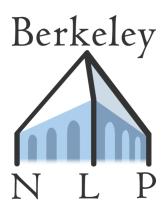
Easy Victories and Uphill Battles in Coreference Resolution

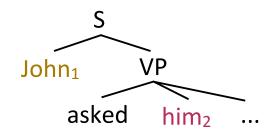


Greg Durrett and Dan Klein
UC Berkeley



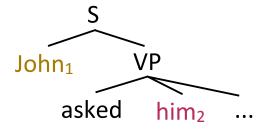


Binding theory (Chomsky, 1981)





Binding theory (Chomsky, 1981)



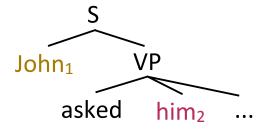
Pronoun agreement (Hobbs, 1977)

Male Female Female

John₁ talked to Jane₂. She₂ asked...



Binding theory (Chomsky, 1981)



Pronoun agreement (Hobbs, 1977)

MALE FEMALE FEMALE

John₁ talked to Jane₂. She₂ asked...

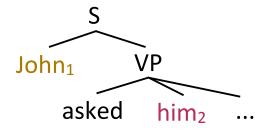
Centering (Grosz et al., 1995)

SUBJ. OBJ. SUBJ.

John talked to Bill. He asked...



Binding theory (Chomsky, 1981)



Pronoun agreement (Hobbs, 1977)

Male Female Female

John₁ talked to Jane₂. She₂ asked...

Centering (Grosz et al., 1995)

SUBJ. OBJ. SUBJ.

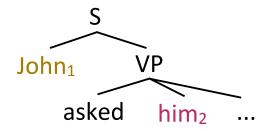
John₁ talked to Bill₂. He₁ asked...

Semantic compatibility

The president₁ ... The leader₁



Binding theory (Chomsky, 1981)



Pronoun agreement (Hobbs, 1977)

Male Female Female

John₁ talked to Jane₂. She₂ asked...

Centering (Grosz et al., 1995)

SUBJ. OBJ. SUBJ.

John₁ talked to Bill₂. He₁ asked...

Semantic compatibility

The president₁ ... The leader₁

Definiteness

The president

A president





The president



The president

```
if (startsWith("the"))
   DEFINITE
else if (startsWith("a"))
   INDEFINITE
else
   NO_ART
```



The president if (startsWith("the")) DEFINITE else if (startsWith("a")) INDEFINITE else NO ART **DEFINITE** INDEFINITE NO ART



Definiteness: Data-Driven

The president



Definiteness: Data-Driven



Definiteness: Data-Driven

```
The president

The these that

U.S. all his no no an some more John their Barack Israeli
```

• • •



Classical approach



Classical approach

 \blacktriangleright Learning with heuristic features $\left\{ \text{ Soon et al. (2001), inter alia} \right\}$



Classical approach

- \blacktriangleright Learning with heuristic features $\left\{ \text{ Soon et al. (2001), inter alia} \right\}$
- Rule-based

```
{ Soon et al. (2001), inter aliand Haghighi and Klein (2009), Lee et al. (2011)
```



Classical approach

- ▶ Learning with heuristic features { Soon et al. (2001), inter alia
- Rule-based

Mixed approach

Add data-driven features on a few axes

```
egin{cases} {
m Soon~et~al.~(2001),~inter~alia)} \ {
m Haghighi~and~Klein~(2009),} \ {
m Lee~et~al.~(2011)} \end{cases}
```

Bengtson and Roth (2008), Rahman and Ng (2011), Björkelund and Nugues (2011)



Classical approach

- ▶ Learning with heuristic features { Soon et al. (2001), inter alia
- Rule-based

Soon et al. (2001), inter aliand Haghighi and Klein (2009), Lee et al. (2011)

Mixed approach

Add data-driven features on a few axes Bengtson and Roth (2008), Rahman and Ng (2011), Björkelund and Nugues (2011)

Data-driven approach (this work)

>400,000 comprehensive, uniform features





[Voters]₁ agree when [they]₁ are given [a chance]₂ to decide if [they]₁ ...



 A_1

New



[Voters]₁ agree when [they]₁ are given [a chance]₂ to decide if [they]₁ ...



$$egin{array}{cccc} A_1 & & A_2 \\ \textit{New} & & \textit{New} \\ \circlearrowleft & & 1 & \circlearrowleft \end{array}$$

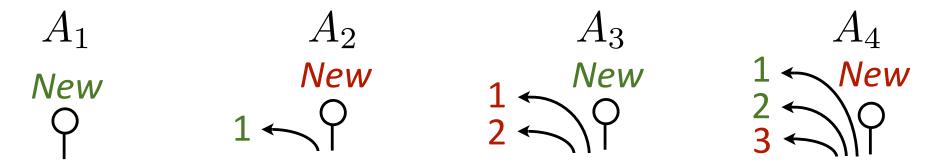
[Voters]₁ agree when [they]₁ are given [a chance]₂ to decide if [they]₁ ...





[Voters]₁ agree when [they]₁ are given [a chance]₂ to decide if [they]₁ ...

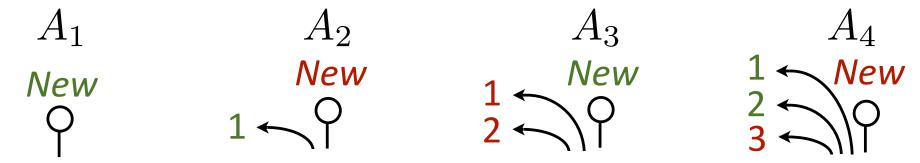




[Voters]₁ agree when [they]₁ are given [a chance]₂ to decide if [they]₁ ...



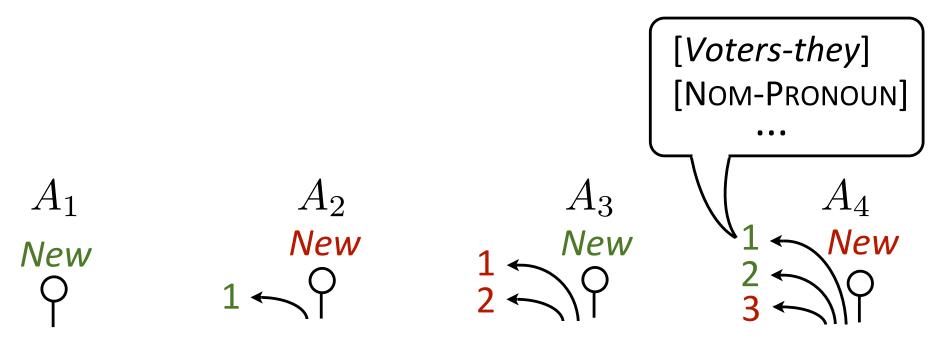
$$Pr(A_i = a|x) \propto \exp(w^{\top} f(i, a, x))$$



[Voters]₁ agree when [they]₁ are given [a chance]₂ to decide if [they]₁ ...



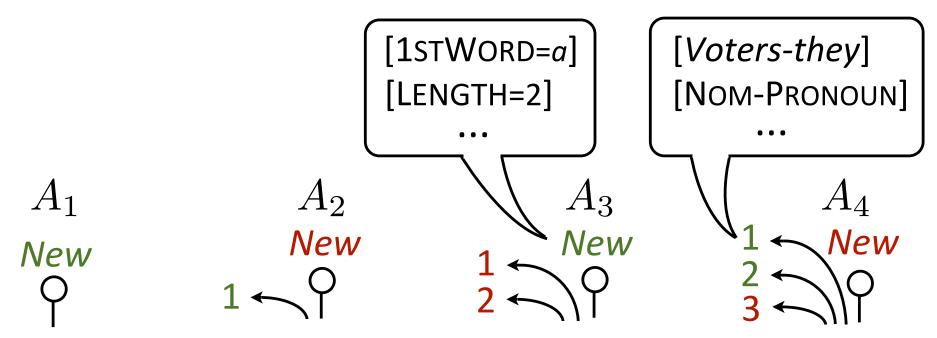
$$Pr(A_i = a|x) \propto \exp(w^{\top} f(i, a, x))$$



[Voters]₁ agree when [they]₁ are given [a chance]₂ to decide if [they]₁ ...



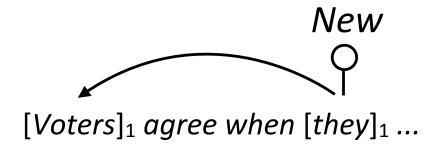
$$Pr(A_i = a|x) \propto \exp(w^{\top} f(i, a, x))$$



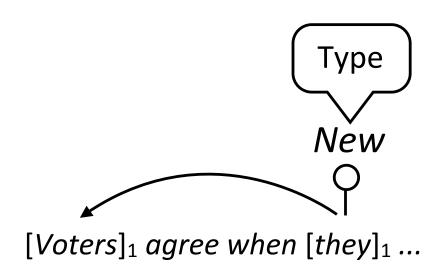
[Voters]₁ agree when [they]₁ are given [a chance]₂ to decide if [they]₁ ...



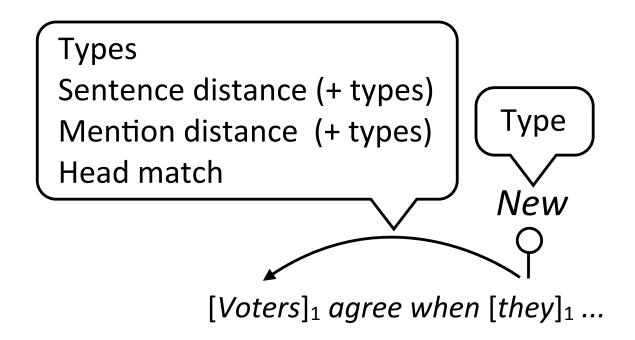














55



50

45 44.6

(Dev set CoNLL-F1, predicted mentions)



Definiteness

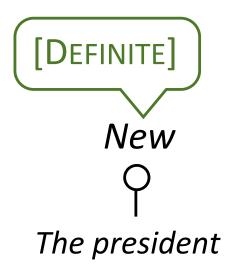


Definiteness

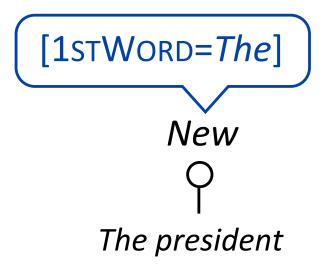
New
O
The president



Definiteness









55

BASIC

50

45
44.6
40
(Dev set CoNLL-F1, predicted mentions)

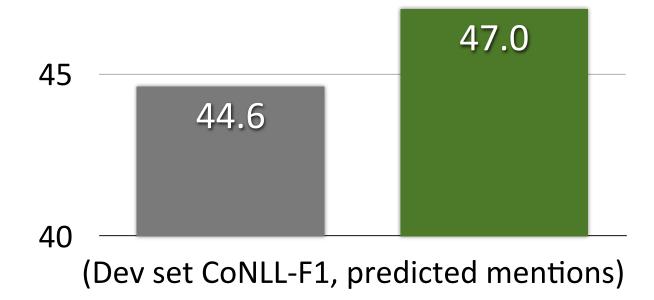


55

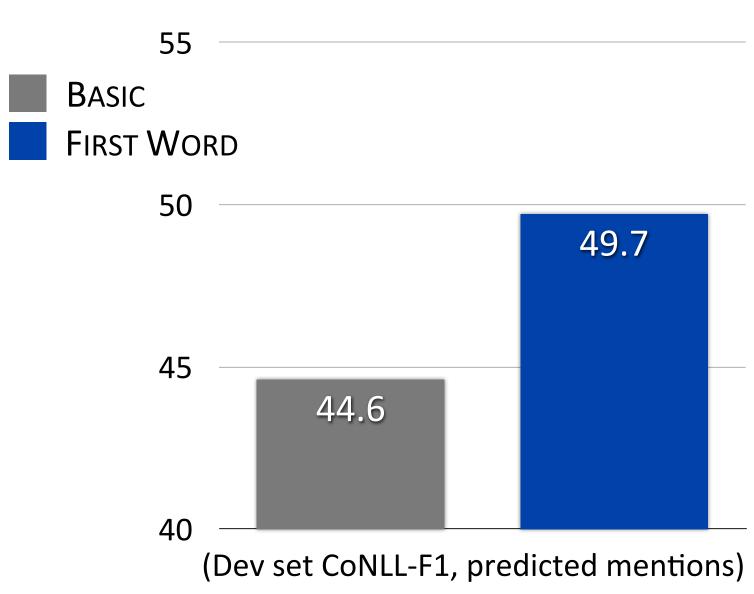
BASIC

DEFINITENESS

50









DEFINITENESS 47.0



DEFINITENESS

47.0

FIRST WORD: "the, a, an"

47.0



DEFINITENESS

47.0

FIRST WORD: "the, a, an"

47.0

+ "some, all, no" + 9 more

47.7

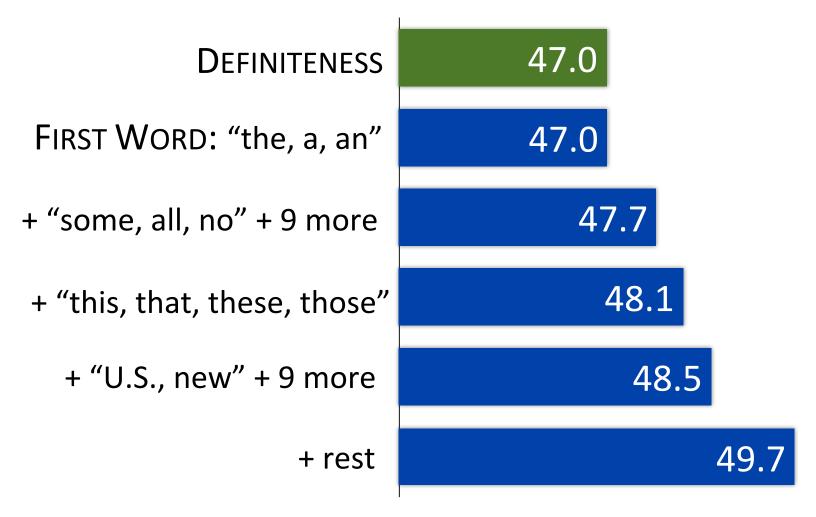


DEFINITENESS	47.0
FIRST WORD: "the, a, an"	47.0
+ "some, all, no" + 9 more	47.7
+ "this, that, these, those"	48.1



Definiteness	47.0
FIRST WORD: "the, a, an"	47.0
+ "some, all, no" + 9 more	47.7
+ "this, that, these, those"	48.1
+ "U.S., new" + 9 more	48.5



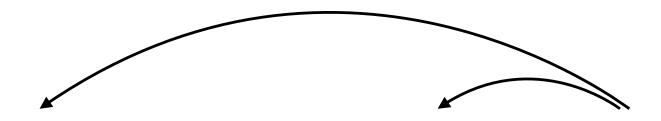






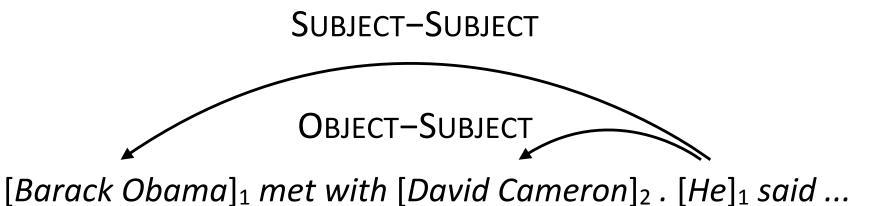
[Barack Obama]₁ met with [David Cameron]₂. [He]₁ said ...



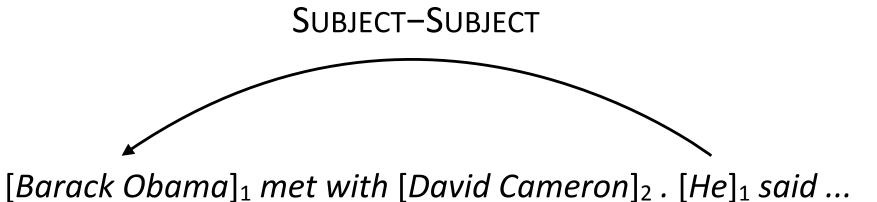


 $[Barack\ Obama]_1\ met\ with\ [David\ Cameron]_2\ .\ [He]_1\ said\ ...$









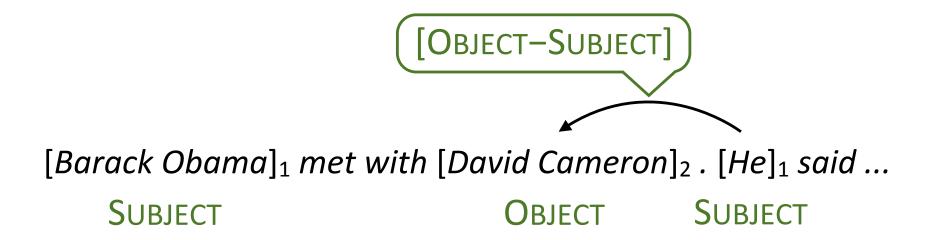


[Barack Obama]₁ met with [David Cameron]₂. [He]₁ said ...



 $[Barack\ Obama]_1\ met\ with\ [David\ Cameron]_2\ .\ [He]_1\ said\ ...$ Subject Subject







 $[Barack\ Obama]_1\ met\ with\ [David\ Cameron]_2\ .\ [He]_1\ said\ ...$



[Barack Obama]₁ met with [David Cameron]₂ . [He]₁ said ... with [X] .



[Barack Obama] $_1$ met with [David Cameron] $_2$. [He] $_1$ said ... with [X] . . [X] said

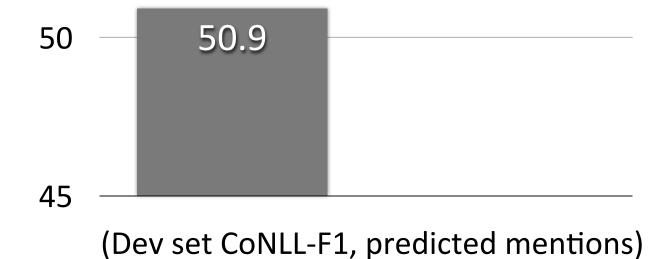


```
[with \ X-.\ Y] \\ [with \ X-Y \ said] \\ ... [Barack\ Obama]_1\ met\ with\ [David\ Cameron]_2\ .\ [He]_1\ said\ ... with\ [X]\ .\qquad .\ [X]\ said
```

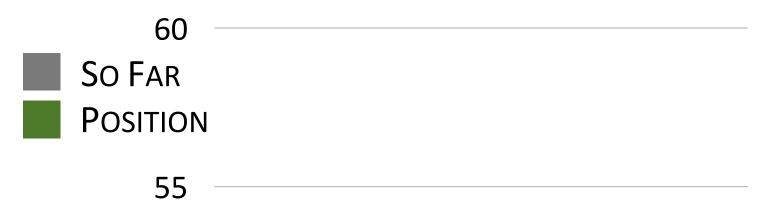


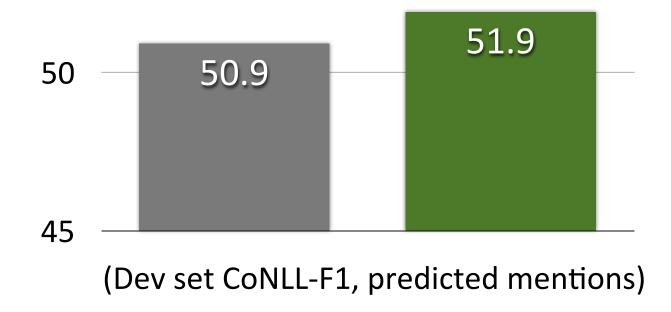


55

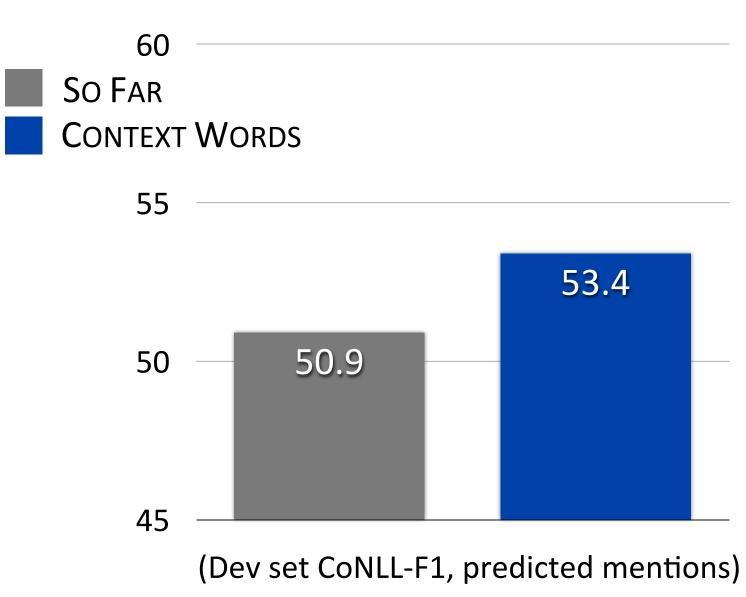










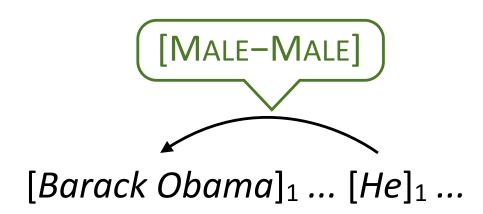




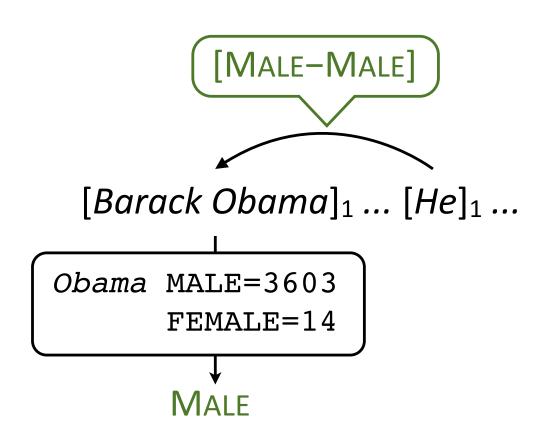


 $[Barack\ Obama]_1 \dots [He]_1 \dots$



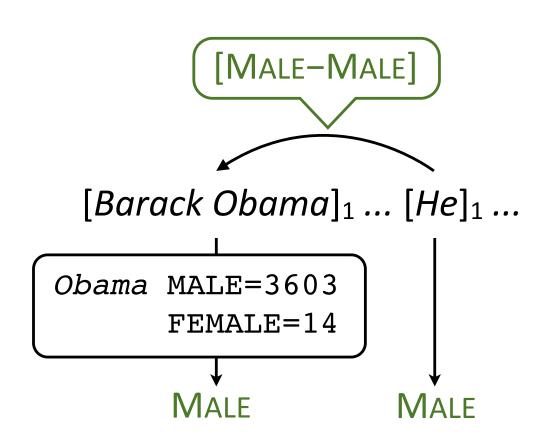






Bergsma and Lin (2006)



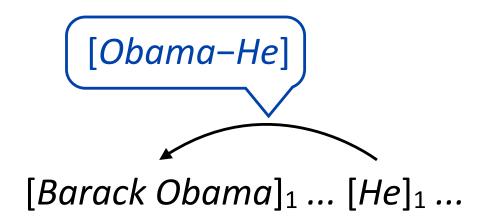


Bergsma and Lin (2006)











60

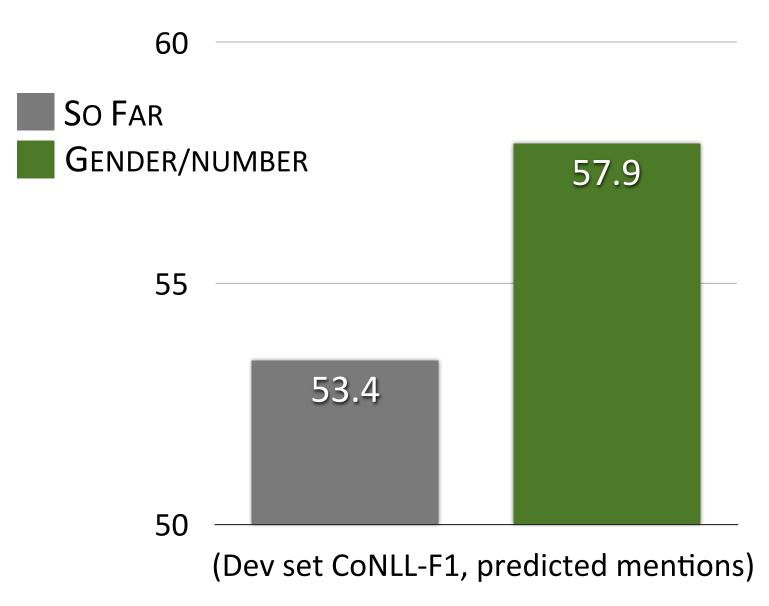
So Far

55

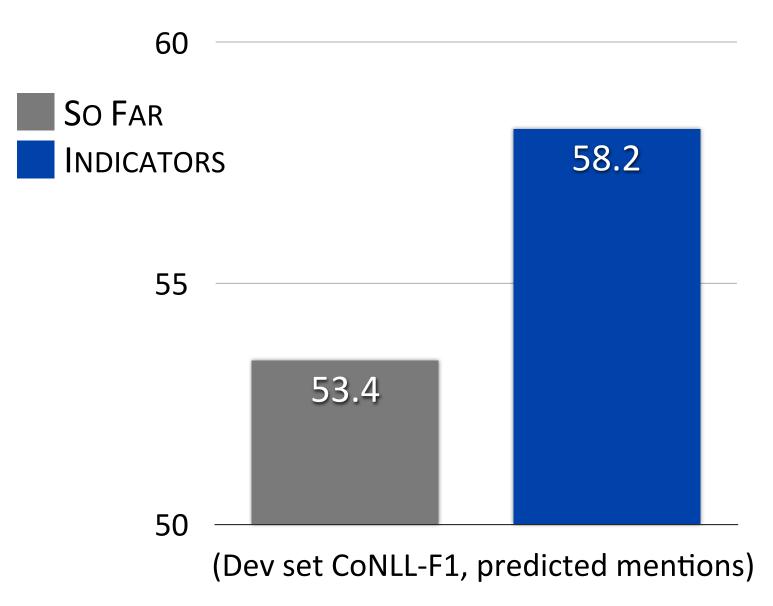
53.4

(Dev set CoNLL-F1, predicted mentions)











SURFACE Information

Features conjoin surface-level mention attributes



Features conjoin surface-level mention attributes

<s>[President Barack Obama] signed the bill ... Afterwards [he] said ...



Features conjoin surface-level mention attributes

<s>[President Barack Obama] signed the bill ... Afterwards [he] said ...



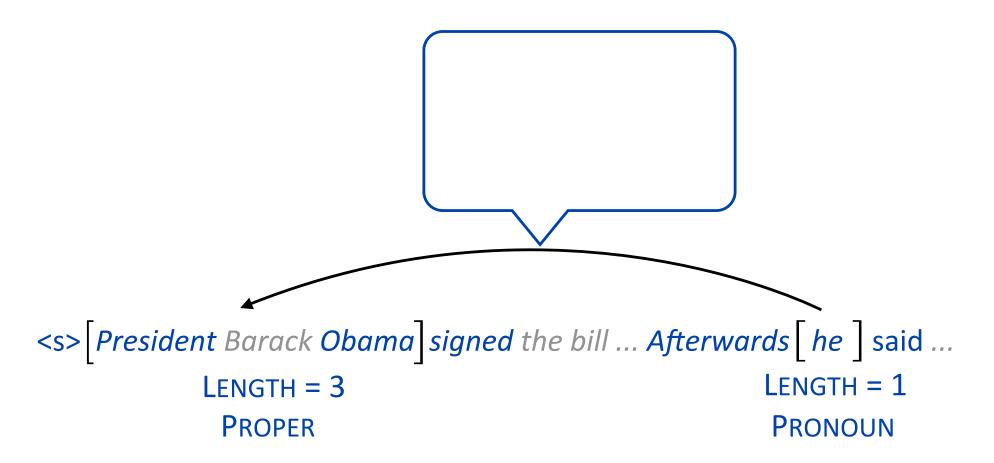
Features conjoin surface-level mention attributes

<s>[President Barack Obama] signed the bill ... Afterwards [he] said ...

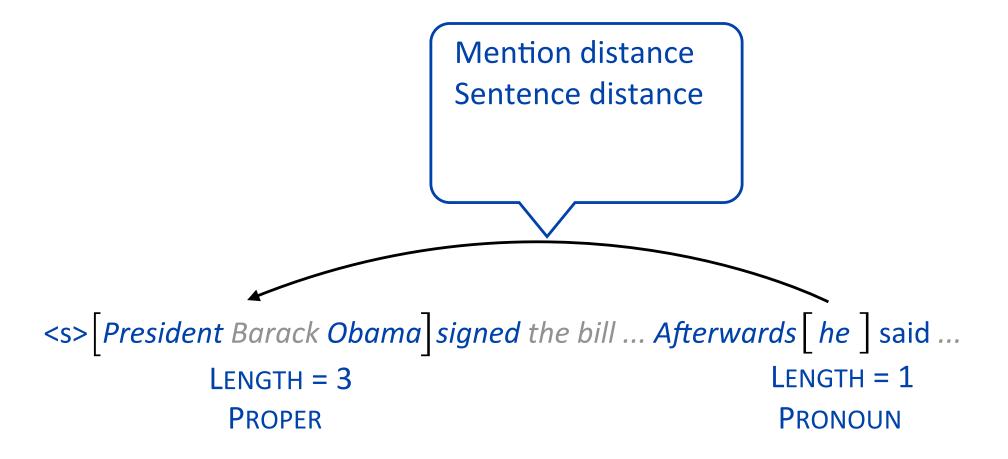














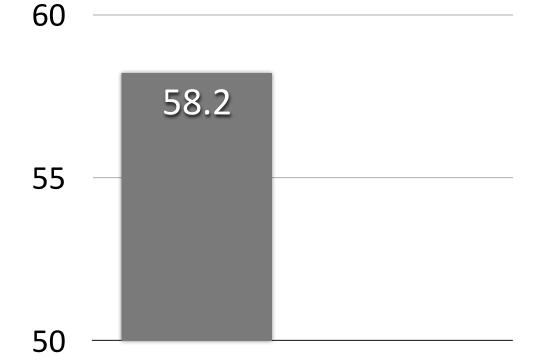
```
Mention distance
                               Sentence distance
                               Head match
                               Exact match
<s> | President Barack Obama | signed the bill ... Afterwards | he | said ...
                                                       LENGTH = 1
            LENGTH = 3
              PROPER
                                                       PRONOUN
```



SURFACE Accuracy

65





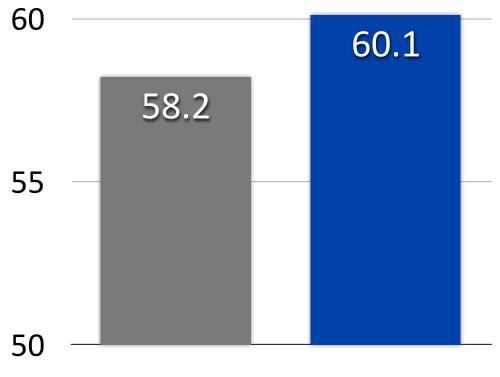


SURFACE Accuracy



So Far

SURFACE







Anaphoric pronouns

<u>Obama</u> ← <u>he</u>



Anaphoric pronouns

<u>Obama</u> ← <u>he</u>

72.0%



Anaphoric pronouns

Obama ← he

72.0%

Referring: head match

the U.S. <u>president</u> ← <u>president</u>



Anaphoric pronouns

Obama ← he

Referring: head match

the U.S. <u>president</u> ← <u>president</u>

72.0%

82.7%



Anaphoric pronouns

Obama ← he

72.0%

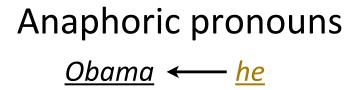
Referring: head match the U.S. <u>president</u> ← <u>president</u>

82.7%

Referring: no head match

David <u>Cameron</u> ← prime <u>minister</u>

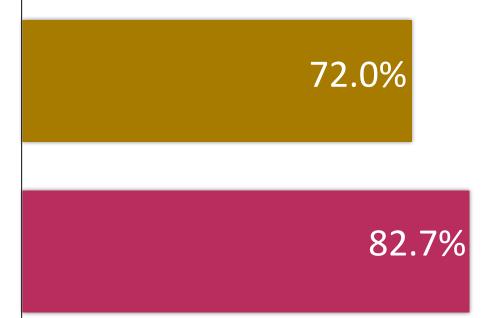




Referring: head match the U.S. <u>president</u> ← <u>president</u>

Referring: no head match

David <u>Cameron</u> ← prime <u>minister</u>







[David Cameron]₁ ... [The prime minister]₁ ...



[David Cameron]₁ ... [The prime minister]₁ ...



[David Cameron]₁ ... [The prime minister]₁ ...

Number, gender



[David Cameron]₁ ... [The prime minister]₁ ...

- Number, gender
- Named entity type



[David Cameron]₁ ... [The prime minister]₁ ...

- Number, gender
- Named entity type
- Unsupervised clustering labels



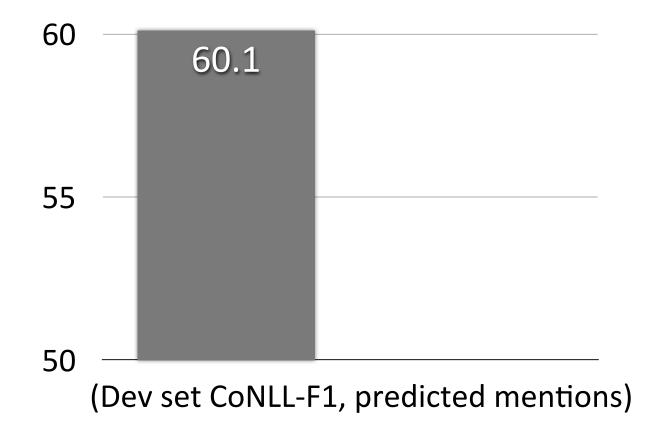
[David Cameron]₁ ... [The prime minister]₁ ...

- Number, gender
- Named entity type
- Unsupervised clustering labels
- WordNet hypernymy / synonymy



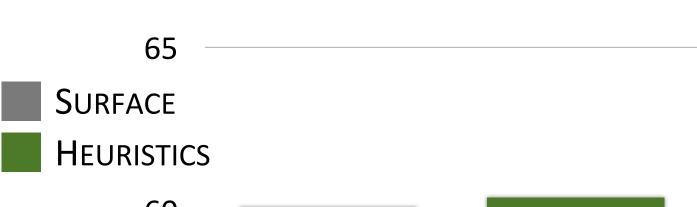
SURFACE Accuracy

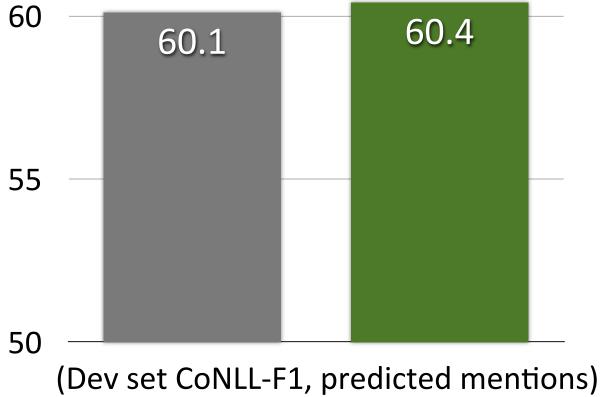
SURFACE





SURFACE Accuracy







What works?



What works?

Importing external information with sophisticated heuristics

Ponzetto and Strube (2006) Rahman and Ng (2011) Bansal and Klein (2012)



What works?

Importing external information with sophisticated heuristics

Ponzetto and Strube (2006) Rahman and Ng (2011) Bansal and Klein (2012)

We can support additional heuristic features, including number and gender information (Bergsma and Lin, 2006)



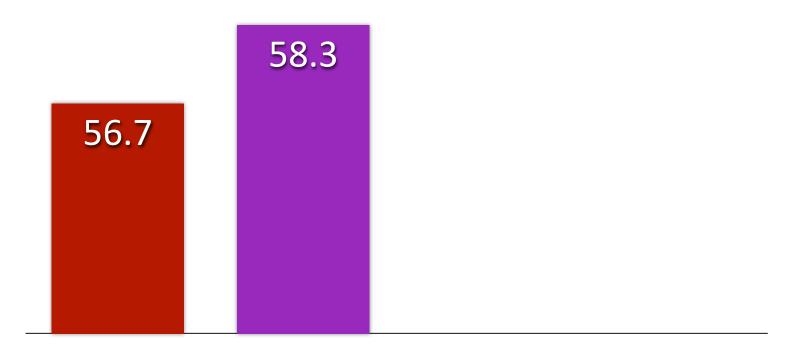


STANFORD (Lee et al., 2011)



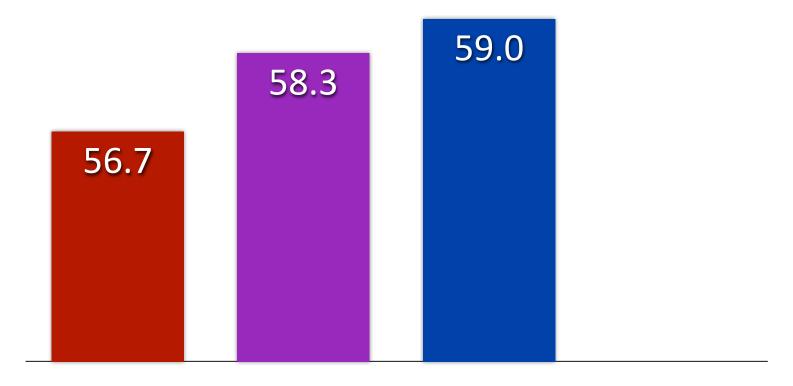


- STANFORD (Lee et al., 2011)
- IMS (Björkelund and Farkas, 2012)

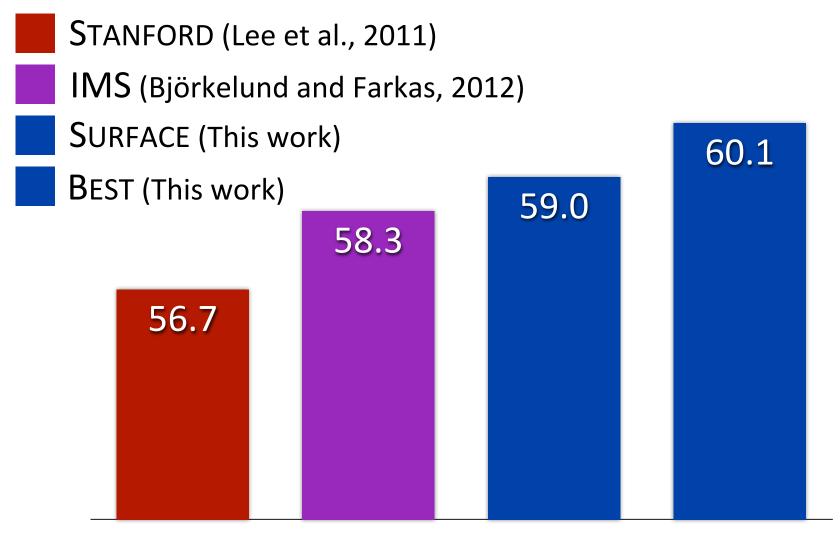




- STANFORD (Lee et al., 2011)
- IMS (Björkelund and Farkas, 2012)
- SURFACE (This work)









Conclusion



Conclusion

Surface lexical features capture a wide range of linguistic phenomena in a unified way



Conclusion

- Surface lexical features capture a wide range of linguistic phenomena in a unified way
- Semantic errors require heavy-duty information from other knowledge sources



Conclusion

- Surface lexical features capture a wide range of linguistic phenomena in a unified way
- Semantic errors require heavy-duty information from other knowledge sources
- Extensible system that achieves state-of-the-art performance



Conclusion

- The Berkeley Coreference Resolution System: http://nlp.cs.berkeley.edu
 - Full end-to-end system (accepts raw text as input)



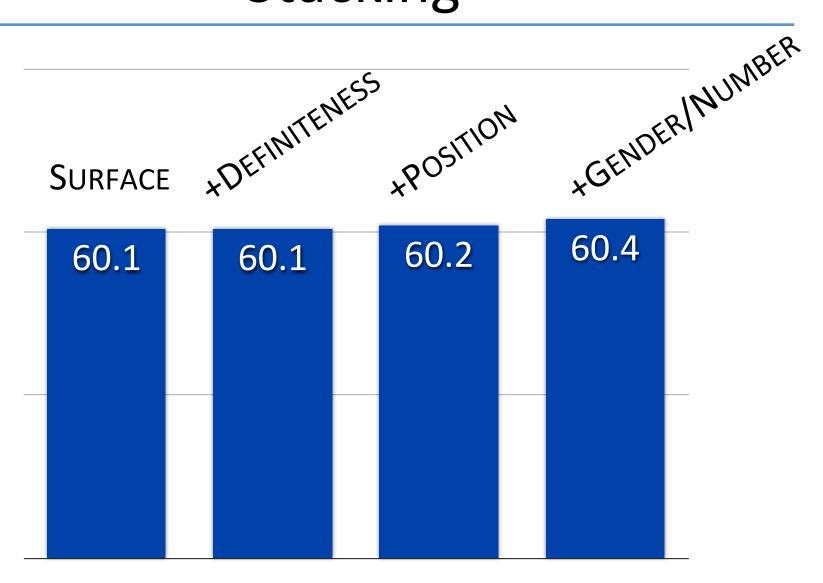
Conclusion

- The Berkeley Coreference Resolution System: http://nlp.cs.berkeley.edu
 - Full end-to-end system (accepts raw text as input)

Thank you!



Stacking







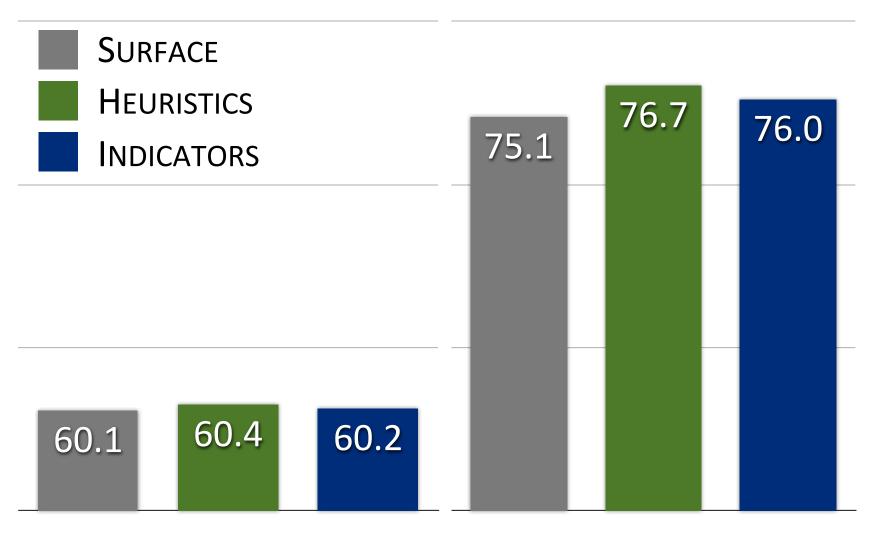






Predicted mentions

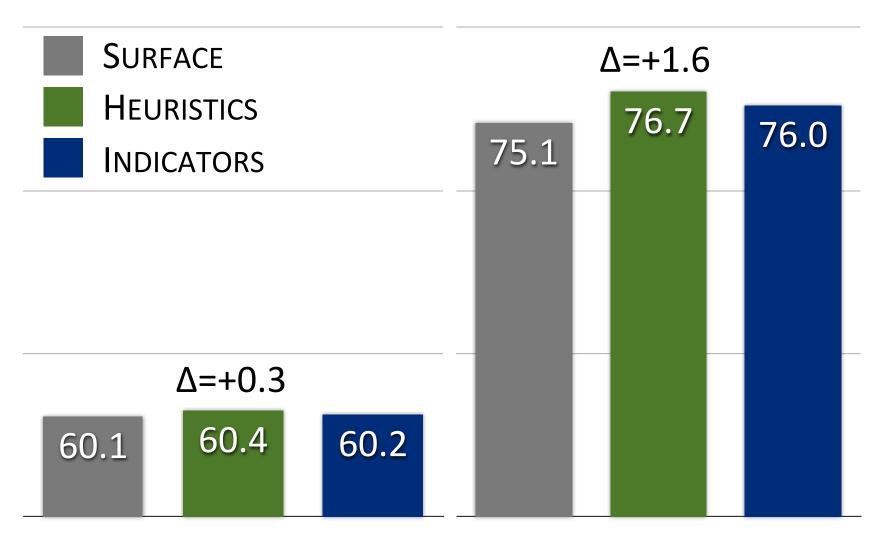




Predicted mentions

Gold mentions

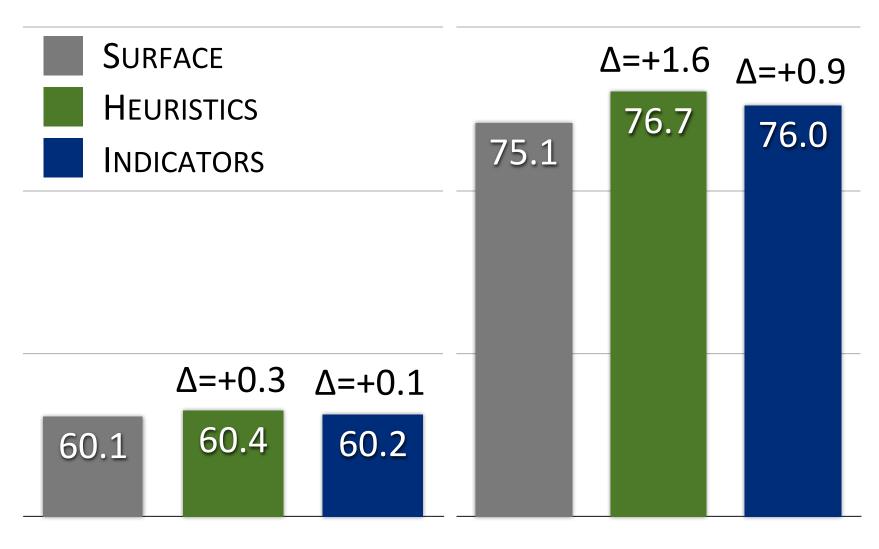




Predicted mentions

Gold mentions





Predicted mentions

Gold mentions

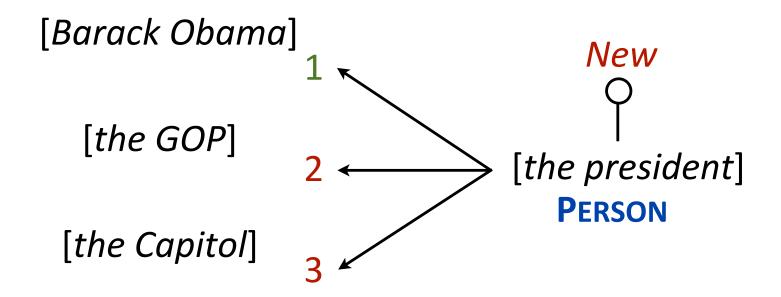


New
O
[the president]

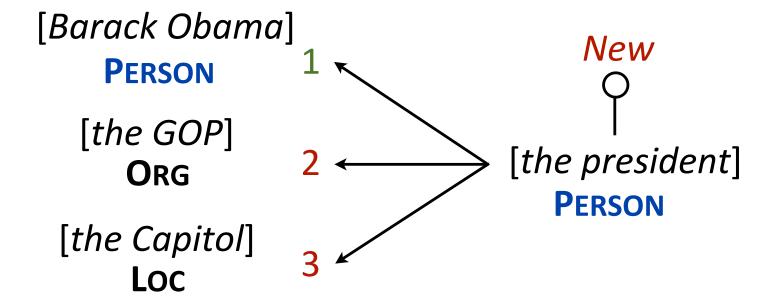




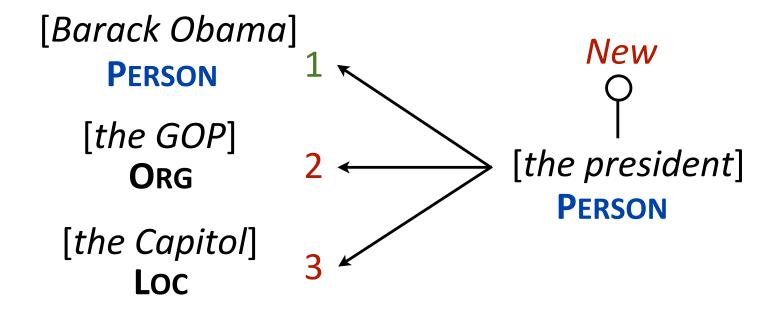












3x more mentions when singletons are included



