

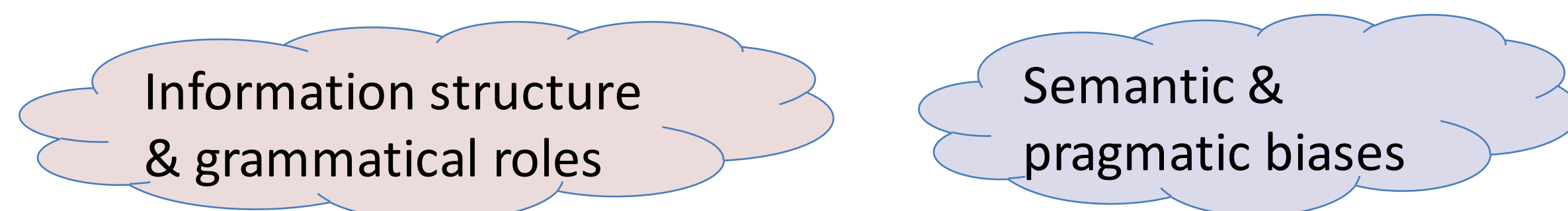
Introduction

How do pronoun interpretation and production relate to each other?

- **Bayesian pronoun interpretation model** (Kehler & Rohde, 2013; Rohde & Kehler, 2014)

$$P(\text{referent}|\text{pronoun}) = \frac{P(\text{pronoun}|\text{referent})P(\text{referent})}{\sum_{\text{referent} \in \text{refs}} P(\text{pronoun}|\text{referent})P(\text{referent})}$$

Posterior: Pronoun interpretation biases \propto Likelihood: Pronoun production biases \times Prior: Next-mention biases



- **Expectancy (prior-only) model** (Arnold, 2001)

$$P(\text{referent}|\text{pronoun}) \leftarrow P(\text{referent})$$

- **Mirror (likelihood-only) model**

$$P(\text{referent}|\text{pronoun}) \leftarrow \frac{P(\text{pronoun}|\text{referent})}{\sum_{\text{referent} \in \text{refs}} P(\text{pronoun}|\text{referent})}$$

Research questions

Does the Bayesian pronoun interpretation theory:

- generalize crosslinguistically, specifically to Mandarin Chinese? -> Expt. 1
- generalize across different syntactic constructions, specifically active/passive alternations? -> Expt. 2

Methods and Materials

Active voice (Experiments 1 & 2)

- IC-1, Free prompt** Meihui dadong-le Jieyi. _____
“[Meihui]_{NP1} impressed [Jieyi]_{NP2}. _____”
- IC-1, Pronoun prompt** Meihui dadong-le Jieyi. Ta _____
“[Meihui]_{NP1} impressed [Jieyi]_{NP2}. She _____”
- IC-2, Free prompt** Meihui jiegu-le Jieyi. _____
“[Meihui]_{NP1} fired [Jieyi]_{NP2}. _____”
- IC-2, Pronoun prompt** Meihui jiegu-le Jieyi. Ta _____
“[Meihui]_{NP1} fired [Jieyi]_{NP2}. She _____”

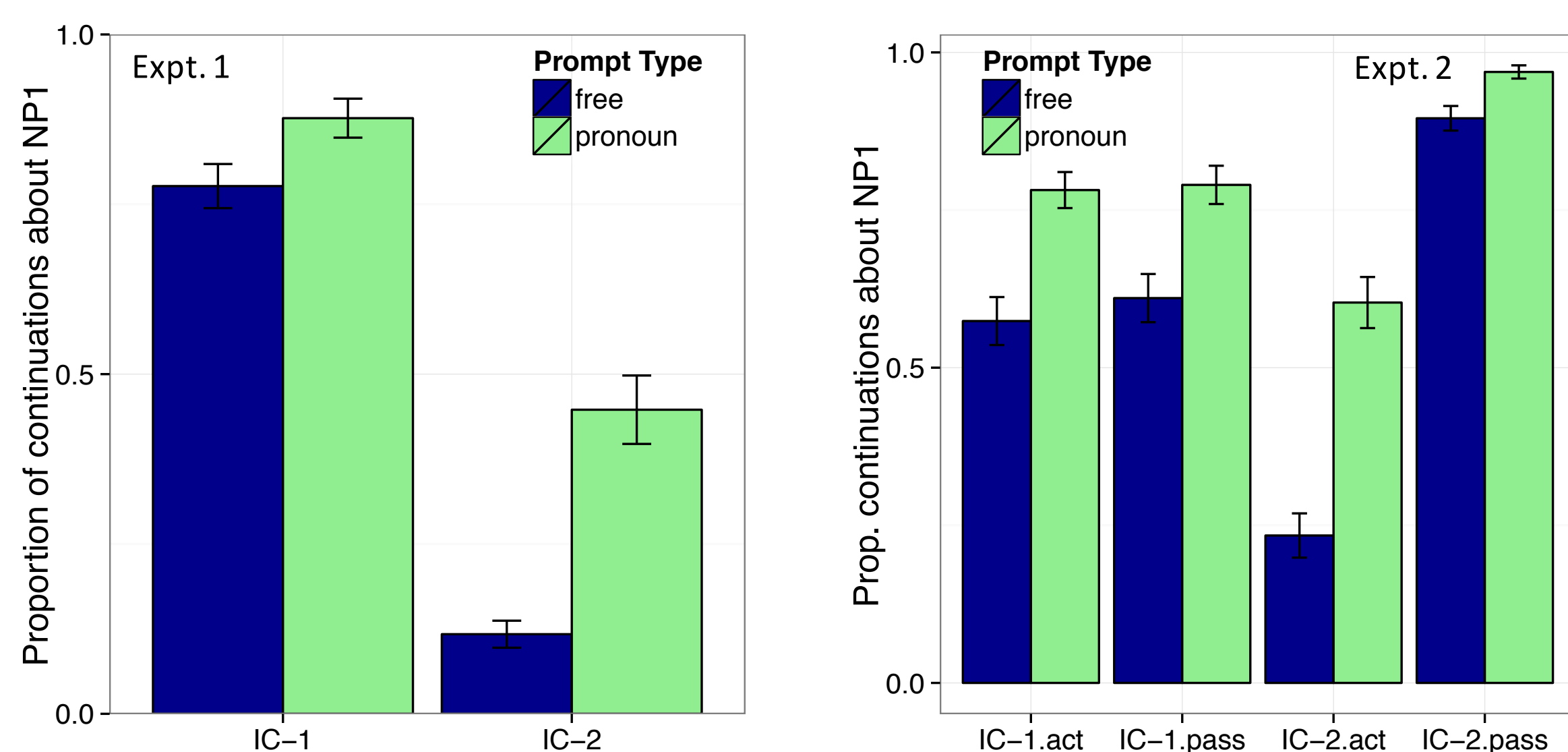
Passive voice (Experiment 2)

- IC-1, Free prompt** Jieyi bei Meihui dadong-le. _____
“[Jieyi]_{NP1} was_by [Meihui]_{NP2} impressed. _____”
- IC-1, Pronoun prompt** Jieyi bei Meihui dadong-le. Ta _____
“[Jieyi]_{NP1} was_by [Meihui]_{NP2} impressed. She _____”
- IC-2, Free prompt** Jieyi bei Meihui jiegu-le. _____
“[Jieyi]_{NP1} was_by [Meihui]_{NP2} fired. _____”
- IC-2, Pronoun prompt** Jieyi bei Meihui jiegu-le. Ta _____
“[Jieyi]_{NP1} was_by [Meihui]_{NP2} fired. She _____”

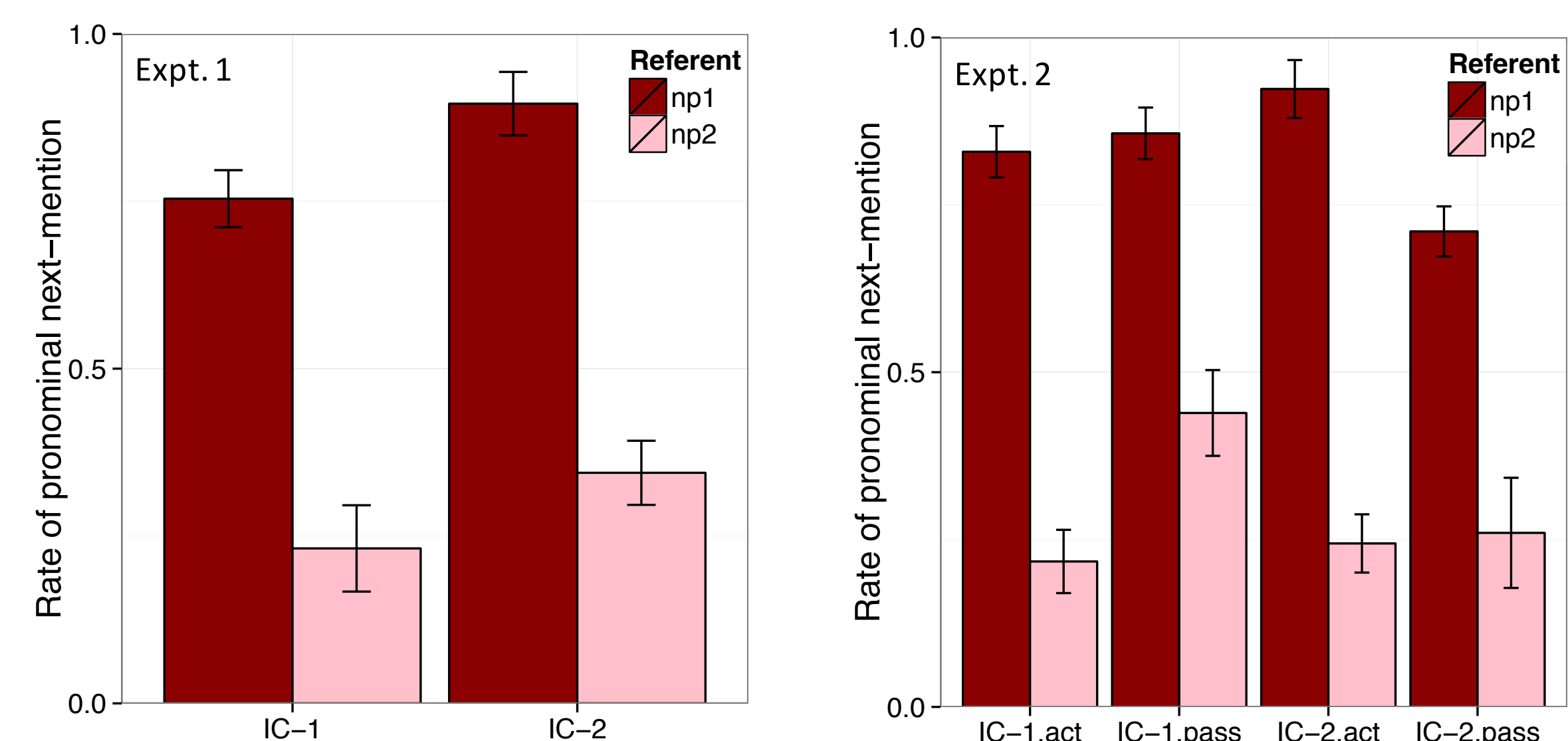
Implicit causality (IC) verbs: IC-1 and IC-2 favors NP1 and NP2 respectively.
16 IC-1 and 20 IC-2 verbs were used in each experiment.

Results

Prior $P(\text{NP1})$ vs. **Posterior** $P(\text{NP1}|\text{pronoun})$: Who was the intended referent?
Significant difference between Free prompt and Pronoun prompt

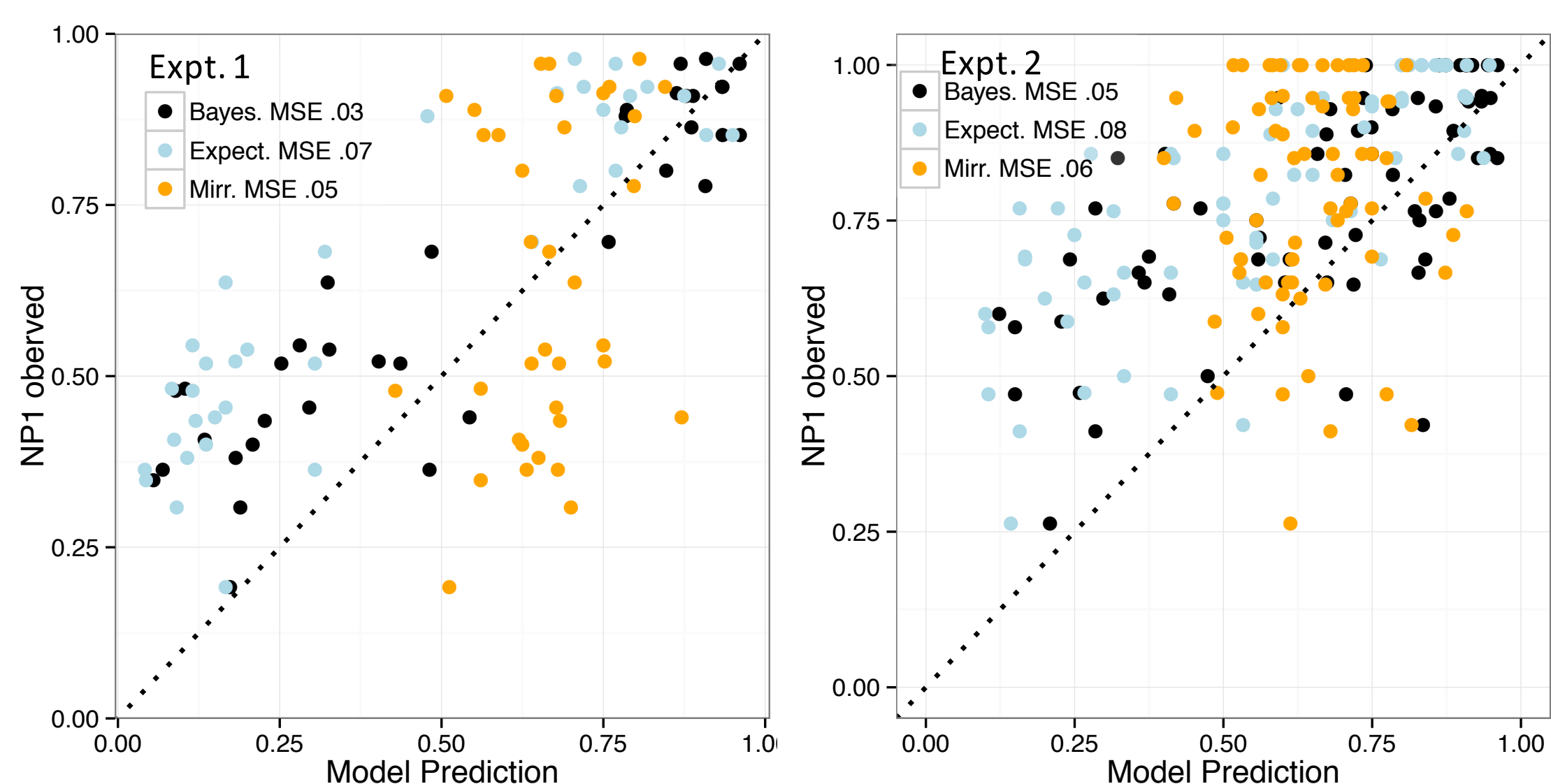


Likelihood $P(\text{pronoun}|\text{NP1})$ vs. $P(\text{pronoun}|\text{NP2})$: Pronoun used given a referent?
Significant difference between NP1 and NP2, not between passivized NP1 and active NP1



Quantitative model comparisons (Bayesian vs. Expectancy vs. Mirror):

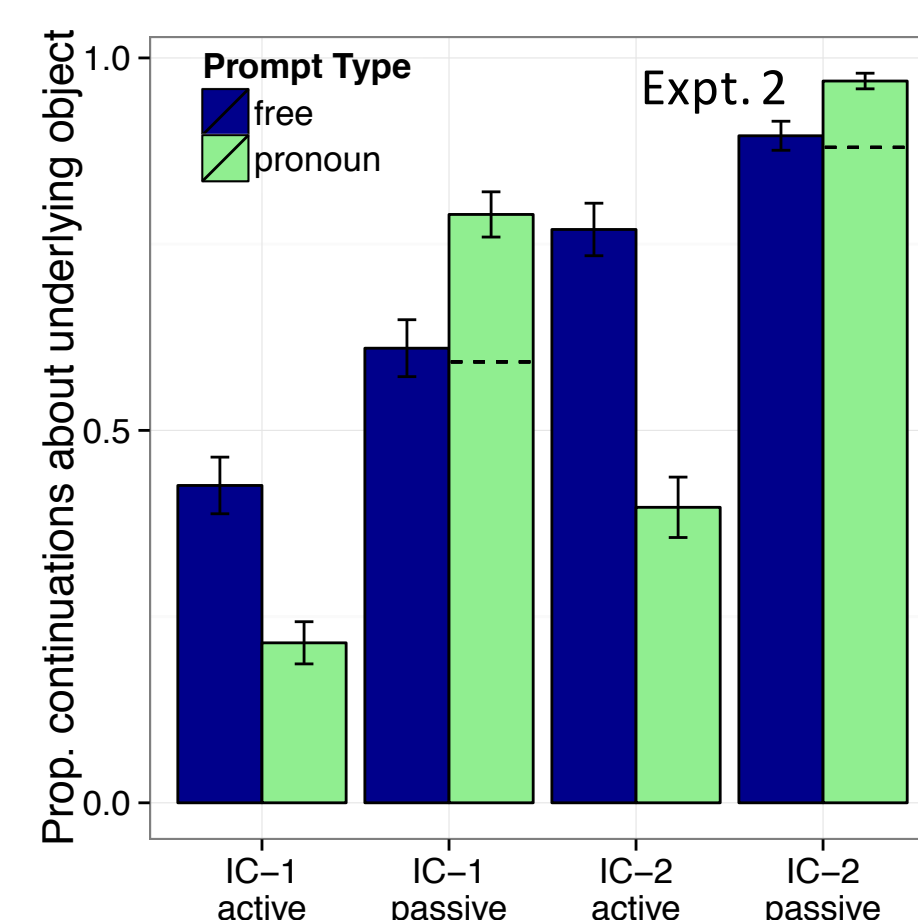
Both prior and likelihood essential? Yes, Bayesian model is a better fit than either of the two competing models



- Observed NP1 interpretation rate against item-specific prediction of three models
- $x = y$ dotted line would be perfect model fit
- **Bayesian model**: least mean squared error
- **Mirror model**: underpredicts cross-item/condition variability in interpretation preference
- **Expectancy model**: underpredicts the rate of NP1 interpretation

We also found:

- Passivization increased the next-mentions of the logical object ($P(\text{Jieyi})$)



- This increase is reflected in pronoun interpretation
- Dashed lines indicate next-mention rates predicted using the active condition prior

General Discussion

- Verb semantics affect prior next mention biases
- Grammatical roles affect pronoun production biases
- Disassociation between next-mention biases and pronoun production biases
- No consistent evidence showing passivized subject is more likely to be pronominalized
- Full Bayesian model best predicted human data in pronoun interpretation
- Passivization increases next-mention biases of the logical object
- Results support a crosslinguistically general Bayesian theory of pronoun interpretation

