

# Prominence and Coherence in a Bayesian Theory of Pronoun Interpretation

Andrew Kehler  
UC San Diego

(Joint work with Hannah Rohde, University of Edinburgh)

# Pronoun Interpretation and Production

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- ❖ Common wisdom: There is a unified notion of prominence that determines when...
  - ❖ a speaker will produce a pronoun to mention a referent, and hence
  - ❖ a comprehender will successfully interpret the reference
- ❖ The task is to identify what factors affect prominence (grammatical role, parallelism, thematic role, information structural, semantic, etc)
- ❖ I will try to disabuse you of this, and argue instead for a different model

# Implicit Causality

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- ❖ Previous work has shown that so-called *implicit causality* verbs are associated with strong pronoun biases (Garvey and Caramazza, 1974 and many others)

*Amanda amazes Brittany because she \_\_\_\_\_ [subject-biased]*

*Amanda detests Brittany because she \_\_\_\_\_ [object-biased]*

- ❖ The connective *because* indicates an Explanation coherence relation: the second sentence describes a cause or reason for the eventuality described by the first
- ❖ For free prompts, IC verbs result in a greater number of Explanation continuations (60%) than non-IC controls (24%) (Kehler et al. 2008)

# Background

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- \* A study by Stevenson et al (1994) compared pronoun prompts with free prompts:

*Amanda detests Brittany. She \_\_\_\_\_*

*Amanda detests Brittany. \_\_\_\_\_*

- \* Two results:
  - \* Interpretation: Greater number of subject references in the pronoun-prompt condition than the free-prompt condition
  - \* Production: In the free-prompt condition, a strong tendency to use a pronoun to refer to the subject and a name to refer to a non-subject

# Bayesian Interpretation (Kehler et al. 2008)

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- \* Bayesian formulation:

$$P(\text{referent} \mid \text{pronoun}) = \frac{P(\text{pronoun} \mid \text{referent}) P(\text{referent})}{\sum_{\text{referent} \in \text{referents}} P(\text{pronoun} \mid \text{referent}) P(\text{referent})}$$

Diagram illustrating the Bayesian formulation:

- Interpretation (left) points down to the numerator of the equation.
- Production (Subject Bias) (top left) points down to the first term in the numerator.
- Prior Expectation (Semantics/Coherence) (top right) points down to the second term in the numerator.

- \* Data is consistent with a scenario in which semantics/coherence-driven biases primarily affect probability of *next-mention*, whereas grammatical biases affect *choice of referential form*
- \* Results in the counterintuitive prediction that production biases are insensitive to a set of factors that affect the ultimate interpretation bias

# Implicit Causality (Ambiguous Contexts)

(Rohde, 2008; Fukumura & van Gompel 2010; Rohde & Kehler 2014)

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- Measure next mention bias  $P(\text{referent})$   
and production bias  $P(\text{pronoun} \mid \text{referent})$
- \* Free prompts:
    - \* *Amanda amazed Brittany.* \_\_\_\_\_ [IC, subject-biased]
    - \* *Amanda detested Brittany.* \_\_\_\_\_ [IC, object-biased]
    - \* *Amanda chatted with Brittany.* \_\_\_\_\_ [non-IC]
  
  - \* Pronoun prompts:
    - \* *Amanda amazed Brittany. She* \_\_\_\_\_ [IC, subject-biased]
    - \* *Amanda detested Brittany. She* \_\_\_\_\_ [IC, object-biased]
    - \* *Amanda chatted with Brittany. She* \_\_\_\_\_ [non-IC]
- Measure interpretation bias  
 $P(\text{referent} \mid \text{pronoun})$

# Implicit Causality (Ambiguous Contexts)

(Rohde, 2008; Fukumura & van Gompel 2010; Rohde & Kehler 2014)

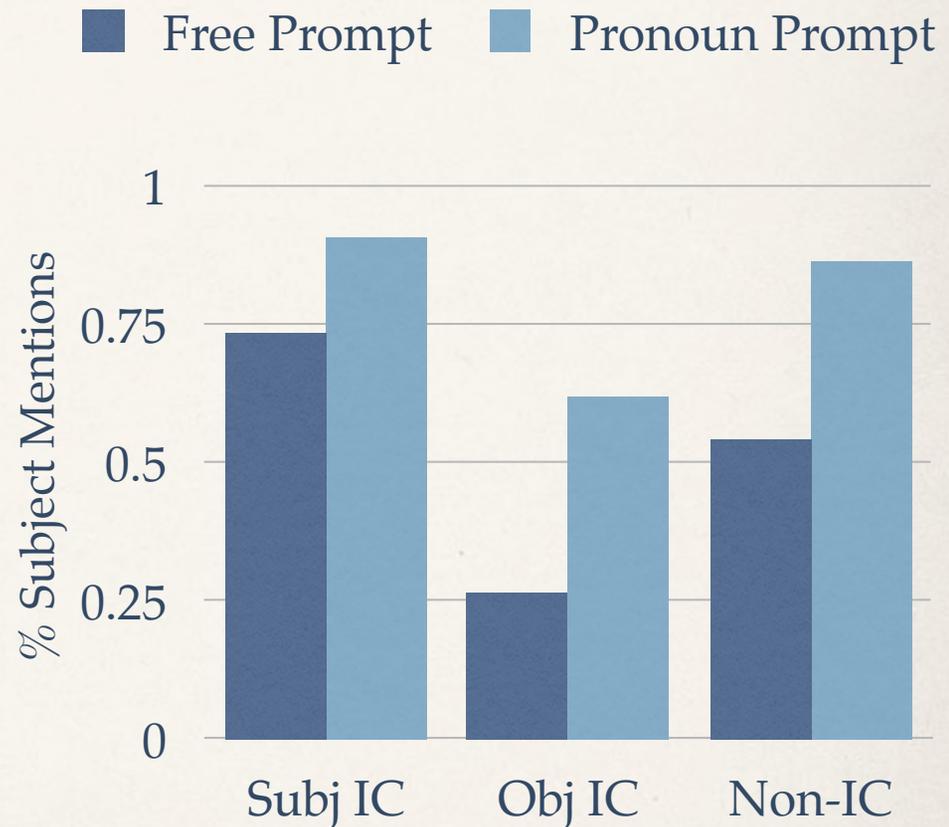
- \* Rohde (2008), Rohde & Kehler (2014): IC affects interpretation

- \* *Amanda amazed Brittany.*  
(She) \_\_\_\_\_ [IC, subject-biased]

- \* *Amanda detested Brittany.*  
(She) \_\_\_\_\_ [IC, object-biased]

- \* *Amanda chatted with Brittany.*  
(She) \_\_\_\_\_ [non-IC]

- \* Result: IC bias affects next-mention (prior) and pronoun interpretation



# Production Biases (Ambiguous Contexts)

(Rohde, 2008; Fukumura & van Gompel 2010; Rohde & Kehler 2014)

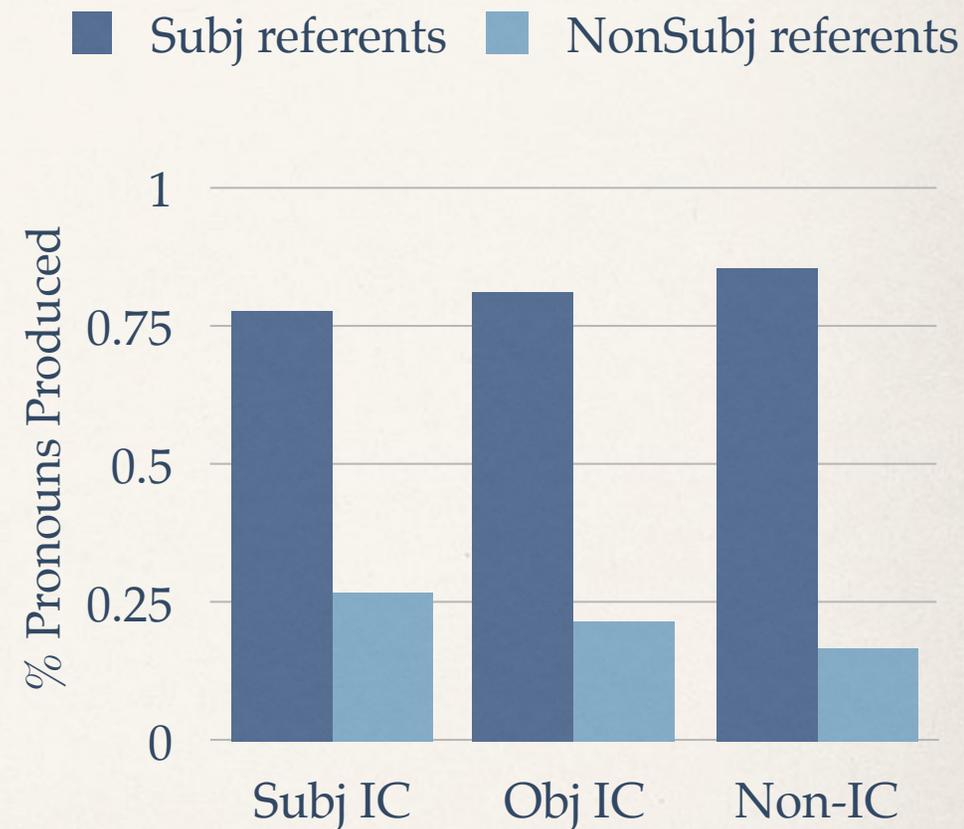
- \* Rohde (2008), Rohde & Kehler (2014): IC doesn't affect production

- \* *John amazed Mary.* \_\_\_\_\_  
[IC, subject-biased]

- \* *John detested Mary.* \_\_\_\_\_  
[IC, object-biased]

- \* *John chatted with Mary.* \_\_\_\_\_  
[non-IC]

- \* Result: grammatical role matters, but semantic bias does not



# Testing the Theory: Inferred Causes

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## \* Passage completion study:

*The boss fired the employee who was hired in 2002. He \_\_\_\_\_ [Control]*

*The boss fired the employee who was embezzling money. He \_\_\_\_\_ [ExplRC]*

*The boss fired the employee who was hired in 2002. \_\_\_\_\_ [Control]*

*The boss fired the employee who was embezzling money. \_\_\_\_\_ [ExplRC]*

## \* Analyze:

- \* Coherence relations (Explanation or Other)
- \* Next-mentioned referent (Subject or Object)
- \* Form of Reference (free-prompt condition; Pronoun or Other)

# Predictions

RC Type

[ExplRC] *The boss fired the employee who was embezzling money.*  
[Control] *The boss fired the employee who was hired in 2002.*

Coherence Relations

ExplRC: fewer Explanations

Production Bias  
 $P(\textit{pronoun} \mid \textit{referent})$

Subjects: more pronouns  
ExplRC: no effect

Next-Mention Biases  
 $P(\textit{referent})$

ExplRC: fewer object next-mentions  
(i.e., more subject references)

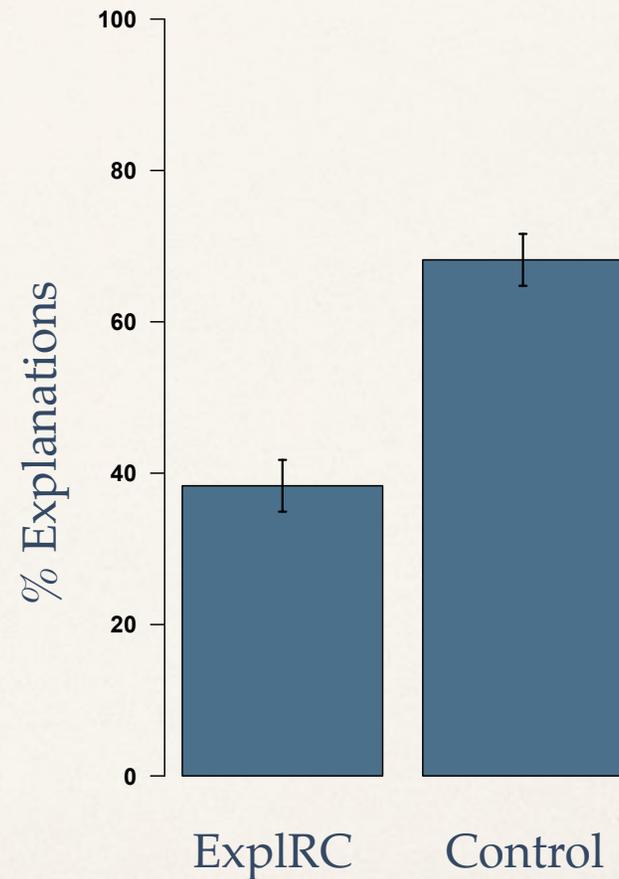
Interpretation Bias  
 $P(\textit{referent} \mid \textit{pronoun})$

ExplRC: fewer object refs (= more subjects)  
Pronoun prompt: more subject references

# Prediction 1: Coherence Relations

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- \* Predict a smaller percentage of Explanation relations in the ExplRC condition than the Control condition
- \* Confirmed: ( $\beta=2.06$ ;  $p<.001$ )



*[ExplRC] The boss fired the employee who was embezzling money.*

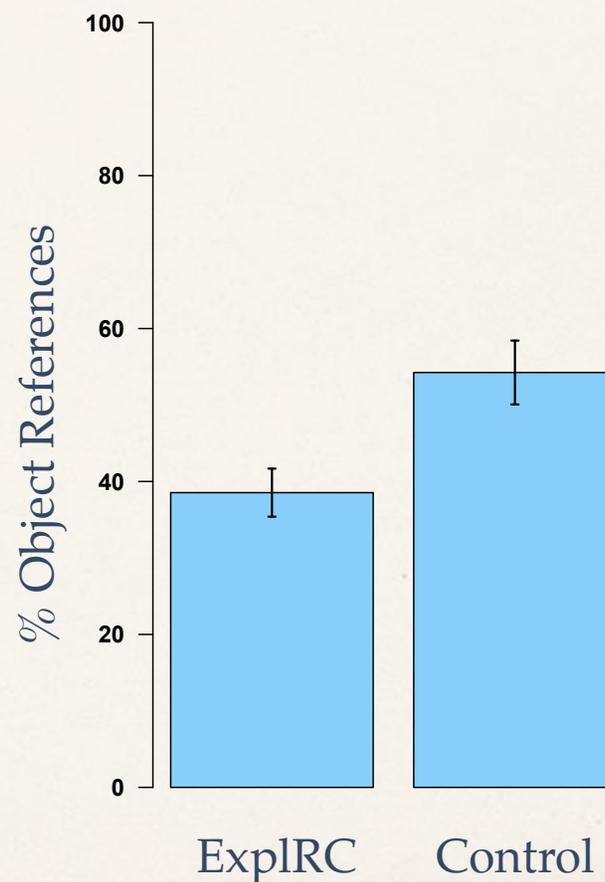
*[Control] The boss fired the employee who was hired in 2002.*

# Prediction 2: Next-Mention Biases

$$P(\text{referent} \mid \text{pronoun}) \propto P(\text{pronoun} \mid \text{referent}) P(\text{referent})$$

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- \* For free-prompt condition, predict a smaller percentage of next mentions of the object in ExplRC condition than the Control condition
- \* Confirmed: ( $\beta=.720$ ;  $p<.05$ )



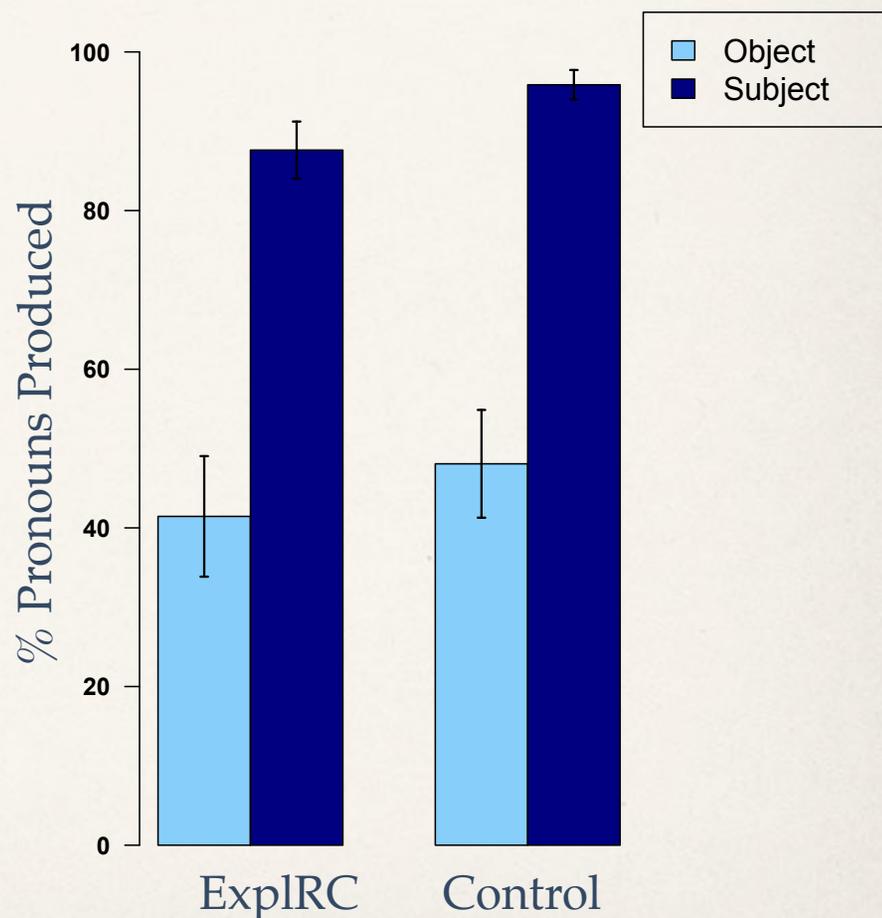
[ExplRC] The boss fired the employee *who was embezzling money*.

[Control] The boss fired the employee *who was hired in 2002*.

# Prediction 3: Rate of Pronominalization

$$P(\text{referent} \mid \text{pronoun}) \propto P(\text{pronoun} \mid \text{referent}) P(\text{referent})$$

- \* Predict an effect of grammatical role on pronominalization rate (favoring subjects; free prompt condition)
  - \* Confirmed: ( $\beta=4.11$ ;  $p<.001$ )
- \* But no interaction with RC condition
  - \* Confirmed ( $\beta=0.12$ ;  $p=.92$ )
  - \* Marginal effect of RC condition ( $\beta=0.94$ ;  $p=.078$ )



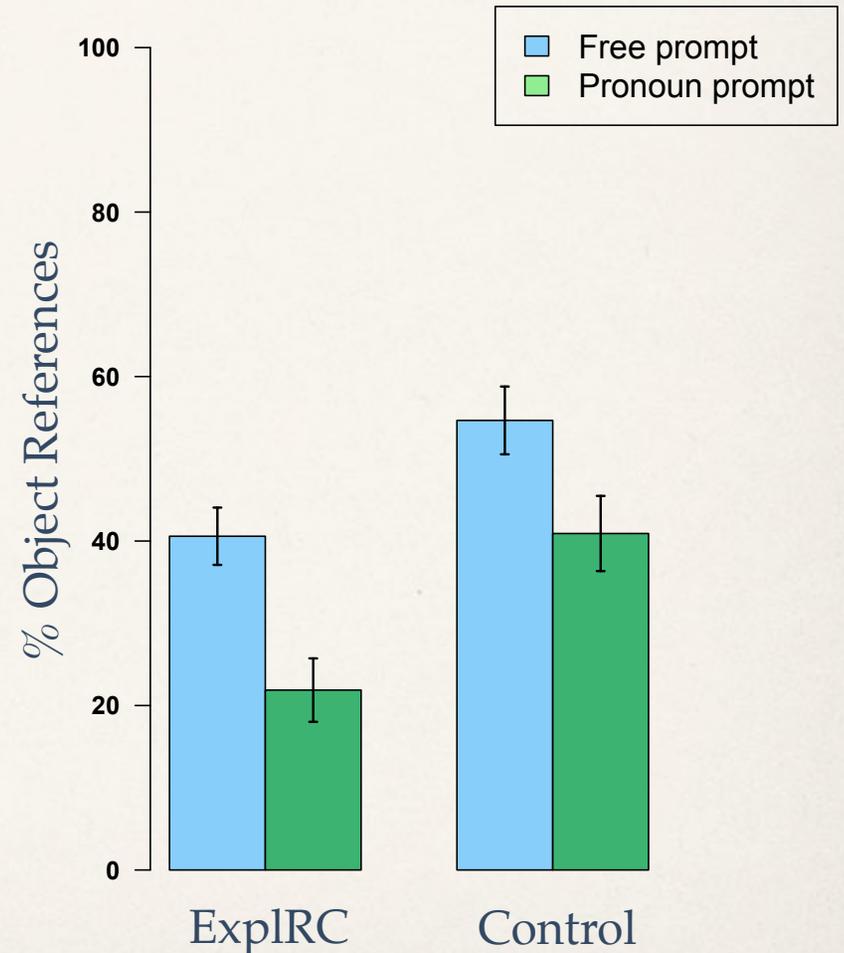
[ExplRC] *The boss fired the employee who was embezzling money.*

[Control] *The boss fired the employee who was hired in 2002.*

# Predictions 4 & 5: Pronoun Interpretation

$$P(\textit{referent} \mid \textit{pronoun}) \propto P(\textit{pronoun} \mid \textit{referent}) P(\textit{referent})$$

- \* Predict a smaller percentage of object mentions in the ExplRC condition than the Control condition...
  - \* Confirmed: ( $\beta=1.17$ ;  $p<.005$ )
- \* ...and in the free-prompt condition than the pronoun-prompt condition
  - \* Confirmed ( $\beta=-1.27$ ;  $p=.001$ )
- \* Marginal interaction ( $\beta=0.85$ ;  $p=.078$ )
- \* Effect in Pronoun subset only ( $\beta=1.46$ ;  $p<.005$ )



[ExplRC] *The boss fired the employee who was embezzling money.*

[Control] *The boss fired the employee who was hired in 2002.*

# Model Comparison

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- ❖ We can evaluate the predictions of the model by estimating the likelihood and prior from the data in the free prompt condition to generate a *predicted* pronoun interpretation bias
- ❖ We then compare that to the *actual* pronoun interpretation bias estimated from the data in the pronoun-prompt condition

$$P(\text{referent} \mid \text{pronoun}) = \frac{P(\text{pronoun} \mid \text{referent}) P(\text{referent})}{\sum_{\text{referent} \in \text{referents}} P(\text{pronoun} \mid \text{referent}) P(\text{referent})}$$

# Competing Model: Mirror Model

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- ❖ Recall the common wisdom: the factors that comprehenders use to interpret pronouns are those that speakers use when choosing to use one.
- ❖ That means the interpreter's biases will be proportional to (their estimates of) the speaker's production biases

$$P(\text{referent} \mid \text{pronoun}) = \frac{P(\text{pronoun} \mid \text{referent}) P(\text{referent})}{\sum_{\text{referent} \in \text{referents}} P(\text{pronoun} \mid \text{referent}) P(\text{referent})}$$

# Competing Model: Expectancy Model

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- \* According to Arnold's Expectancy Hypothesis (1998, 2001, inter alia), comprehenders will interpret a pronoun to refer to whatever referent they expect to be mentioned next

$$P(\text{referent} \mid \text{pronoun}) = \frac{P(\text{pronoun} \mid \text{referent}) P(\text{referent})}{\sum_{\text{referent} \in \text{referents}} P(\text{pronoun} \mid \text{referent}) P(\text{referent})}$$

# Model Comparison: Results

- ❖ Comparison of actual rates of pronominal reference to object (pronoun-prompt condition) to the predicted rates for three competing models (using estimates from free-prompt condition)

	Actual	Bayesian	Mirror	Expectancy
ExplRC	0.215	0.229	0.321	0.385
Control	0.410	0.373	0.334	0.542

$R^2=.48 / .49$

$R^2=.34 / .42$

$R^2=.14 / .12$

# Conclusions

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- ❖ Pronoun interpretation is sensitive to coherence factors, in this case the invited inference of an explanation
- ❖ Pronoun production, however, is not
- ❖ The data thus demonstrate precisely the asymmetry predicted by the Bayesian analysis
- ❖ A corollary is that there is no unified notion of prominence that drives interpretation and production
- ❖ Indeed, perhaps the best *independent* measure of prominence is provided by next-mention expectations, but pronoun biases are not the same

Thank you!

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