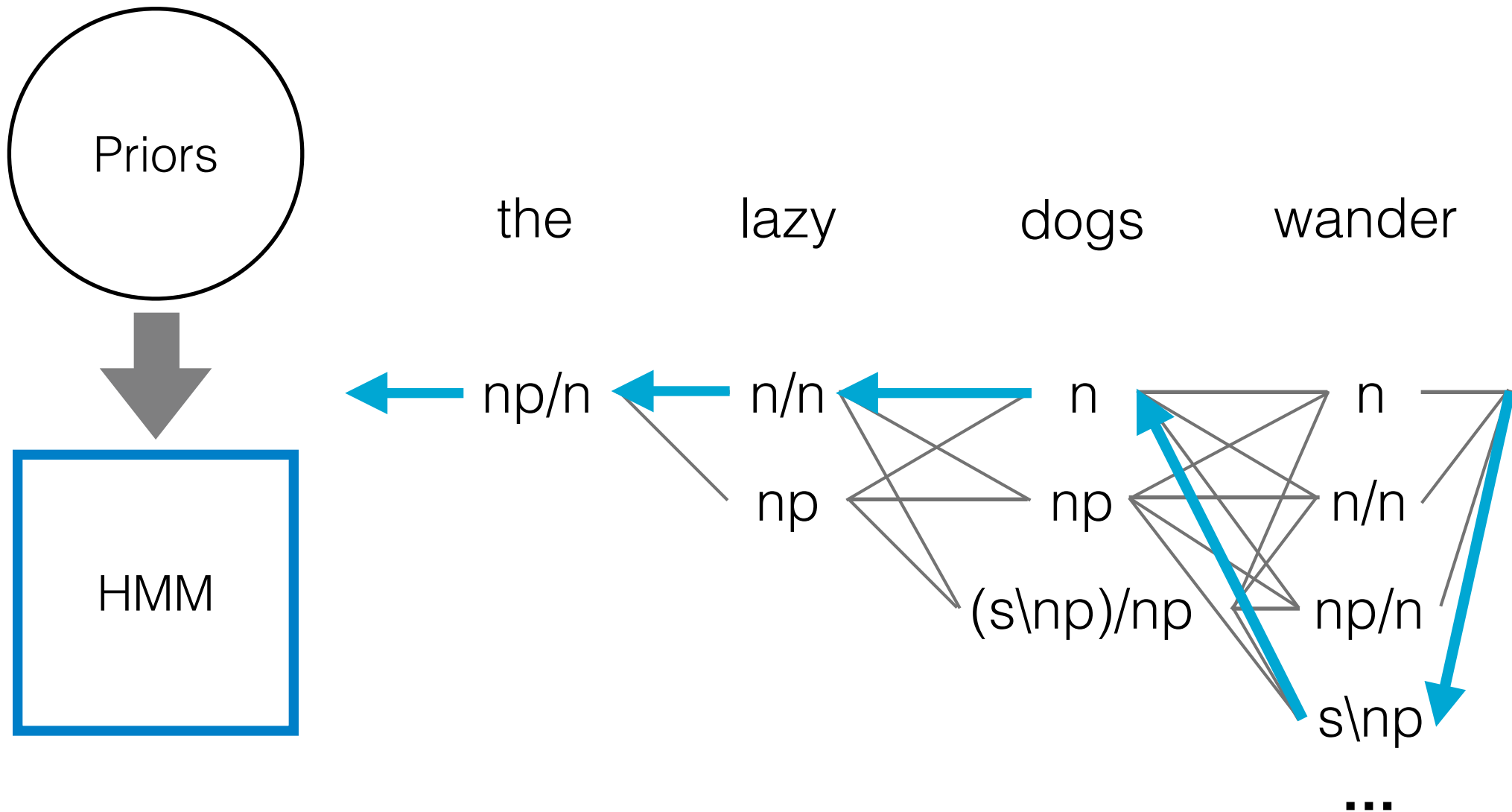


the	lazy	dogs	wander
np/n	n/n	n	n
	np	np	n/n
		(s\np)/np	np/n
			s\np
			...

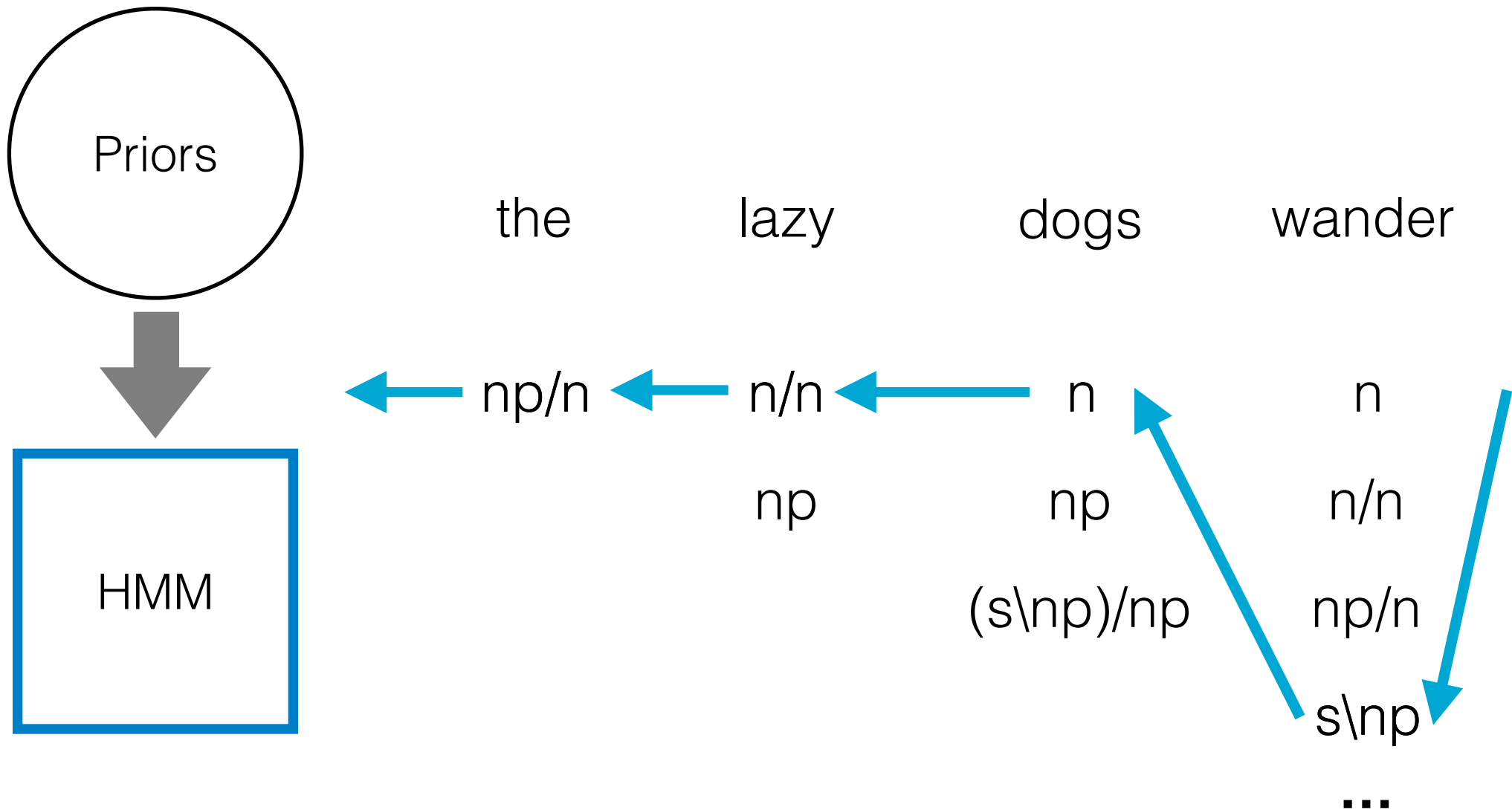
# Posterior Inference



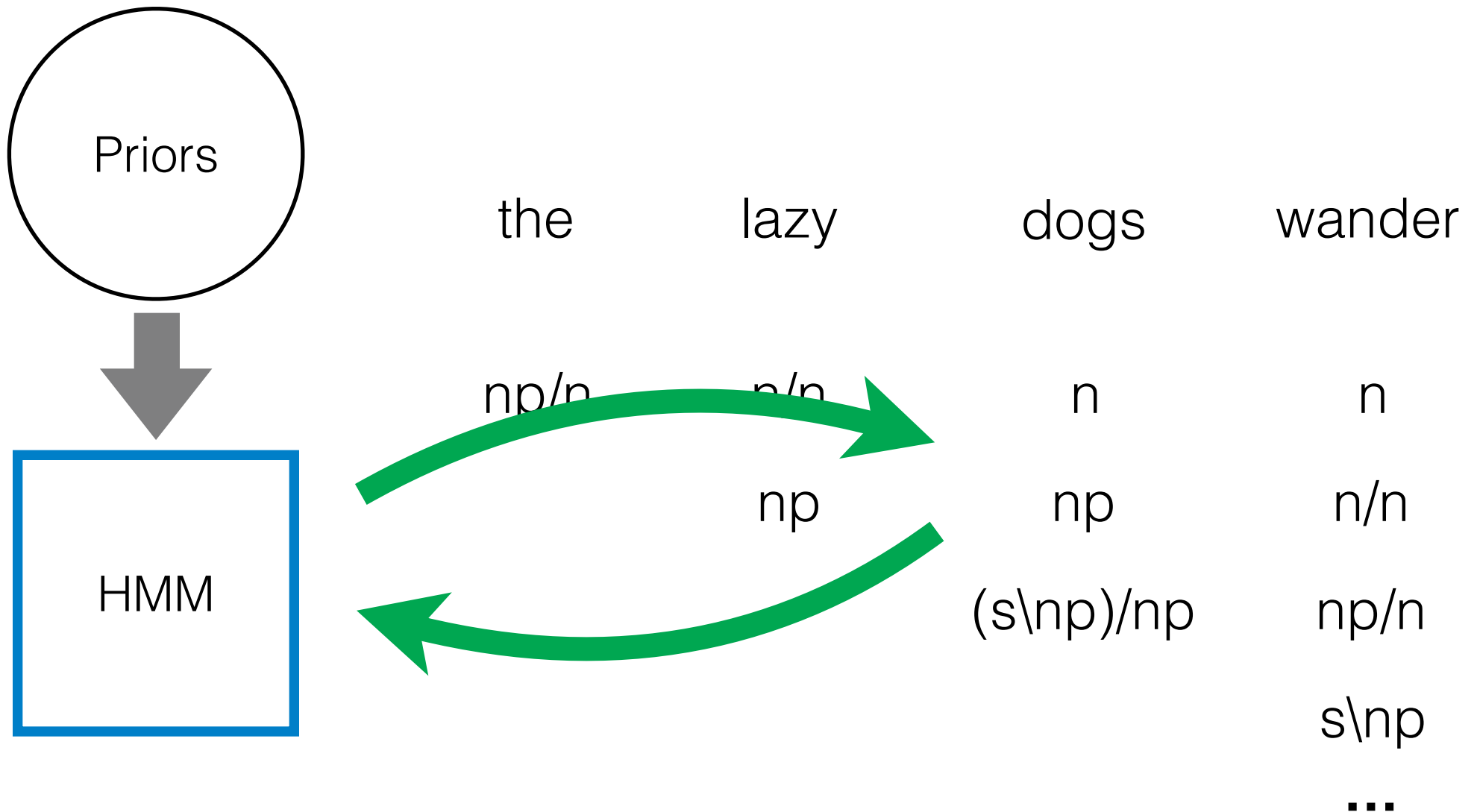
# Posterior Inference



# Posterior Inference



# Posterior Inference



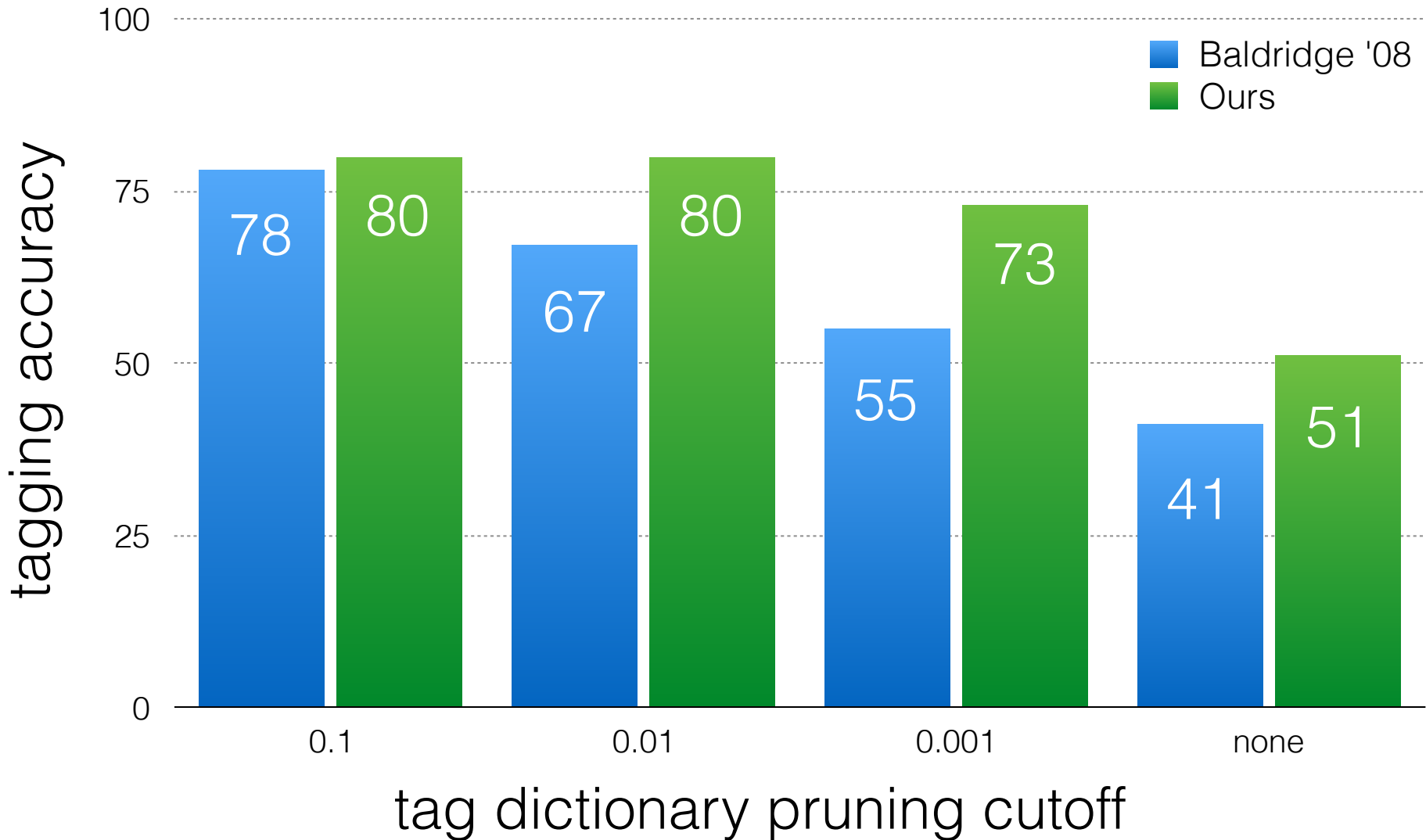
# Experiments

# Baldridge 2008

Use universal properties of CCG to initialize EM

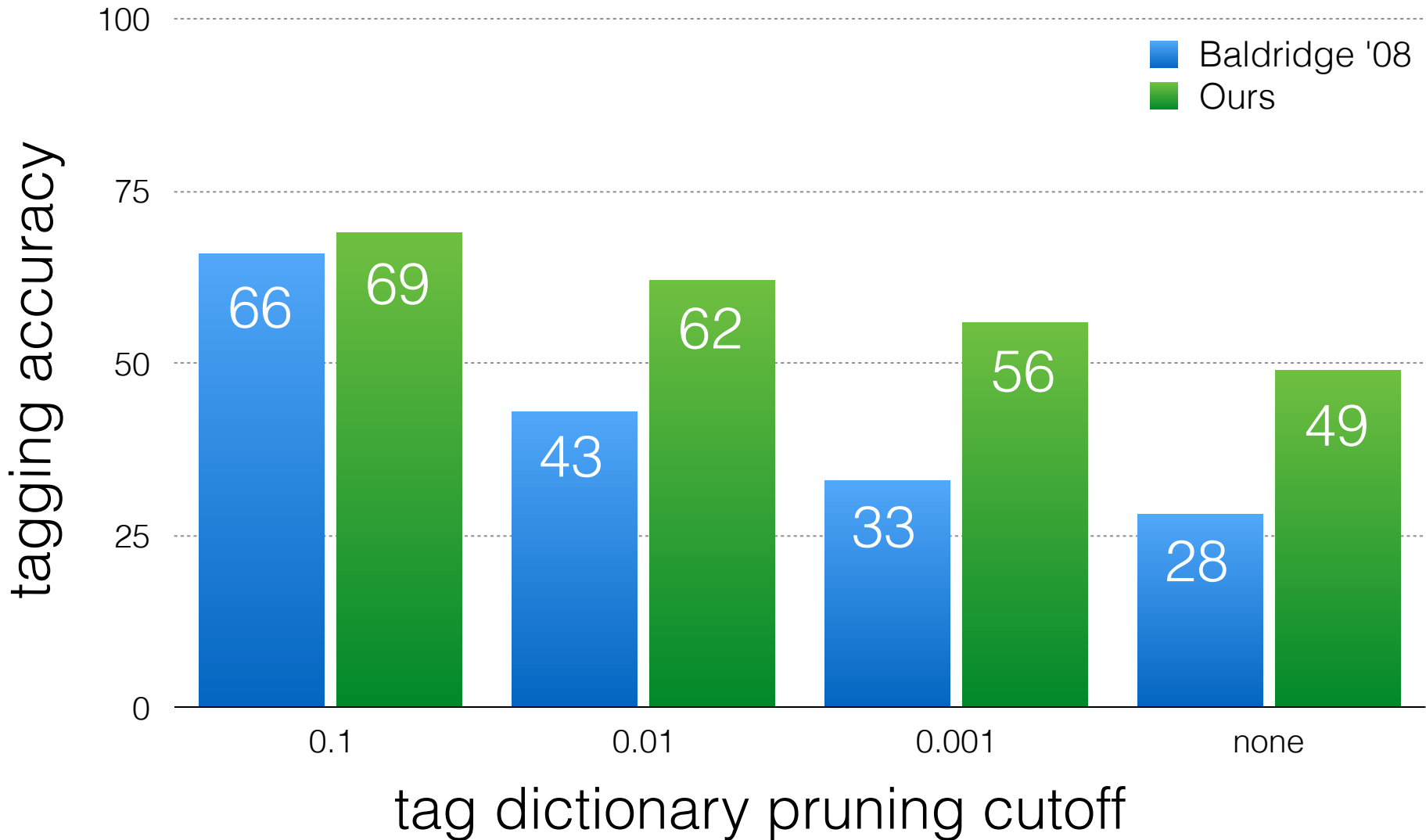
- Simpler definition of category complexity
- No corpus-specific information

# English Supertagging

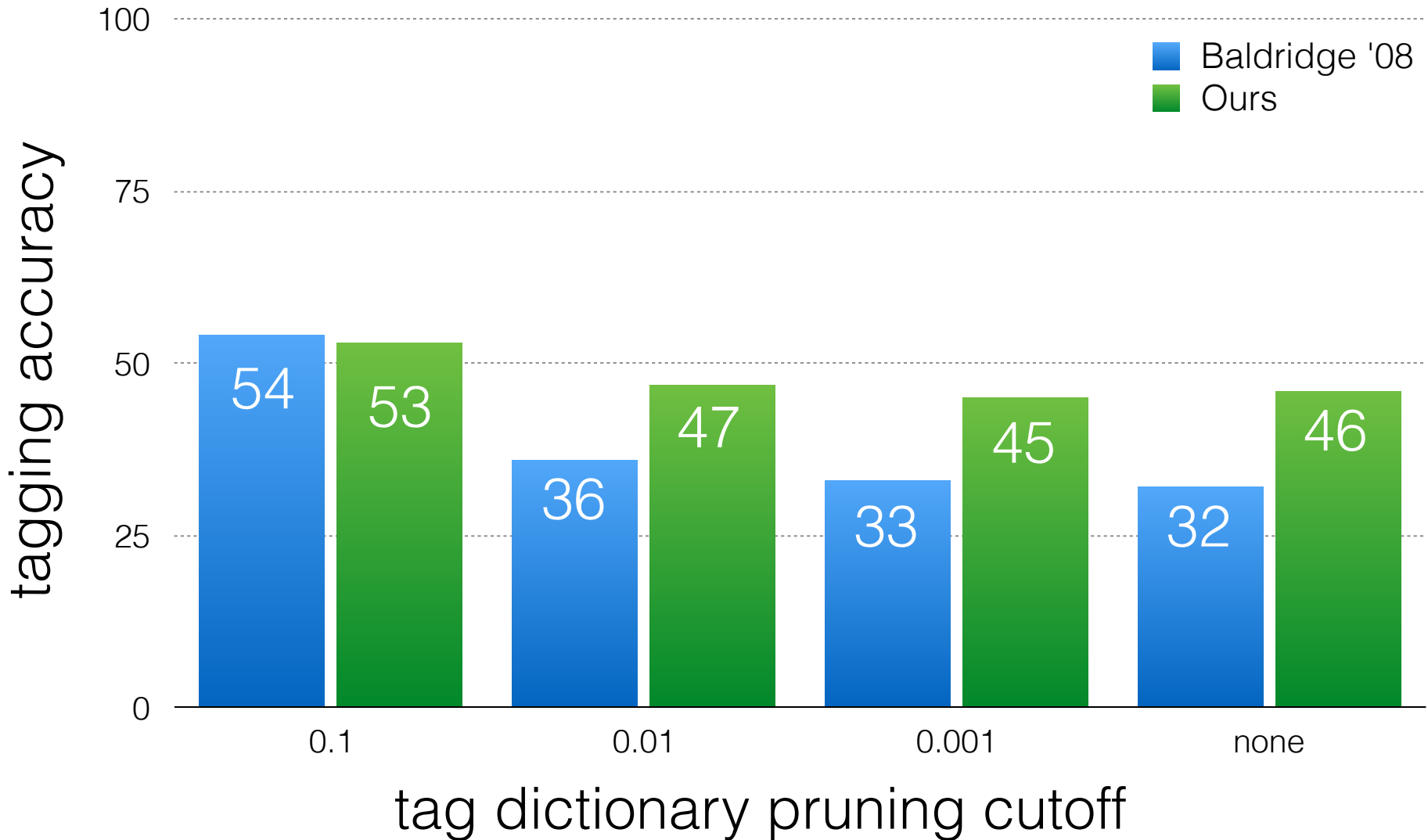




# Chinese Supertagging



# Italian Supertagging



# Code Available

GitHub repository  
linked from my website

# Conclusion

Combining annotation exploitation with  
**universal grammatical knowledge**  
yields good models  
from weak supervision