How children are and aren't like adults when it comes to interpreting pronouns: A developmental modeling investigation

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The girls wave at the teacher...

???

...and then **she** leaves.





The girls wave at the teacher...

277

...and then **she** leaves.

sg

Agreement mismatch: "she" is singular but "girls" is plural





The girls wave at the teacher...

...and then **she** leaves.

Agreement match: both "she" and "teacher" are singular





The girls wave at the teacher...

...and then **she** leaves.

How to choose?

Use linguistic knowledge of agreement.





The girl waves at the teacher...

???? ...and then **she** leaves.



This could work: both "she" and "girl" are singular.





The girl waves at the teacher...

...and then **she** leaves.

But so could this: both "she" and "teacher" are singular.





The girl waves at the teacher...

...and then **she** leaves.

How to choose?



The girl waves at the teacher...

???

...and then **she** leaves.

sg

How to choose?

Use contextual knowledge (who's likely to be leaving)

Maybe the girl is getting ready to leave the classroom.



How to choose?

Use linguistic knowledge about connectives.

Maybe pronouns after "and then" tend to refer to the previous subject.





The girls wave at the teacher...

Subject pl sg 222

...and then **she** leaves.

What about when interpretation cues conflict?





The girls wave at the teacher...

...and then **she** leaves.

What about when interpretation cues conflict?

Here, the connective "and then" signals the subject "the girls", while the agreement signals the object "the teacher".





The girls wave at the teacher...

...and then **she** leaves.

SQ



What about when interpretation cues conflict?

Here, English-speaking adults let agreement matter more than the connective.

So, they interpret "she" as "the teacher".





...and then **she** leaves.





Something English-speaking adults have learned: How to resolve interpretation cue conflicts in context.





The girls wave at the teacher...

...and then **she** leaves.



The need to integrate multiple cues to interpretation doesn't just happen in English, of course.





```
Las niñas saludan a la maestra...

The girls wave at the teacher...

pl sg

????

...y después ella sale.

... and then she leaves.

SO
```

Here's the same sentence in Spanish.





```
Las niñas saludan a la maestra...

The girls wave at the teacher...

pl sg
????

...y después Ø sale.

... and then PRONOUN leaves.

Sg
```

Spanish also allows the form of the pronoun to be null (this means the agreement information is on the verb).





```
Las niñas saludan a la maestra...

The girls wave at the teacher...

pl

sg

????

...y después Ø sale.

... and then PRONOUN leaves.

Sg
```

Just like English, there are multiple cues available to interpret the pronoun.





Las niñas saludan a la maestra...

The girls wave at the teacher...

pl sg

???? ...y después Ø sale.

... and then **PRONOUN** leaves.

sg



Spanish-speaking adults also have interpretation preferences.





Las niñas saludan a la maestra...

The girls

wave

at the teacher...

subject

???

...y después Ø

Ø

sale.

... and then

PRONOUN leaves.

Sg



For Spanish-speaking adults...

...the connective favors the subject.





Las niñas saludan a la maestra...

The girls

wave

at the teacher...

≈subject pl

= SQ

...y después Ø

sale.

... and then

PRONOUN leaves.

sg



For Spanish-speaking adults...

...the (singular) agreement (on the verb) indicates the singular object.





Las niñas saludan a la maestra...

The girls subject

wave at the teacher...

= sg

...y después Ø

sale.

... and then

PRONOUN leaves.

Sg



For Spanish-speaking adults...

...the (null) form favors the subject.





Las niñas saludan a la maestra...

The girls subject

wave

at the teacher...

= sc

...y después

sale.

... and then

PRONOUN leaves.

sg



For Spanish-speaking adults...

...this collection of cues generally causes the pronoun to be interpreted as the singular object (agreement matters the most).





Las niñas saludan a la maestra...

The girls wave at the teacher...

subject pl

????

...y después Ø sale.

... and then **PRONOUN** leaves.

sg



How do Spanish-learning children develop this ability to interpret pronouns in context?





Children's ability to interpret a pronoun in an adult-like way depends on (at least) two things.







First, children need adult-like knowledge of what each cue signals.



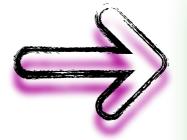


PRONOUN

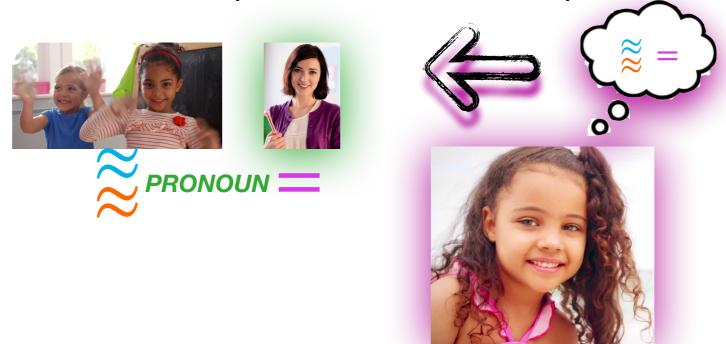


Second, children need adult-like ability to deploy that knowledge in real time.









When both of these are adult-like, we should get adult-like pronoun interpretation.





But if we get non-adult-like pronoun interpretation, then it could be due to immature knowledge, immature deployment of that knowledge, or both!

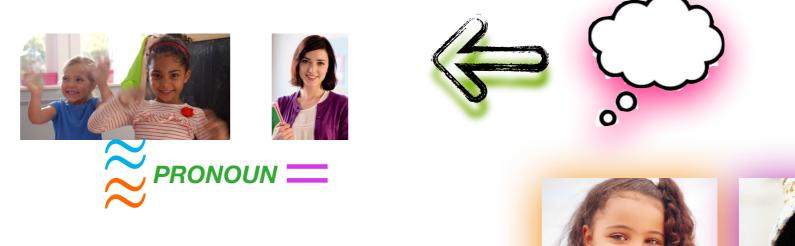




How do we tell what the differences are between child and adult pronoun interpretation? When we understand this better, we'll understand what children need to do to become adults.







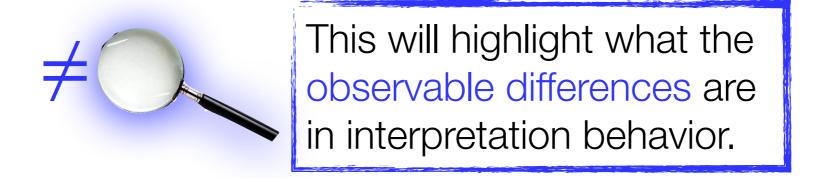
The plan, part 1: Get some empirical data on how children and adults interpret the same pronoun in a context where multiple cues are available.

Case study: Mexican Spanish



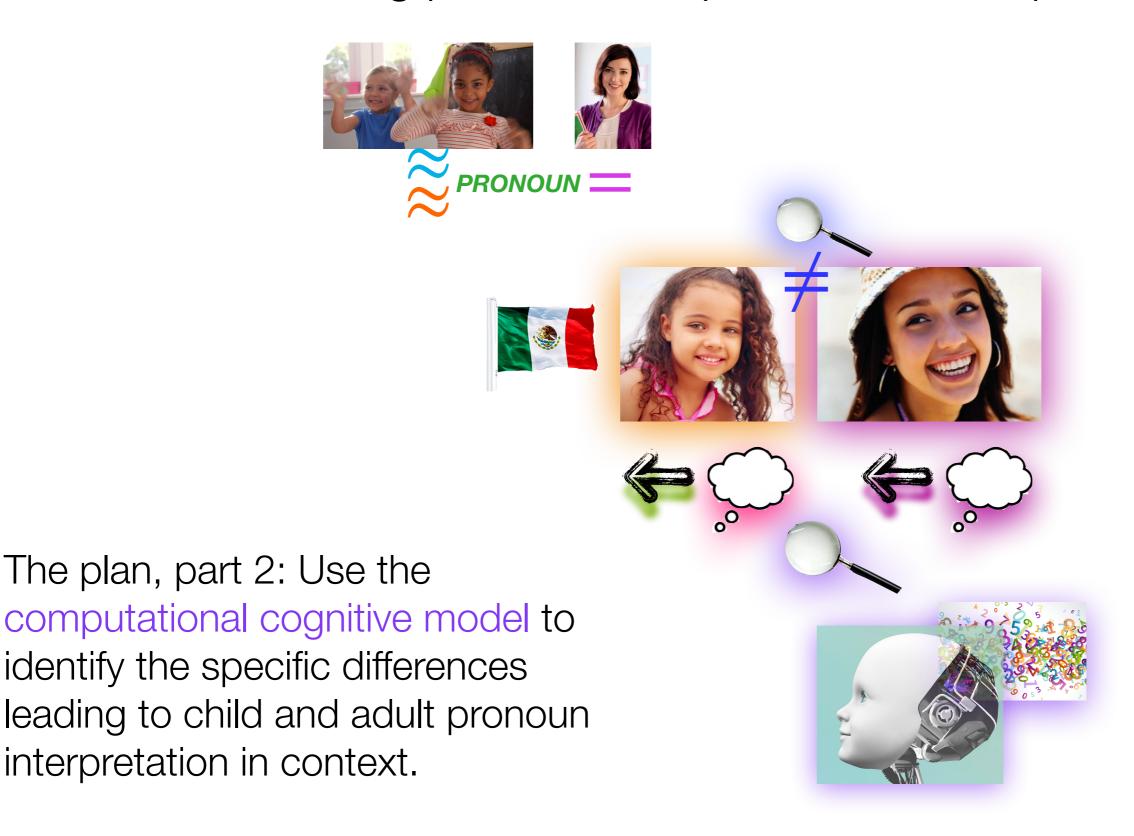


The plan, part 1: Get some empirical data on how children and adults interpret the same pronoun in a context where multiple cues are available.





The plan, part 2: Use computational cognitive modeling to formally articulate the potential process of pronoun interpretation in the context of these multiple cues.





Empirical data on pronoun interpretation









Las niñas saludan a la maestra... The girls at the teacher... wave

subject

...y después sale.

... and then **PRONOUN** leaves.

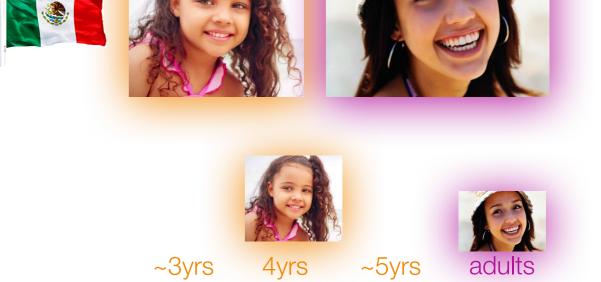
sg

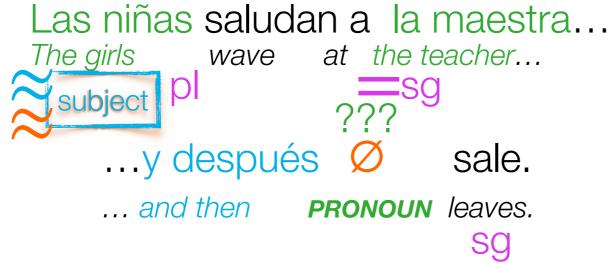
Children (~3, 4, and ~5) and adults are asked to interpret pronouns in the kind of contexts we saw before.

















subject pl ???

...y después Ø sale.

... and then **PRONOUN** leaves.



Rate of subject responses





Choice: Is the pronoun interpreted as the subject or the object?

We can plot the rate of subject responses.



Rate of subject responses





The girls





...y después Ø

wave

sale.

... and then **PRONOUN** leaves.



Context: Does agreement favor the subject or the object?

subject object











Las niñas saludan a la maestra... The girls at the teacher... wave ...y después Ø salen. ... and then **PRONOUN** leave.

Context: Does agreement favor the subject or the object?

subject object Favored by agreement







...y después

od sale.

... and then **PRONOUN** leaves.

sg



~5yrs

Context:

Rate of subject responses

~3yrs

Does agreement favor the subject or the object?

subject object

4yrs







Las niñas saludan a la maestra... The girls at the teacher... wave ...y después sale. ... and then leaves. **PRONOUN** sg

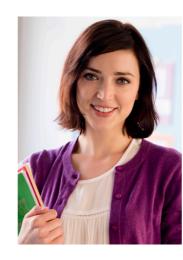


Context: Does the pronoun form favor the subject or the object?

Rate of subject responses

subject object







Las niñas saludan a la maestra... The girls at the teacher... wave subject ...y después sale.

> ... and then leaves. **PRONOUN**

> > sg

Rate of subject responses



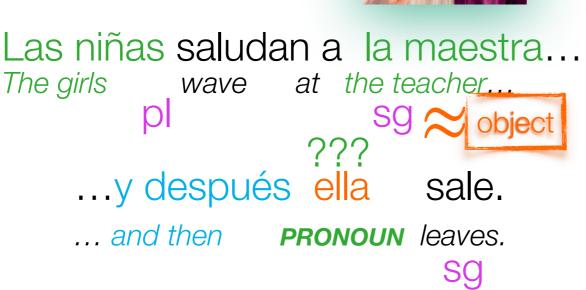
Context: Does the pronoun form favor the subject or the object?

Ø (favors subject)

subject object









~5yrs

Context:

Rate of subject responses

~3yrs

Does the pronoun form favor the subject or the object?

4yrs

overt (favors object)

subject object



Rate of subject responses











Las niñas saludan a la maestra...

The girls wave at the teacher...

pl sg

...y después

sale.

... and then **PRONOUN** leaves.

sg

Context:

Does the connective favor the subject or the object?

Favored by form

Ø (favors subject) overt (favors object)

subject object













Las niñas saludan a la maestra...

The girls wave subject p

at the teacher...

S9 ???

...y después

sale.

... and then **PRONOUN** leaves.

sg

Context:

Does the connective favor the subject or the object?

y después (favors subject)

subject object

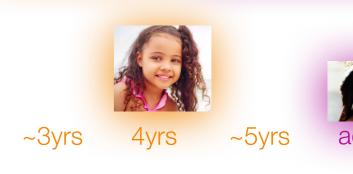












Las niñas saludan a la maestra...

The girls wave at the teacher...

...porque

... because

object object

sale.

PRONOUN leaves.

sg

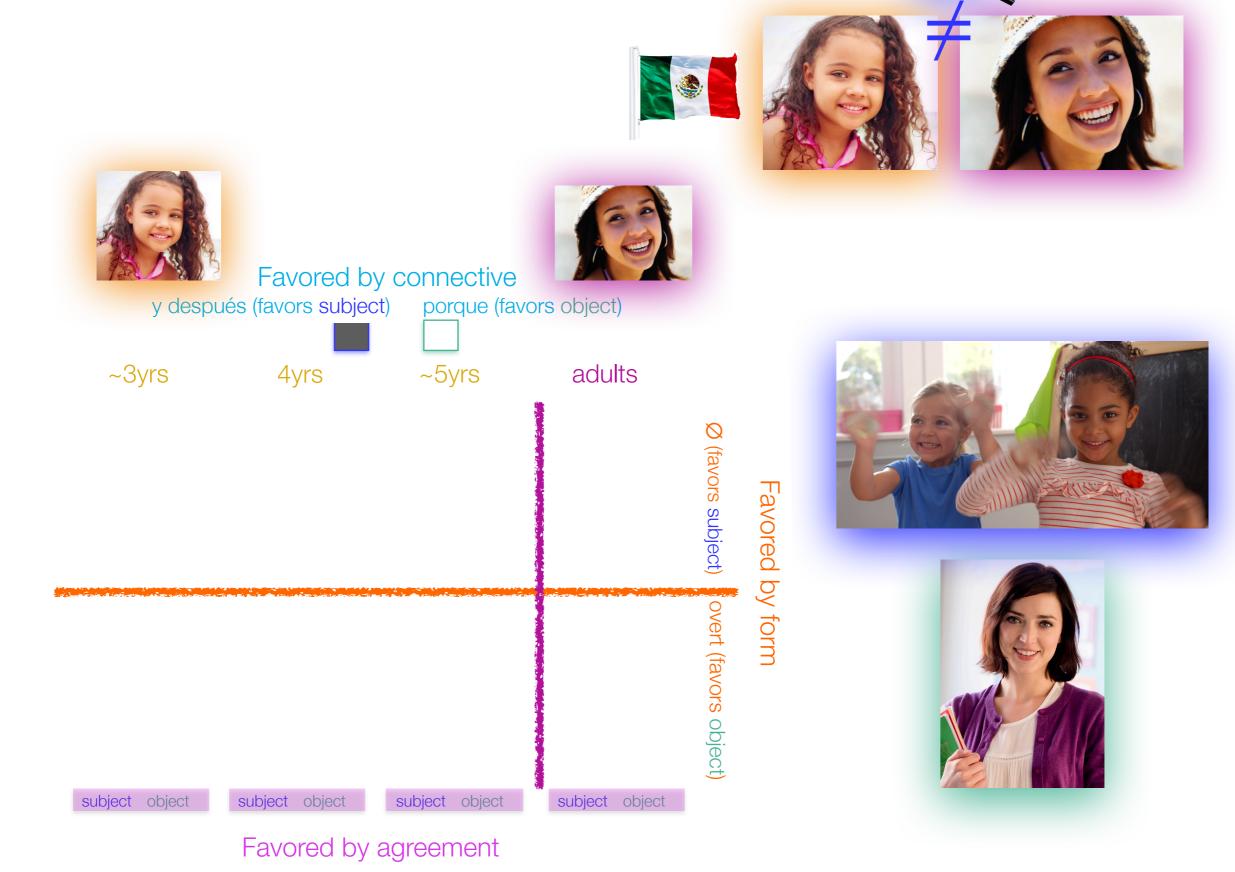
Context:

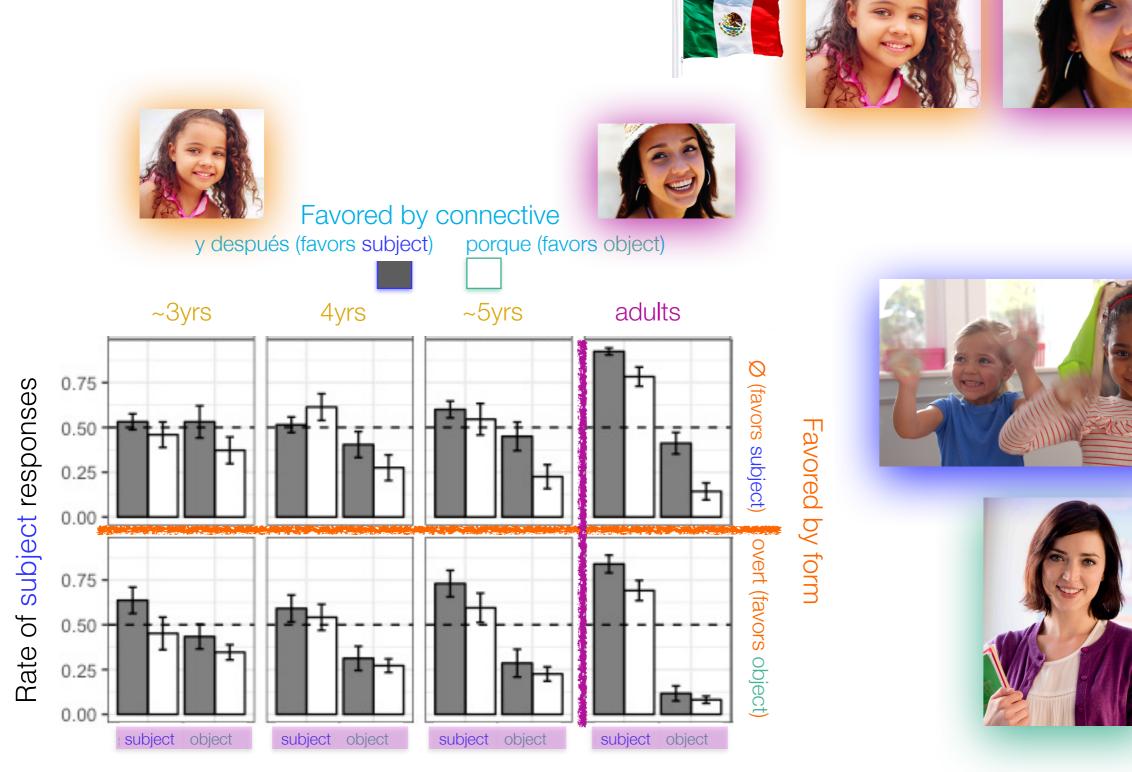
Does the connective favor the subject or the object?

porque (favors object)

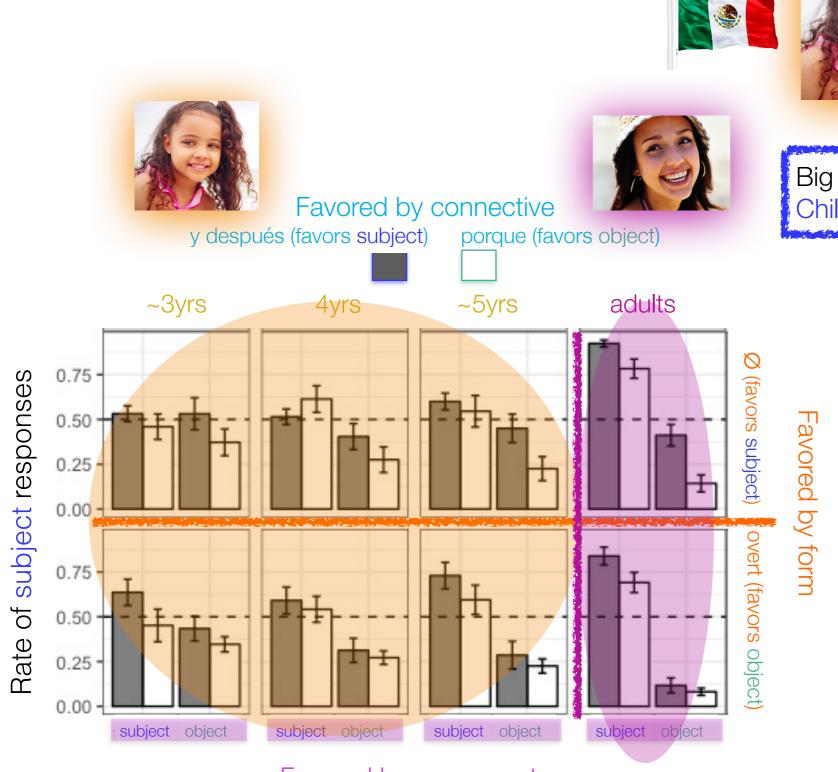
subject object

Rate of subject responses

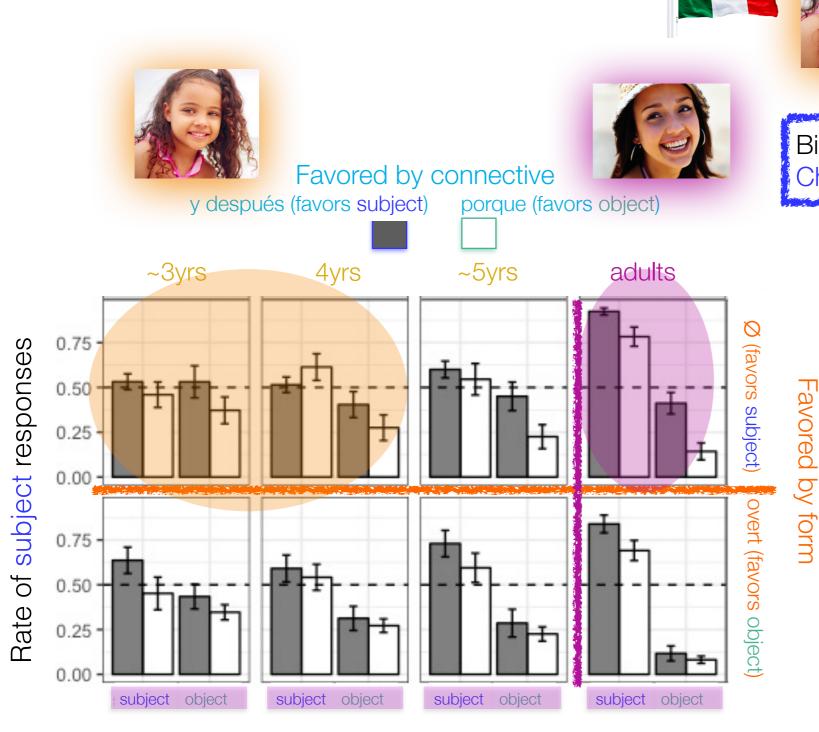




Favored by agreement



Big qualitative picture: Children don't behave like adults.



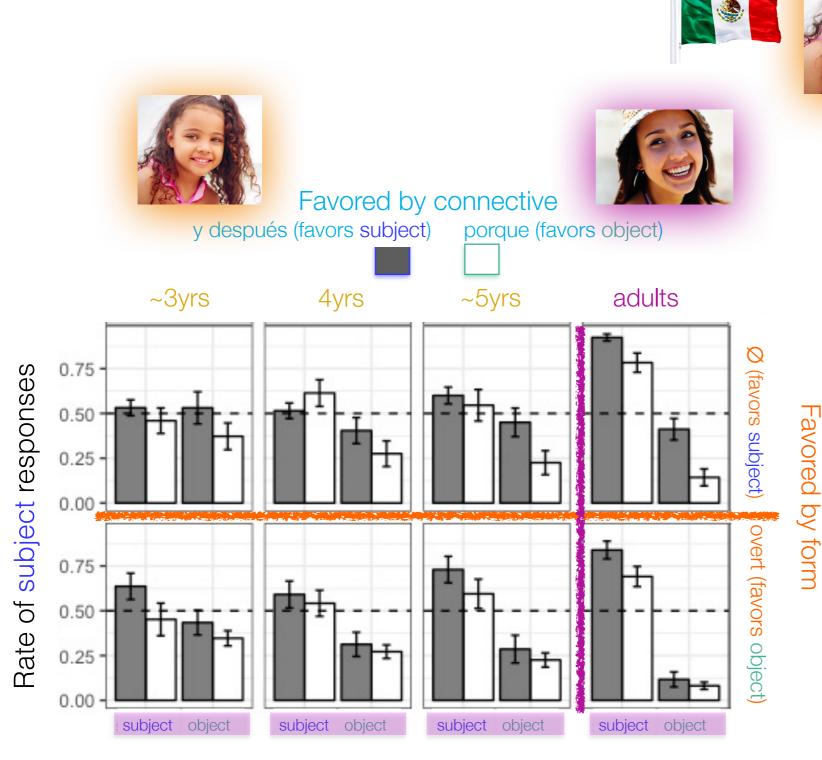
Big qualitative picture:

Children don't behave like adults.

Some nuances:

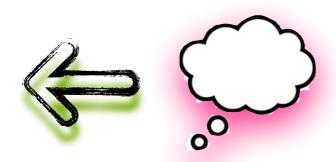
Younger children seem to differ more, especially in certain contexts.

Favored by agreement



Something needs to change for children to become adult-like — but what?

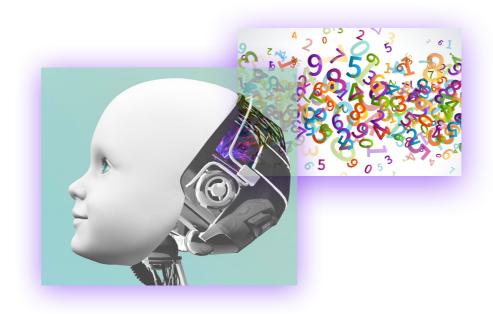




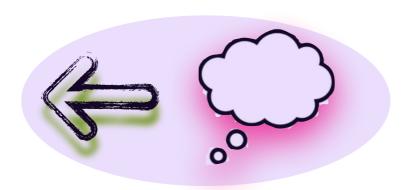
The plan, part 2:

Use computational cognitive modeling to formally articulate the potential process of pronoun interpretation in the context of these multiple cues.



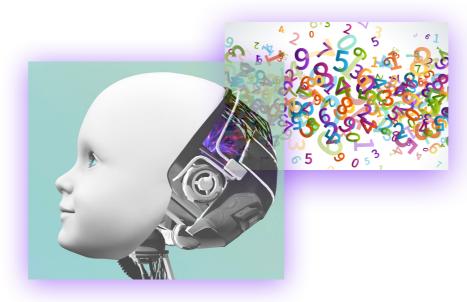




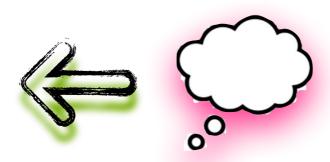


The computational cognitive model formally articulates and implements (what we think are) relevant aspects of pronoun interpretation in context.

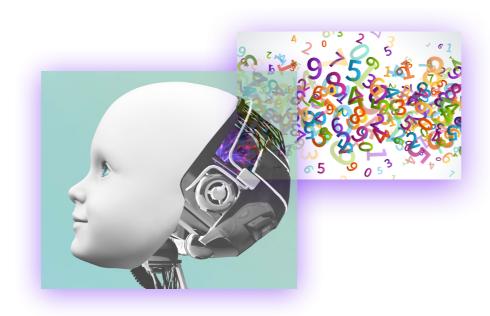




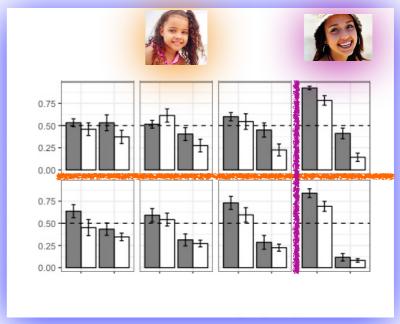




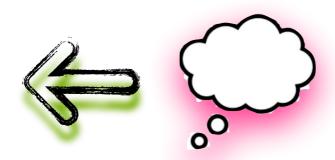
Here: Implement how a modeled listener represents pronoun information and deploys that information in order to predict the probability of a particular interpretation in context.





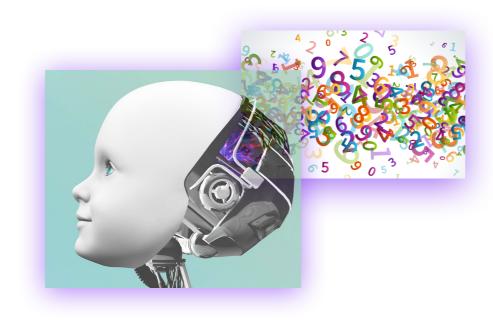


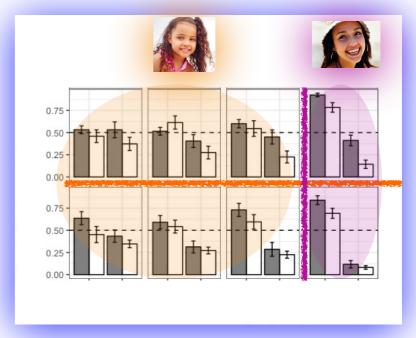




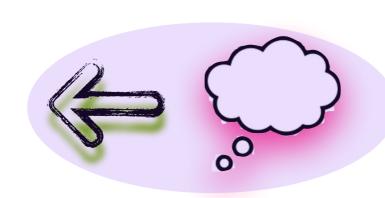


Then, see which options for representation and deployment best match child vs. adult pronoun interpretation behavior.



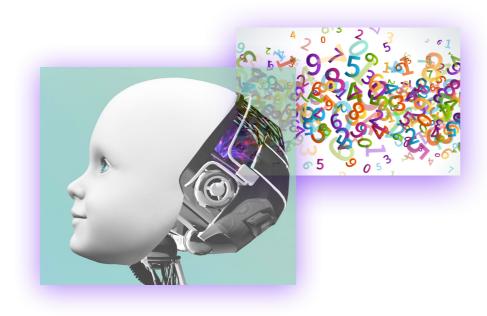


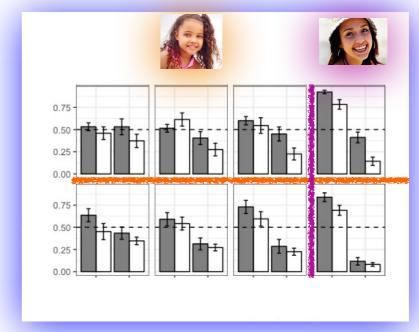


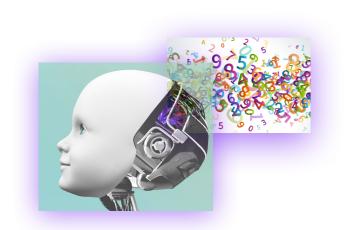


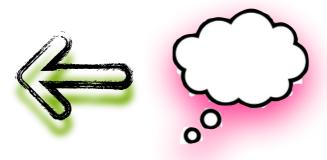


This will articulate how children differ from adults, and what needs to develop in children for them to become adult-like.



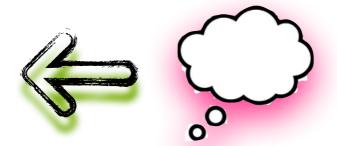






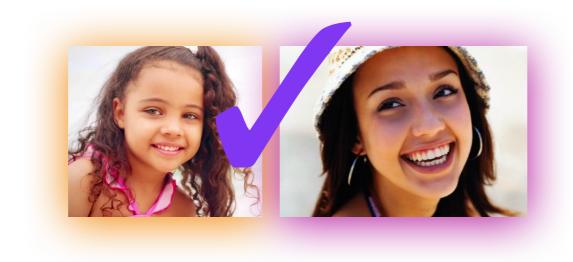
The model uses Bayesian inference to implement the cognitive mechanism that combines information to generate a particular interpretation in context.

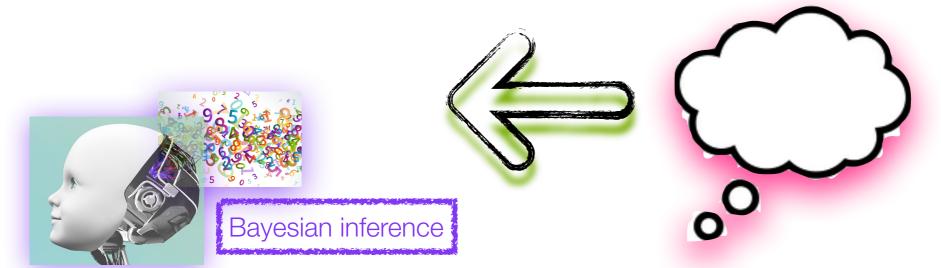




Bayesian inference is commonly used to model human cognition in general and language development in particular, since it matches human behavior quite well (see Pearl in press and b for recent reviews).



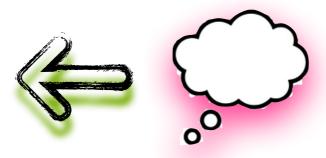




The particular Bayesian model we use is adapted from Gagliardi, Feldman & Lidz (2017), and offers one way to separate out the contributions of information representation vs. information deployment in the moment.









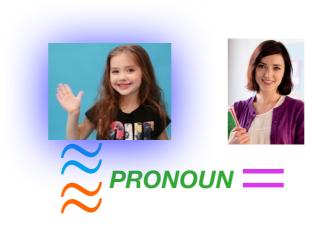
interpretation

 $p(\alpha_{subj.SG}|\text{FORM, CON, MOR}) \propto p(\alpha_{subj.SG}) * p(\text{FORM, CON, MOR}|\alpha_{subj.SG})$

Interpreting the pronoun as the subject, which is singular....



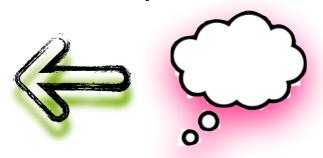


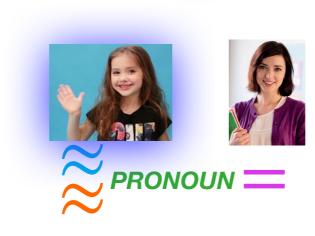


interpretation CONTEXT
$$p(\alpha_{subj.\rm SG}|{\rm FORM,CON,MOR}) \propto p(\alpha_{subj.\rm SG}) * p({\rm FORM,CON,MOR}|\alpha_{subj.\rm SG})$$

...given the particular context involving the pronoun's form,





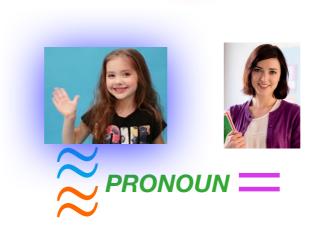


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interpretation CONTEXT p(\alpha_{subj.\text{SG}}|\text{FORM},\text{CON},\text{MOR}) \propto p(\alpha_{subj.\text{SG}}) * p(\text{FORM},\text{CON},\text{MOR}|\alpha_{subj.\text{SG}}) y \ despu\'es \quad porque
```

...given the particular context involving the pronoun's form, the connective,







```
interpretation CONTEXT p(\alpha_{subj.\rm SG}|{\rm FORM,CON,MOR}) \propto p(\alpha_{subj.\rm SG}) * p({\rm FORM,CON,MOR}|\alpha_{subj.\rm SG})
```

...given the particular context involving the pronoun's form, the connective, and the agreement morphology.



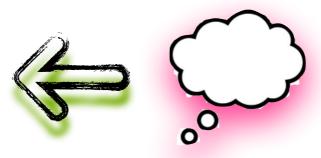




```
interpretation context p(\alpha_{subj.SG}|	extbf{FORM},	extbf{CON},	extbf{MOR}) \propto p(\alpha_{subj.SG}) * p(	extbf{FORM},	extbf{CON},	extbf{MOR}|\alpha_{subj.SG})
```

This is proportional to the prior probability of that interpretation irrespective of this particular context...



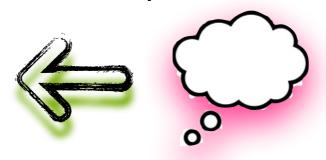




interpretation context
$$p(\alpha_{subj.SG}| extsf{FORM}, extsf{CON}, extsf{MOR}) \propto p(\alpha_{subj.SG}) * p(extsf{FORM}, extsf{CON}, extsf{MOR}|\alpha_{subj.SG})$$

...multiplied by the likelihood of these context values, given this kind of interpretation (a singular subject).



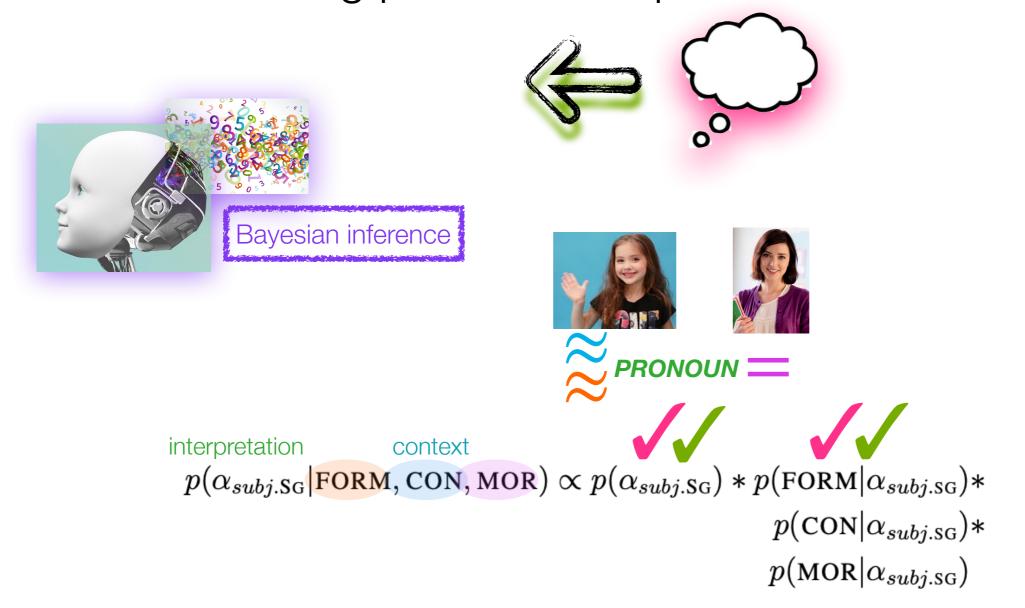




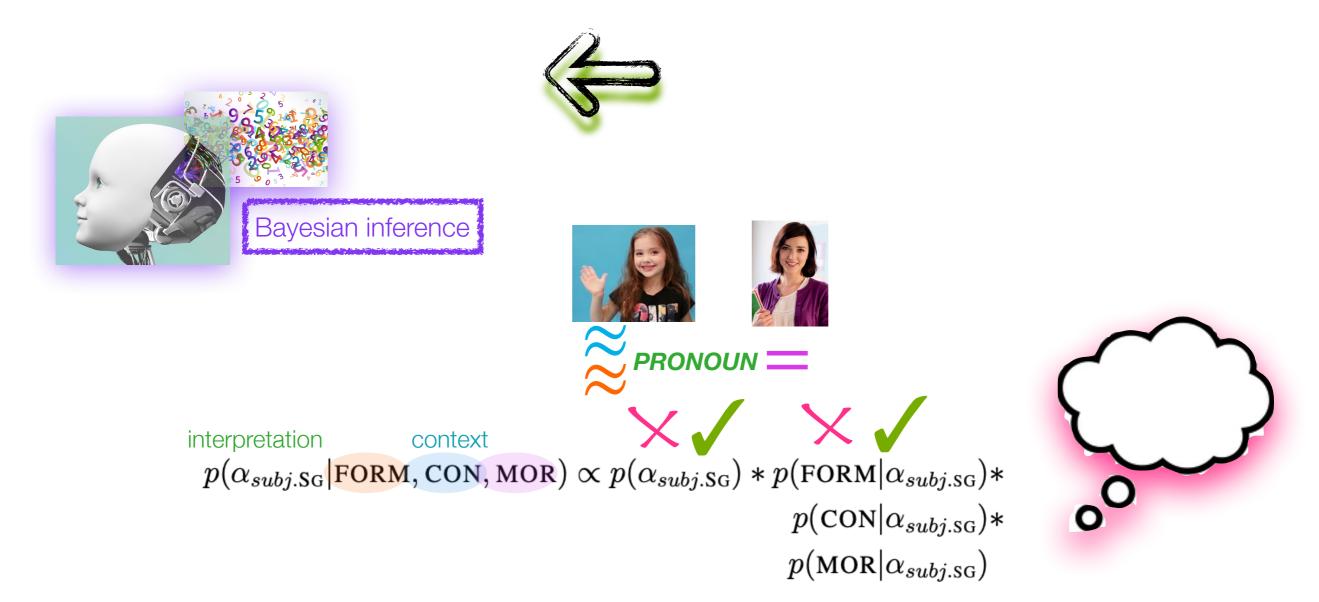
interpretation context
$$p(\alpha_{subj.SG}| extbf{FORM}, extbf{CON}, extbf{MOR}) \propto p(\alpha_{subj.SG}) * p(extbf{FORM}|\alpha_{subj.SG}) * p(extbf{CON}|\alpha_{subj.SG}) * p(extbf{MOR}|\alpha_{subj.SG})$$

Here, we assume these context values are independent, so we can calculate the likelihood this way.

(For example, the probability of a particular pronoun form, given a certain interpretation, is assumed to be independent from the probability of a particular connective, given a certain interpretation.)



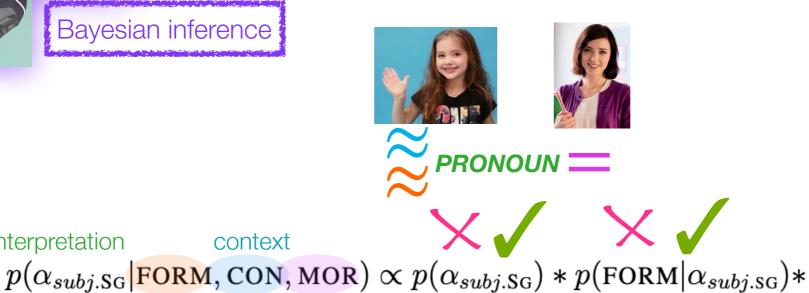
This is the baseline model, which has accurate representations of information and accurately deploys those representations in the moment.



What about a modeled listener who has inaccurate representations? This could involve inaccurately representing the prior or the likelihood information, or both.







 $p(\text{CON}|\alpha_{subj.\text{SG}})*$

 $p(\text{MOR}|\alpha_{subj.sg})$



inaccurate representations

context

interpretation

We implement this as a softmax on the true probability (prior or likelihood), with contrast parameter σ .

 $e^{\sigma^* ln(probability)} = probability^{\sigma}$

















 $p(\alpha_{subj.SG}|\text{FORM}, \text{CON}, \text{MOR}) \propto p(\alpha_{subj.SG}) * p(\text{FORM}|\alpha_{subj.SG}) *$

 $p(\text{CON}|\alpha_{subj.\text{SG}})*$

 $p(\text{MOR}|\alpha_{subj.\text{SG}})$



interpretation











 σ =1: probabilities remain accurate.

 σ >1: probability differences are sharpened.

context

0.324 vs. 0.676

Forsythe & Pearl 2019, in prep













context

 $p(\alpha_{subj.SG}|\text{FORM}, \text{CON}, \text{MOR}) \propto p(\alpha_{subj.SG}) * p(\text{FORM}|\alpha_{subj.SG}) *$





 $p(\text{CON}|\alpha_{subj.\text{SG}})*$

 $p(\text{MOR}|\alpha_{subj.SG})$



probability σ







 $\sigma = 0.5$: probability differences are smoothed away.

 σ =1: probabilities remain accurate.

 σ >1: probability differences are sharpened.

0.409 vs. 0.591

0.324 vs. 0.676

Forsythe & Pearl 2019, in prep







interpretation

context

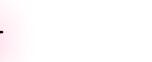
 $p(\alpha_{subj.SG}|\text{FORM}, \text{CON}, \text{MOR}) \propto p(\alpha_{subj.SG}) * p(\text{FORM}|\alpha_{subj.SG}) *$

 $p(\text{CON}|\alpha_{subj.\text{SG}})*$

 $p(\text{MOR}|\alpha_{subj.SG})$



About σ : - \mathfrak{O} -





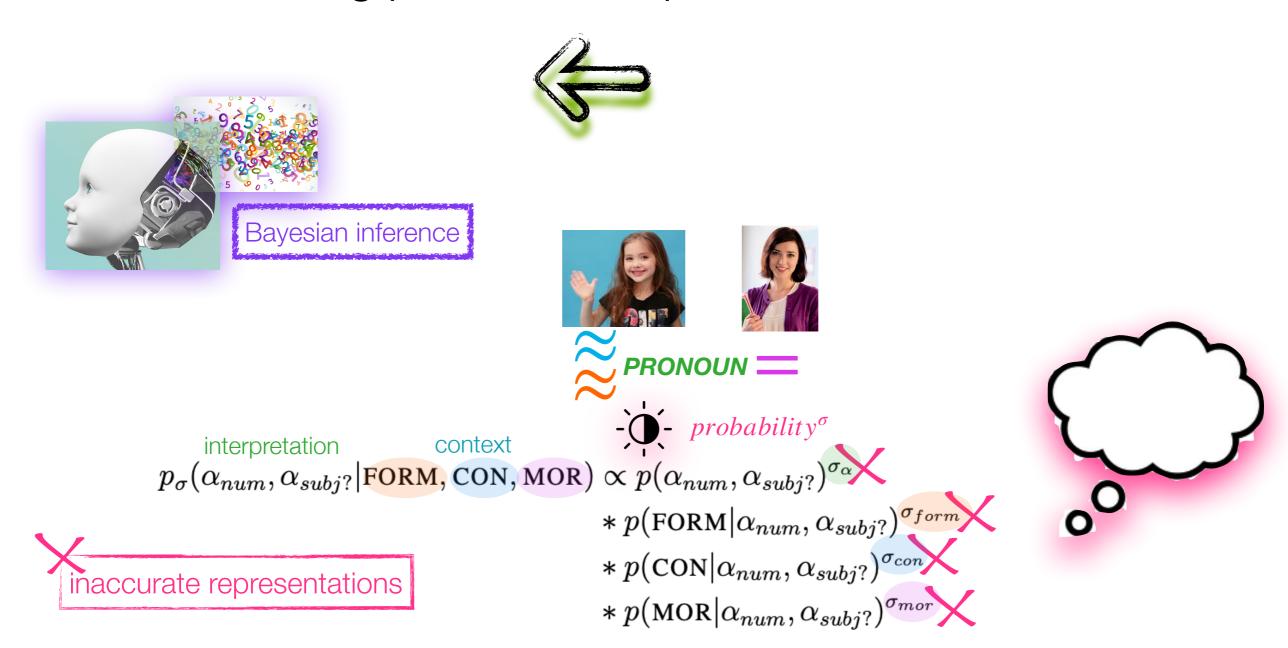
probability $^{\sigma}$

 σ =1: probabilities remain accurate.

 $\sigma = 2$: probability differences are sharpened.

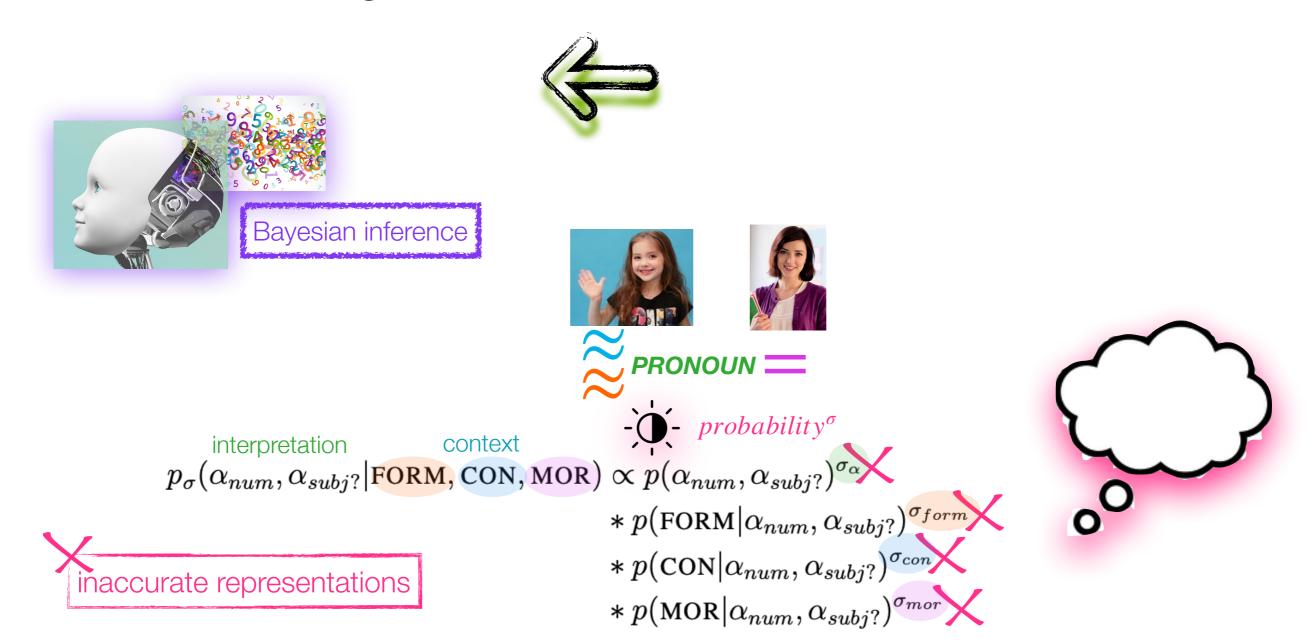


0.324 vs. 0.676 0.187 vs. 0.813 Forsythe & Pearl 2019, in prep



One σ for each information type:

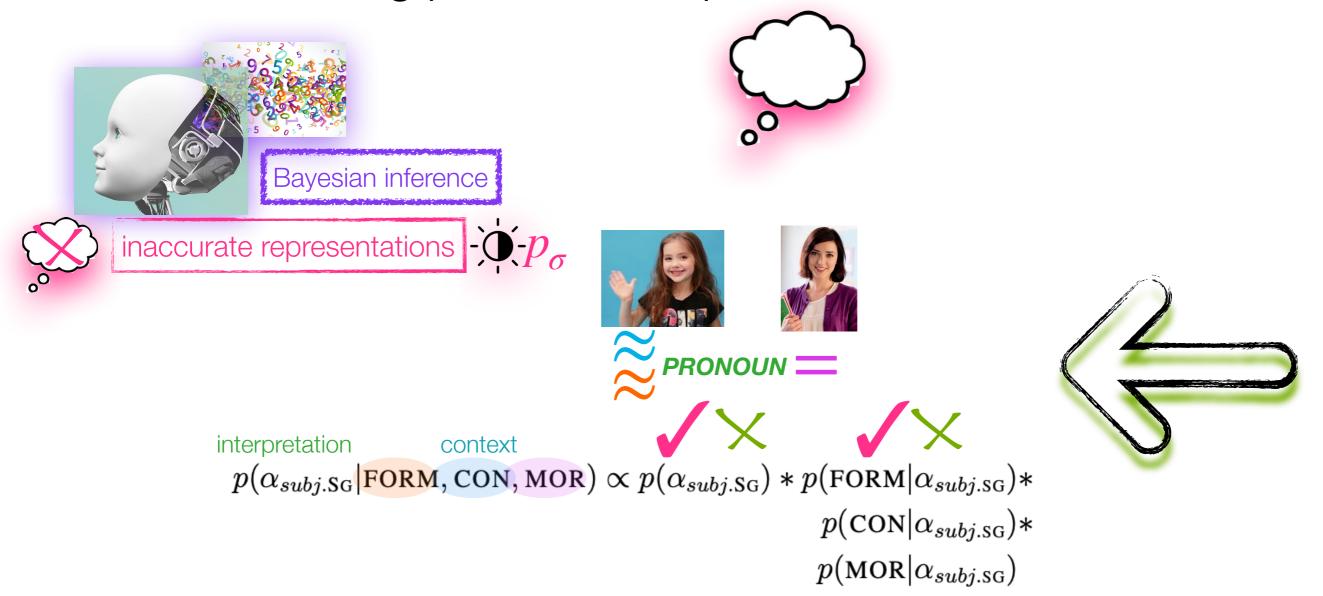
(in the prior) σ_{α} (in the likelihood) σ_{form} , σ_{con} , σ_{mor}



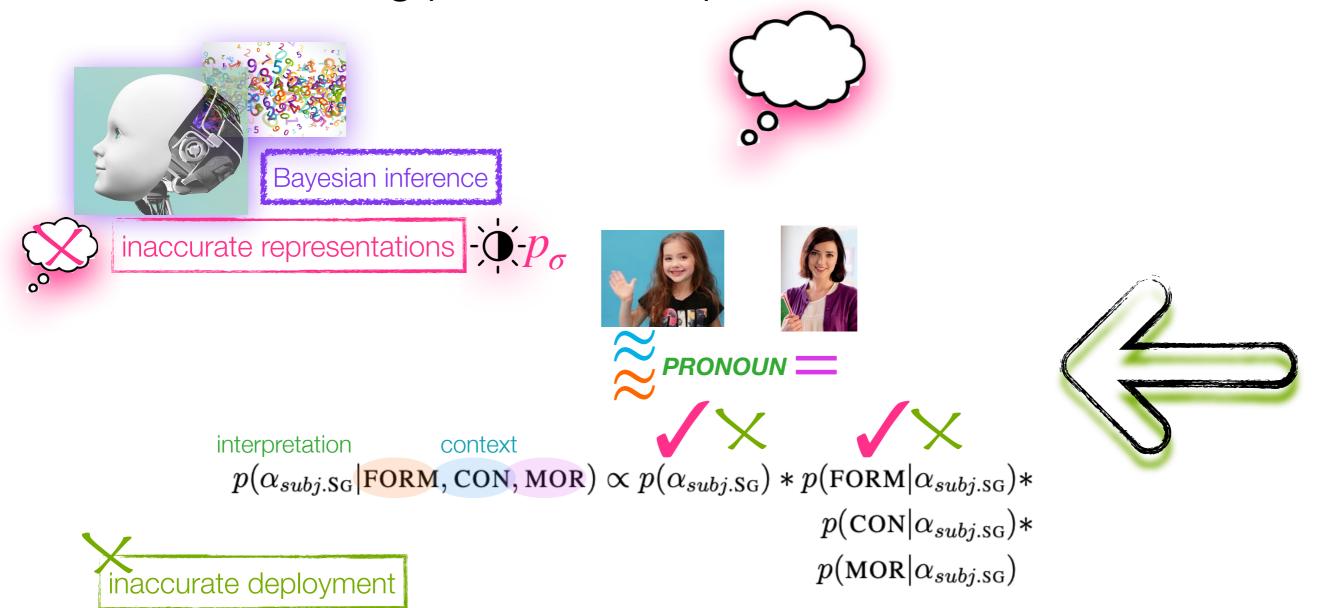
We allow 0.01 $\leq \sigma \leq$ 4, and see which σ value combinations best predict child and adult pronoun interpretation behavior.



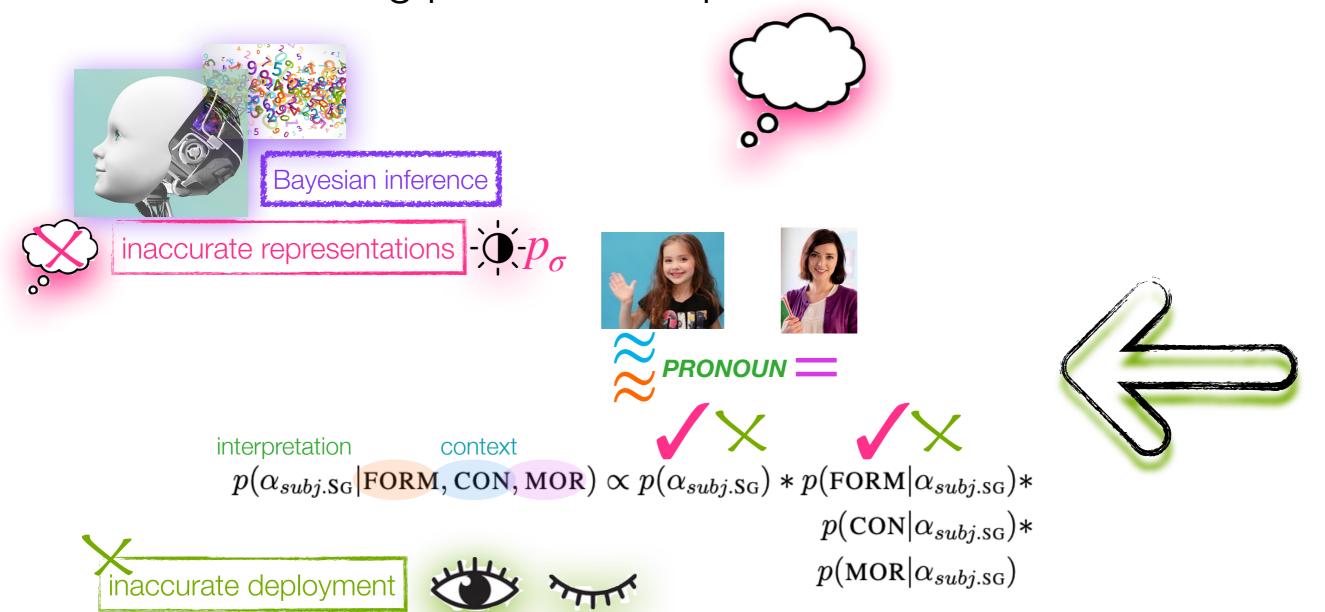




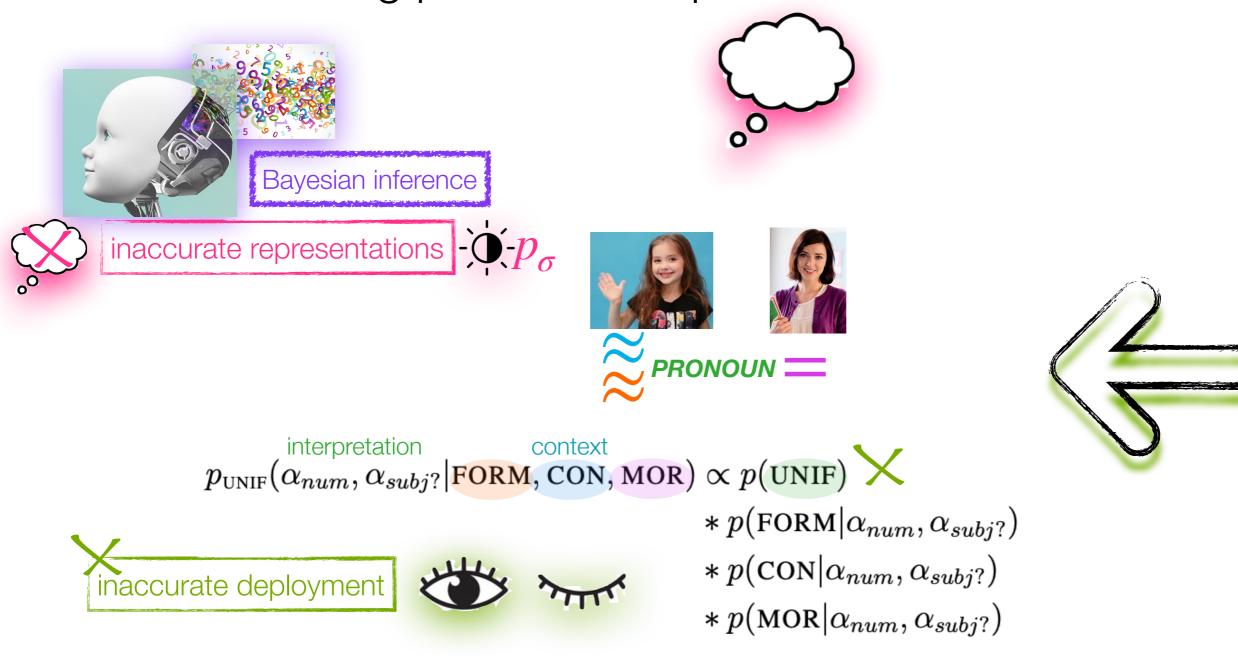
What about a modeled listener who has inaccurate deployment of information in the representations? This could involve inaccurately deploying the prior or the likelihood information, or both.



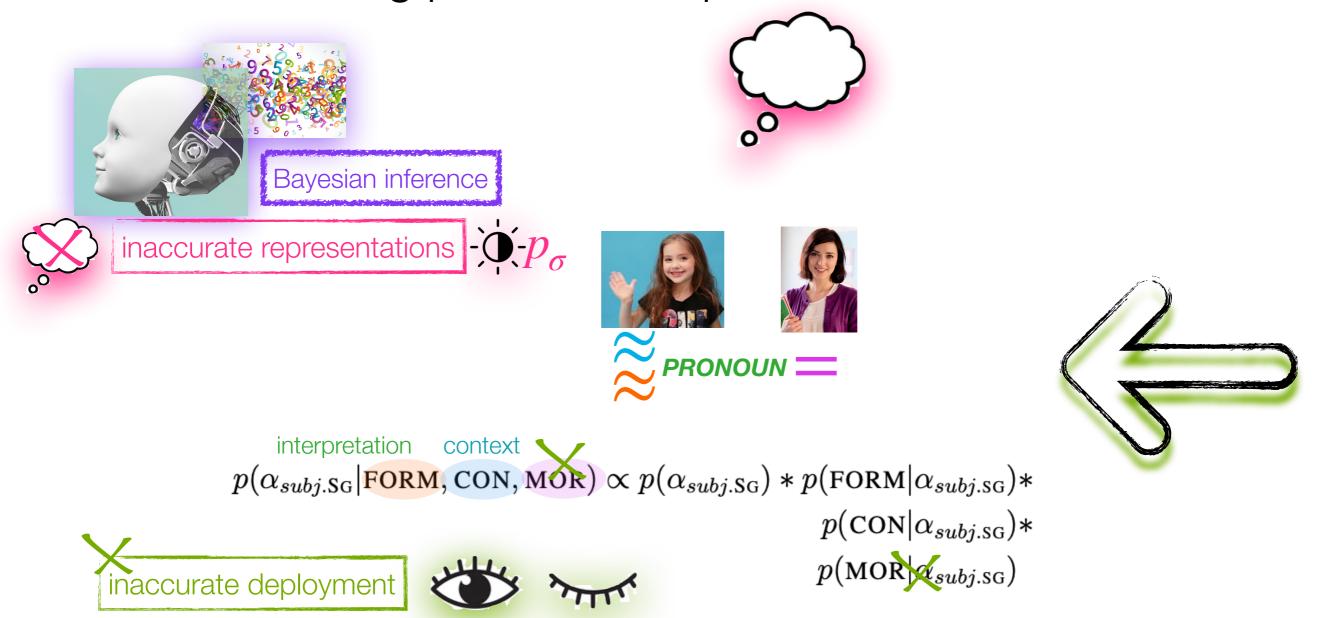
We implement this as ignoring that information. So, for any piece of information, the modeled listener either pays attention to it (and so uses it) or ignores it in the moment.



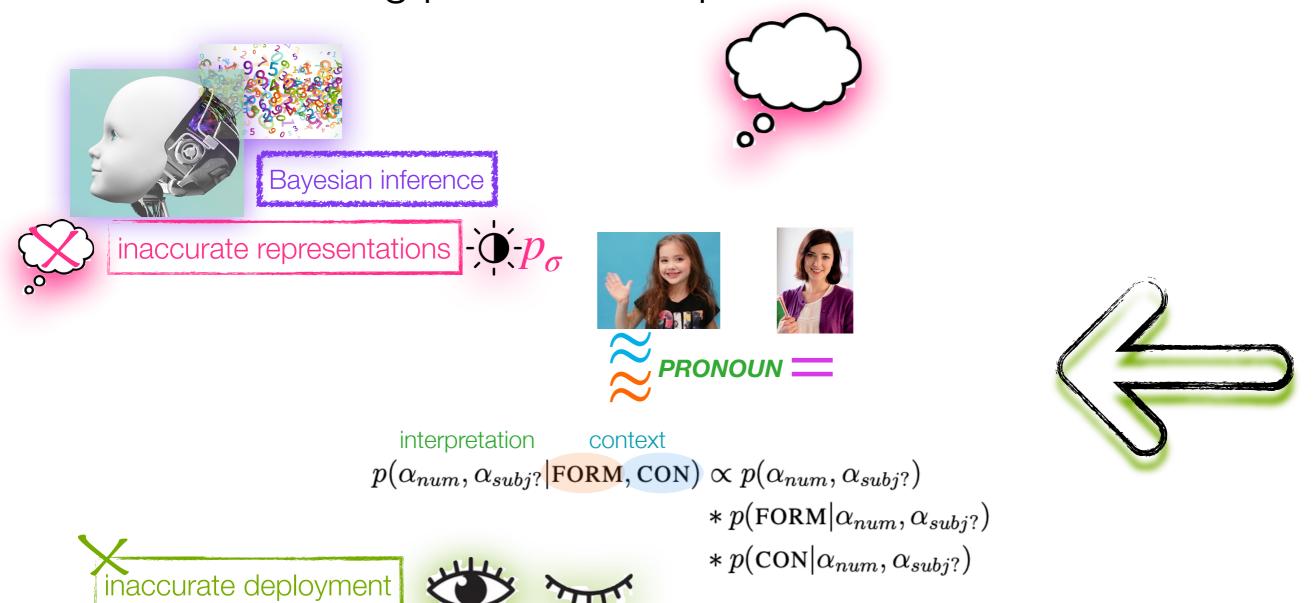
Not using information means not incorporating it into the inference.



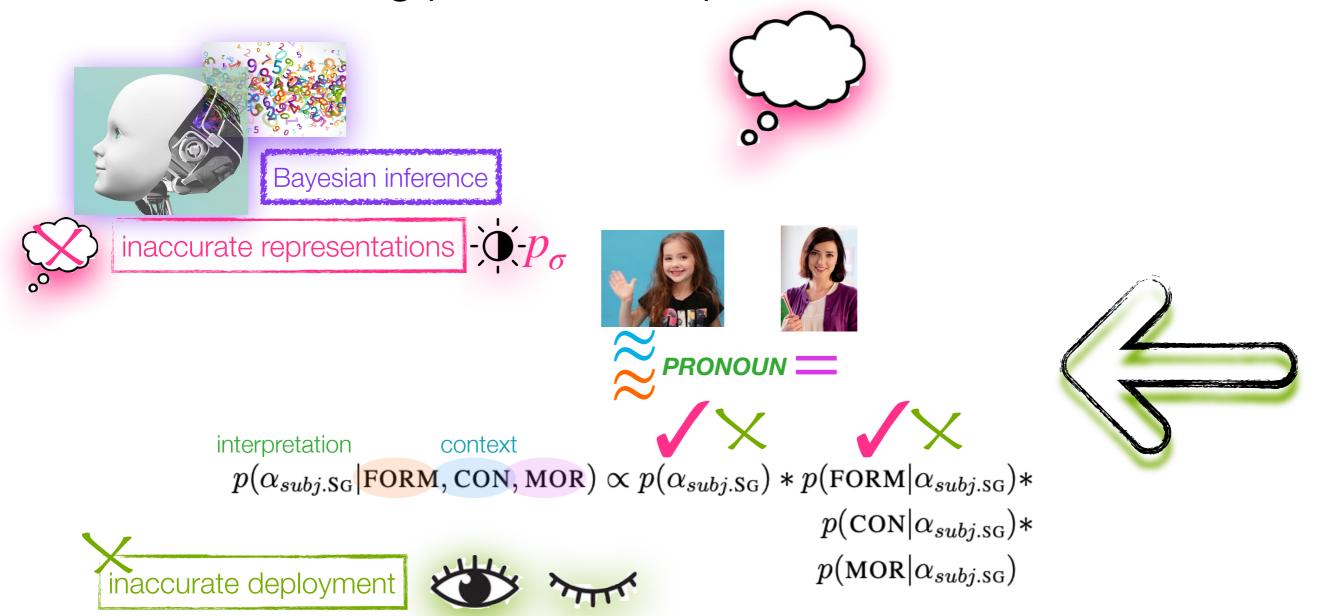
Not using the prior means relying on a uniform prior.



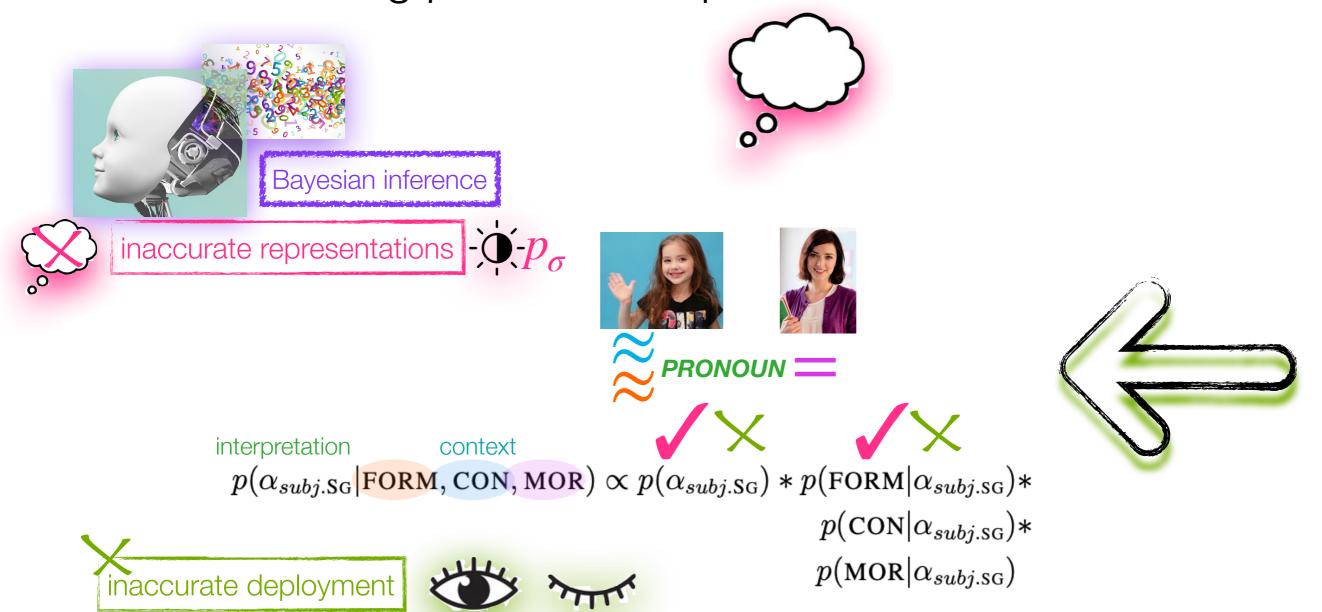
Not using likelihood information for a cue means not using that cue's information. For example, ignoring morphology information means not using the morphology likelihood.



Not using likelihood information for a cue means not using that cue's information. For example, ignoring morphology information means not using the morphology likelihood.



For any information, the modeled listener could use or not use it in the moment. Use parameter β determines whether a particular information type is used.



Each of the four information types has its own β : (prior) β_{α} (likelihood) β_{form} , β_{con} , β_{mor}

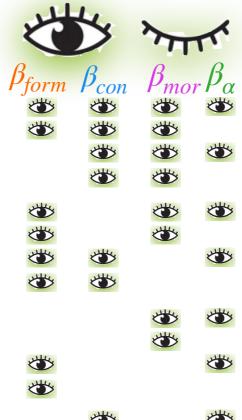


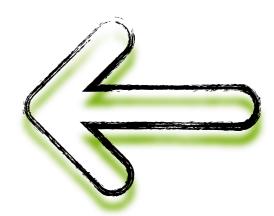


inaccurate deployment

interpretation context $p_{\beta}(\alpha|\text{FORM}, \text{CON}, \text{MOR}, \alpha_{num}, \alpha_{subj?}) =$ $(\beta_{form})(\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|\text{FORM, CON, MOR}, \alpha_{num}, \alpha_{subj?}) +$ $(\beta_{form})(\beta_{con})(\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM, CON, MOR, }\alpha_{num},\alpha_{subj?})+$ $(1 - \beta_{form})(\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha | CON, MOR, \alpha_{num}, \alpha_{subj?}) +$ $(1 - \beta_{form})(\beta_{con})(\beta_{mor})(1 - \beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{CON}, \text{MOR}, \alpha_{num}, \alpha_{subi?}) +$ $(\beta_{form})(1-\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|FORM, MOR, \alpha_{num}, \alpha_{subj?}) +$ $(\beta_{form})(1-\beta_{con})(\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM},\text{MOR},\alpha_{num},\alpha_{subj?})+$ $(\beta_{form})(\beta_{con})(1-\beta_{mor})(\beta_{\alpha})*p(\alpha|\text{FORM},\text{CON},\alpha_{num},\alpha_{subj?})+$ $(\beta_{form})(\beta_{con})(1-\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM},\text{CON},\alpha_{num},\alpha_{subj?})+$ $(1 - \beta_{form})(1 - \beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|MOR, \alpha_{num}, \alpha_{subi?}) +$ $(1 - \beta_{form})(1 - \beta_{con})(\beta_{mor})(1 - \beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{MOR}, \alpha_{num}, \alpha_{subj?}) +$ $(\beta_{form})(1-\beta_{con})(1-\beta_{mor})(\beta_{\alpha}) * p(\alpha|FORM, \alpha_{num}, \alpha_{subi?}) +$ $(\beta_{form})(1-\beta_{con})(1-\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM},\alpha_{num},\alpha_{subj?})+$ $(1 - \beta_{form})(\beta_{con})(1 - \beta_{mor})(\beta_{\alpha}) * p(\alpha | CON, \alpha_{num}, \alpha_{subi?}) +$ $(1 - \beta_{form})(\beta_{con})(1 - \beta_{mor})(1 - \beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{CON}, \alpha_{num}, \alpha_{subj?}) +$ $(1 - \beta_{form})(1 - \beta_{con})(1 - \beta_{mor})(\beta_{\alpha}) * p(\alpha_{num}, \alpha_{subj?}) +$ $(1 - \beta_{form})(1 - \beta_{con})(1 - \beta_{mor})(1 - \beta_{o}) * p(UNIF)$





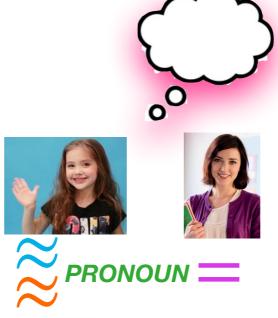


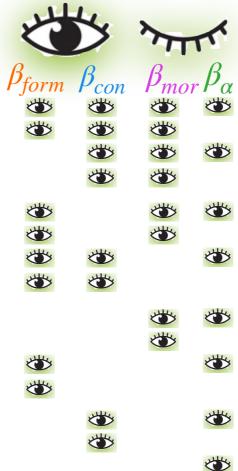
This yields 16 possible use combinations for any particular moment, implemented with a mixture model p_{β} .

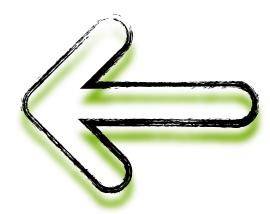




interpretation context $p_{\beta}(\alpha|\text{FORM}, \text{CON}, \text{MOR}, \alpha_{num}, \alpha_{subj?}) =$ $(\beta_{form})(\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|\text{FORM, CON, MOR}, \alpha_{num}, \alpha_{subj?}) +$ $(\beta_{form})(\beta_{con})(\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM, CON, MOR},\alpha_{num},\alpha_{subj?})+$ $(1 - \beta_{form})(\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha | CON, MOR, \alpha_{num}, \alpha_{subj?}) +$ $(1 - \beta_{form})(\beta_{con})(\beta_{mor})(1 - \beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{CON}, \text{MOR}, \alpha_{num}, \alpha_{subj?}) +$ $(\beta_{form})(1-\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|FORM, MOR, \alpha_{num}, \alpha_{subj?}) +$ $(\beta_{form})(1-\beta_{con})(\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM},\text{MOR},\alpha_{num},\alpha_{subj?})+$ $(\beta_{form})(\beta_{con})(1-\beta_{mor})(\beta_{\alpha})*p(\alpha|\text{FORM},\text{CON},\alpha_{num},\alpha_{subj?})+$ $(\beta_{form})(\beta_{con})(1-\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM},\text{CON},\alpha_{num},\alpha_{subj?})+$ $(1 - \beta_{form})(1 - \beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|MOR, \alpha_{num}, \alpha_{subi?}) +$ $(1 - \beta_{form})(1 - \beta_{con})(\beta_{mor})(1 - \beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{MOR}, \alpha_{num}, \alpha_{subj?}) +$ $(\beta_{form})(1-\beta_{con})(1-\beta_{mor})(\beta_{\alpha}) * p(\alpha|FORM, \alpha_{num}, \alpha_{subi?}) +$ $(\beta_{form})(1-\beta_{con})(1-\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM},\alpha_{num},\alpha_{subj?})+$ $(1 - \beta_{form})(\beta_{con})(1 - \beta_{mor})(\beta_{\alpha}) * p(\alpha | CON, \alpha_{num}, \alpha_{subi?}) +$ $(1 - \beta_{form})(\beta_{con})(1 - \beta_{mor})(1 - \beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{CON}, \alpha_{num}, \alpha_{subj?}) +$ $(1 - \beta_{form})(1 - \beta_{con})(1 - \beta_{mor})(\beta_{\alpha}) * p(\alpha_{num}, \alpha_{subj?}) +$ $(1 - \beta_{form})(1 - \beta_{con})(1 - \beta_{mor})(1 - \beta_{o}) * p(UNIF)$



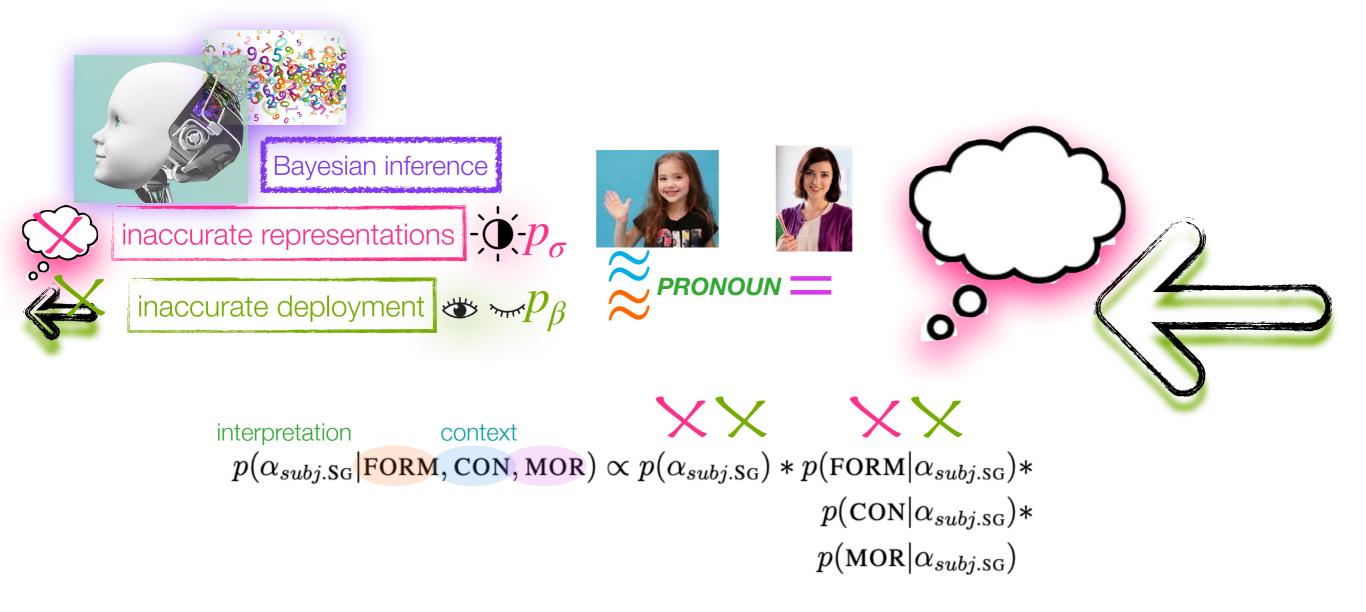




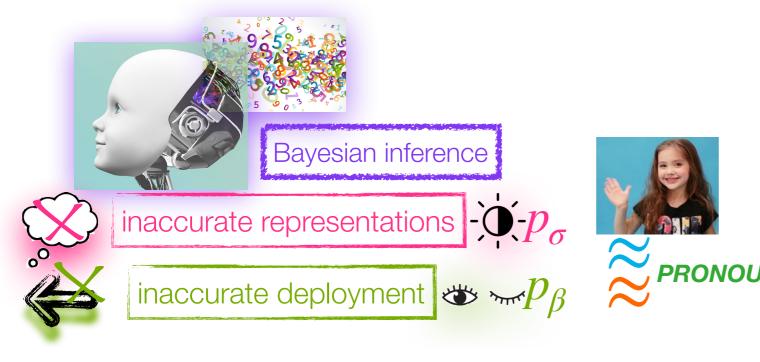
We allow $0 \le \beta \le 1$, and see which β value combinations best predict child and adult pronoun interpretation behavior.



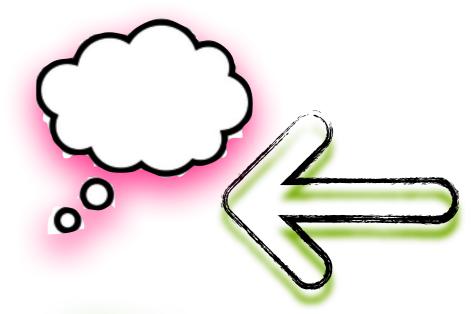




What about a modeled listener who has both inaccurate representations of information and inaccurate deployment of those representations?







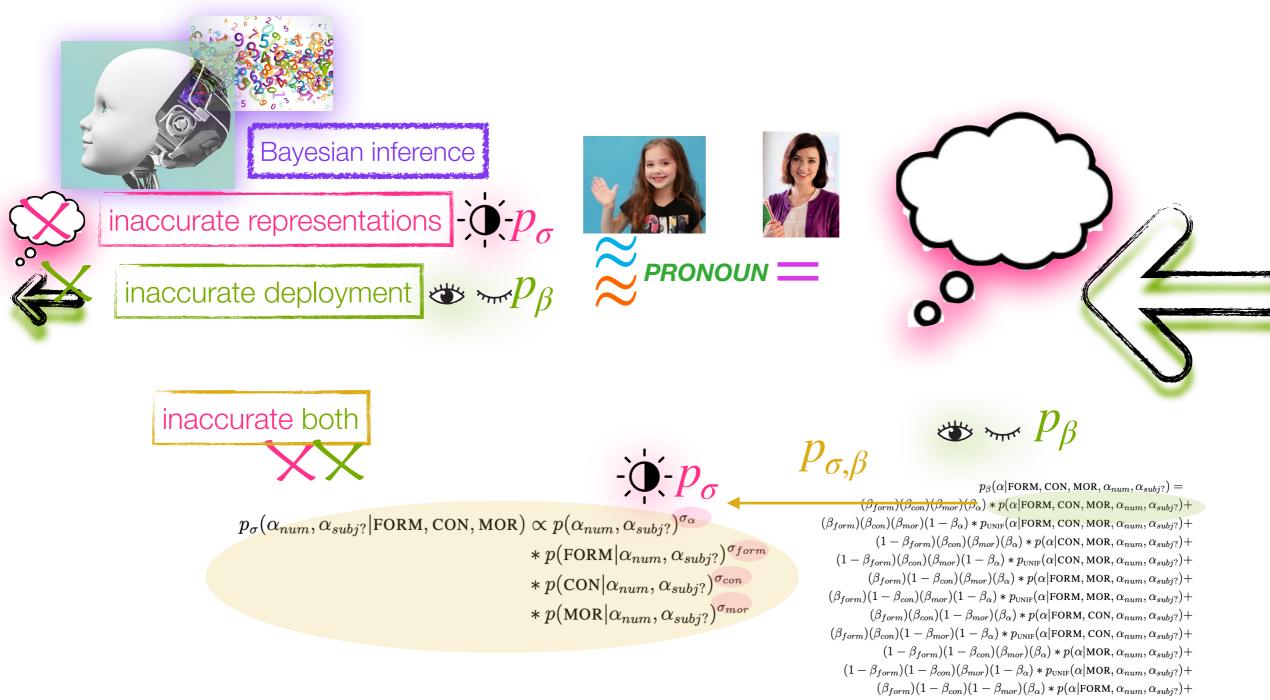


$$p_{\sigma}(\alpha_{num}, \alpha_{subj?}| ext{FORM, CON, MOR}) \propto p(\alpha_{num}, \alpha_{subj?})^{\sigma_{\alpha}} \\ * p(ext{FORM}|\alpha_{num}, \alpha_{subj?})^{\sigma_{form}} \\ * p(ext{CON}|\alpha_{num}, \alpha_{subj?})^{\sigma_{con}} \\ * p(ext{MOR}|\alpha_{num}, \alpha_{subj?})^{\sigma_{mor}}$$

We implement this as a combination of the previous two modeled listeners, including σ values for inaccurate representations and β values for inaccurate deployment.

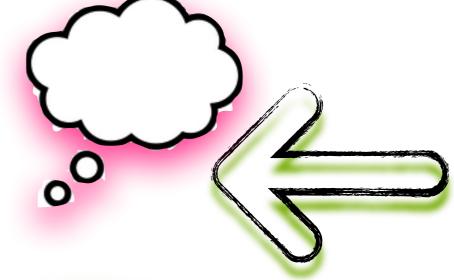
```
P_{\beta}
```

```
p_{\sigma,\beta}
                                                                  p_{\beta}(\alpha|\text{FORM}, \text{CON}, \text{MOR}, \alpha_{num}, \alpha_{subj?}) =
                       (\beta_{form})(\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|FORM, CON, MOR, \alpha_{num}, \alpha_{subj?}) +
         (\beta_{form})(\beta_{con})(\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM, CON, MOR},\alpha_{num},\alpha_{subj?})+
                             (1 - \beta_{form})(\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha | CON, MOR, \alpha_{num}, \alpha_{subj?}) +
              (1 - \beta_{form})(\beta_{con})(\beta_{mor})(1 - \beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{CON}, \text{MOR}, \alpha_{num}, \alpha_{subj?}) +
                          (\beta_{form})(1-\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|FORM, MOR, \alpha_{num}, \alpha_{subj?}) +
           (\beta_{form})(1-\beta_{con})(\beta_{mor})(1-\beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{FORM}, \text{MOR}, \alpha_{num}, \alpha_{subj?}) +
                           (\beta_{form})(\beta_{con})(1-\beta_{mor})(\beta_{\alpha}) * p(\alpha|FORM, CON, \alpha_{num}, \alpha_{subj?}) +
             (\beta_{form})(\beta_{con})(1-\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM, CON},\alpha_{num},\alpha_{subj?})+
                               (1 - \beta_{form})(1 - \beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|MOR, \alpha_{num}, \alpha_{subj?}) +
                 (1 - \beta_{form})(1 - \beta_{con})(\beta_{mor})(1 - \beta_{\alpha}) * p_{\text{UNIF}}(\alpha | \text{MOR}, \alpha_{num}, \alpha_{subj?}) +
                             (\beta_{form})(1-\beta_{con})(1-\beta_{mor})(\beta_{\alpha}) * p(\alpha|FORM, \alpha_{num}, \alpha_{subj?}) +
              (\beta_{form})(1-\beta_{con})(1-\beta_{mor})(1-\beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{FORM}, \alpha_{num}, \alpha_{subj?}) +
                                (1 - \beta_{form})(\beta_{con})(1 - \beta_{mor})(\beta_{\alpha}) * p(\alpha | CON, \alpha_{num}, \alpha_{subj?}) +
                 (1 - \beta_{form})(\beta_{con})(1 - \beta_{mor})(1 - \beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{CON}, \alpha_{num}, \alpha_{subj?}) +
                                       (1 - \beta_{form})(1 - \beta_{con})(1 - \beta_{mor})(\beta_{\alpha}) * p(\alpha_{num}, \alpha_{subj?}) +
                                            (1 - \beta_{form})(1 - \beta_{con})(1 - \beta_{mor})(1 - \beta_{\alpha}) * p(UNIF)
```

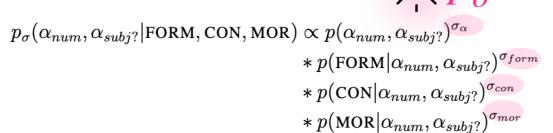


Each term of p_{β} has σ values for all information types that are used.





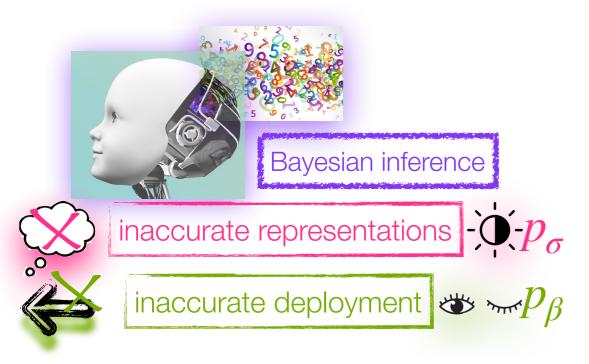




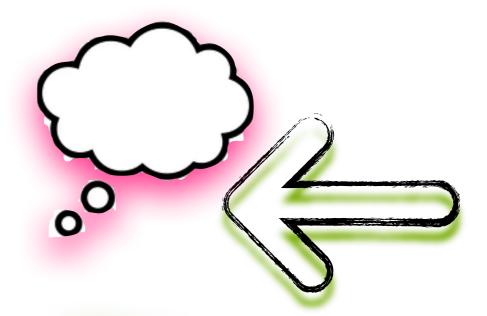
```
So, p_{\sigma,\beta} has 8 parameter values: (in the prior) \sigma_{\alpha}, \beta_{\alpha}
(in the likelihood) \sigma_{form}, \sigma_{con}, \sigma_{mor}, \beta_{form}, \beta_{con}, \beta_{mor}
```

 p_{β} $p_{\sigma,\beta}$

```
p_{\beta}(\alpha|\text{FORM}, \text{CON}, \text{MOR}, \alpha_{num}, \alpha_{subj?}) =
              (\beta_{form})(\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|FORM, CON, MOR, \alpha_{num}, \alpha_{subj?}) +
(\beta_{form})(\beta_{con})(\beta_{mor})(1-\beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{FORM, CON, MOR}, \alpha_{num}, \alpha_{subj?}) +
                    (1 - \beta_{form})(\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha | CON, MOR, \alpha_{num}, \alpha_{subj?}) +
     (1 - \beta_{form})(\beta_{con})(\beta_{mor})(1 - \beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{CON}, \text{MOR}, \alpha_{num}, \alpha_{subj?}) +
                 (\beta_{form})(1-\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|FORM, MOR, \alpha_{num}, \alpha_{subj?}) +
  (\beta_{form})(1-\beta_{con})(\beta_{mor})(1-\beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{FORM, MOR, }\alpha_{num},\alpha_{subj?}) +
                  (\beta_{form})(\beta_{con})(1-\beta_{mor})(\beta_{\alpha}) * p(\alpha|FORM, CON, \alpha_{num}, \alpha_{subj?}) +
    (\beta_{form})(\beta_{con})(1-\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM, CON},\alpha_{num},\alpha_{subj?})+
                      (1 - \beta_{form})(1 - \beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|MOR, \alpha_{num}, \alpha_{subj?}) +
        (1 - \beta_{form})(1 - \beta_{con})(\beta_{mor})(1 - \beta_{\alpha}) * p_{\text{UNIF}}(\alpha | \text{MOR}, \alpha_{num}, \alpha_{subj?}) +
                    (\beta_{form})(1-\beta_{con})(1-\beta_{mor})(\beta_{\alpha}) * p(\alpha|FORM, \alpha_{num}, \alpha_{subj?}) +
     (\beta_{form})(1-\beta_{con})(1-\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM},\alpha_{num},\alpha_{subj?})+
                       (1 - \beta_{form})(\beta_{con})(1 - \beta_{mor})(\beta_{\alpha}) * p(\alpha | CON, \alpha_{num}, \alpha_{subj?}) +
        (1 - \beta_{form})(\beta_{con})(1 - \beta_{mor})(1 - \beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{CON}, \alpha_{num}, \alpha_{subj?}) +
                              (1 - \beta_{form})(1 - \beta_{con})(1 - \beta_{mor})(\beta_{\alpha}) * p(\alpha_{num}, \alpha_{subj?}) +
                                  (1 - \beta_{form})(1 - \beta_{con})(1 - \beta_{mor})(1 - \beta_{\alpha}) * p(UNIF)
```











```
p_{\sigma}(\alpha_{num}, \alpha_{subj?}|\text{FORM, CON, MOR}) \propto p(\alpha_{num}, \alpha_{subj?})^{\sigma_{\alpha}}
                                                                           * p(\text{FORM}|\alpha_{num}, \alpha_{subj?})^{\sigma_{form}}
                                                                           * p(\text{CON}|\alpha_{num}, \alpha_{subj?})^{\sigma_{con}}
                                                                           * p(MOR|\alpha_{num},\alpha_{subj?})^{\sigma_{mor}}
```

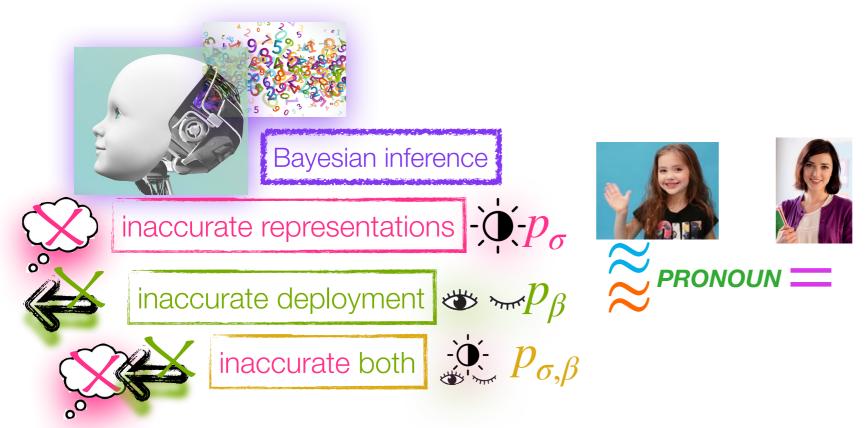
We allow 0.01 $\leq \sigma \leq$ 4 and 0 $\leq \beta \leq$ 1, and see which σ and β value combinations best predict child and adult pronoun interpretation behavior.





```
p_{\beta}(\alpha|\text{FORM}, \text{CON}, \text{MOR}, \alpha_{num}, \alpha_{subj?}) =
               (\beta_{form})(\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|FORM, CON, MOR, \alpha_{num}, \alpha_{subj?}) +
(\beta_{form})(\beta_{con})(\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM, CON, MOR, }\alpha_{num},\alpha_{subj?})+
                    (1 - \beta_{form})(\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha | CON, MOR, \alpha_{num}, \alpha_{subj?}) +
     (1 - \beta_{form})(\beta_{con})(\beta_{mor})(1 - \beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{CON}, \text{MOR}, \alpha_{num}, \alpha_{subj?}) +
                 (\beta_{form})(1-\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|FORM, MOR, \alpha_{num}, \alpha_{subj?}) +
  (\beta_{form})(1-\beta_{con})(\beta_{mor})(1-\beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{FORM}, \text{MOR}, \alpha_{num}, \alpha_{subj?}) +
                  (\beta_{form})(\beta_{con})(1-\beta_{mor})(\beta_{\alpha}) * p(\alpha|FORM, CON, \alpha_{num}, \alpha_{subj?}) +
    (\beta_{form})(\beta_{con})(1-\beta_{mor})(1-\beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{FORM, CON}, \alpha_{num}, \alpha_{subj?}) +
                      (1 - \beta_{form})(1 - \beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|MOR, \alpha_{num}, \alpha_{subj?}) +
        (1 - \beta_{form})(1 - \beta_{con})(\beta_{mor})(1 - \beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{MOR}, \alpha_{num}, \alpha_{subj?}) +
                    (\beta_{form})(1-\beta_{con})(1-\beta_{mor})(\beta_{\alpha}) * p(\alpha|\text{FORM}, \alpha_{num}, \alpha_{subj?}) +
     (\beta_{form})(1-\beta_{con})(1-\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM},\alpha_{num},\alpha_{subj?})+
                       (1 - \beta_{form})(\beta_{con})(1 - \beta_{mor})(\beta_{\alpha}) * p(\alpha | CON, \alpha_{num}, \alpha_{subj?}) +
        (1 - \beta_{form})(\beta_{con})(1 - \beta_{mor})(1 - \beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{CON}, \alpha_{num}, \alpha_{subj?}) +
                              (1 - \beta_{form})(1 - \beta_{con})(1 - \beta_{mor})(\beta_{\alpha}) * p(\alpha_{num}, \alpha_{subj?}) +
                                   (1 - \beta_{form})(1 - \beta_{con})(1 - \beta_{mor})(1 - \beta_{\alpha}) * p(\text{UNIF})
```

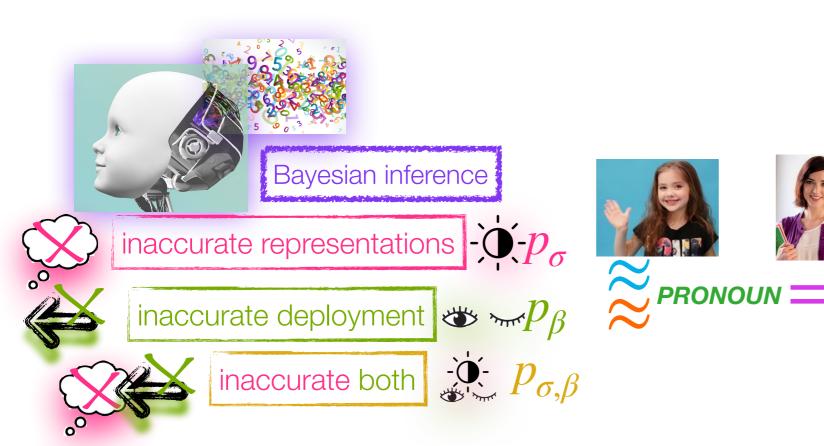
 P_{β}

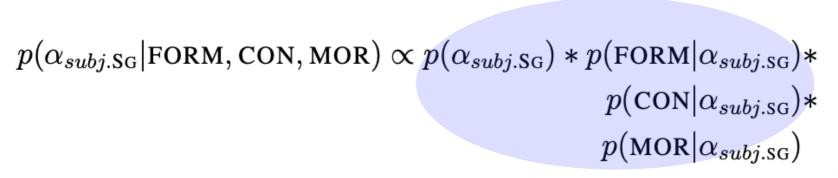




```
p(\alpha_{subj.SG}|\text{FORM}, \text{CON}, \text{MOR}) \propto p(\alpha_{subj.SG}) * p(\text{FORM}|\alpha_{subj.SG}) * 
p(\text{CON}|\alpha_{subj.SG}) * 
p(\text{MOR}|\alpha_{subj.SG})
```

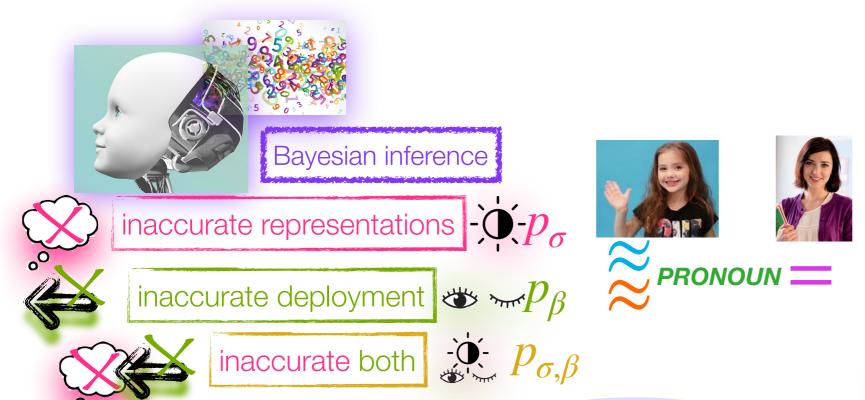
What input is the modeled listener using to represent the various information types?







54,757 utterances of Mexico City spontaneous child-directed speech to children 1;6-5;11 from the Schmitt-Miller corpus (Miller & Schmitt 2012).

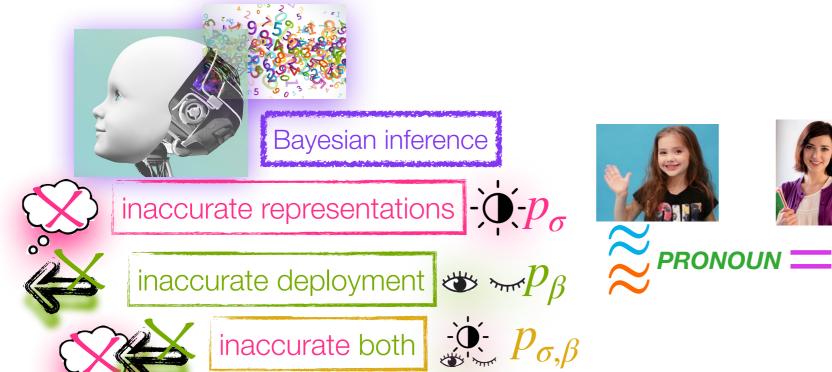


$$p(lpha_{subj.{
m SG}}|{
m FORM,CON,MOR}) \propto p(lpha_{subj.{
m SG}}) * p({
m FORM}|lpha_{subj.{
m SG}}) * \\ p({
m CON}|lpha_{subj.{
m SG}}) * \\ p({
m MOR}|lpha_{subj.{
m SG}})$$

		prior						
			$p(\mathtt{FORM} lpha)$		$p(\mathtt{CON} lpha)$		p(MOR lpha)	
antecedent type		$p(\alpha)$	ø	overt	después	porque	SG	PL
SUBJ	SG	0.362	0.938	0.062	0.324	0.676	0.998	0.002
	PL	0.071	0.984	0.016	0.750	0.250	0.005	0.995
¬SUBJ	SG	0.438	0.817	0.183	0.132	0.868	0.998	0.002
	PL	0.129	0.959	0.041	0.394	0.606	0.005	0.995



From this, we estimate the relevant priors and likelihoods.

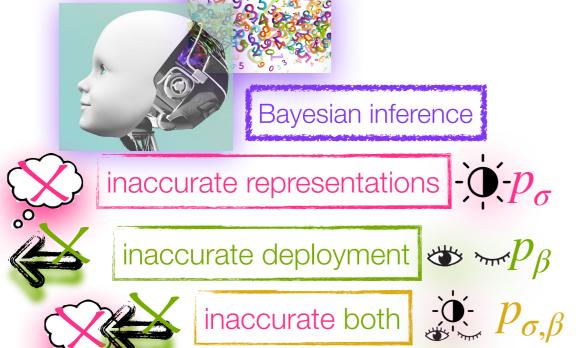


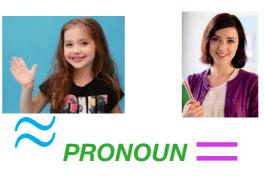
$$p(\alpha_{subj.SG}|{
m FORM,CON,MOR}) \propto p(\alpha_{subj.SG}) * p({
m FORM}|\alpha_{subj.SG}) * \\ p({
m CON}|\alpha_{subj.SG}) * \\ p({
m MOR}|\alpha_{subj.SG})$$

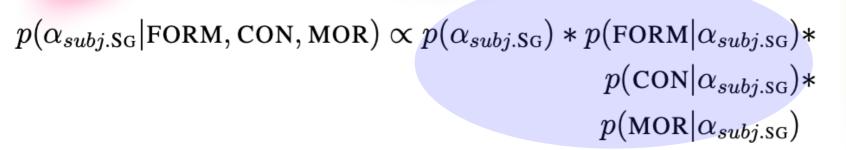
		prior	likelihoods					
		$p(\alpha)$	$p(\mathtt{FORM} lpha)$		$p(\operatorname{CON} lpha)$		$p(\mathtt{MOR} lpha)$	
antecedent type		$p(\alpha)$	ø	overt	después	porque	SG	PL
SUBJ	SG	0.362	0.938	0.062	0.324	0.676	0.998	0.002
	PL	0.071	0.984	0.016	0.750	0.250	0.005	0.995
¬SUBJ	SG	0.438	0.817	0.183	0.132	0.868	0.998	0.002
	PL	0.129	0.959	0.041	0.394	0.606	0.005	0.995



Singular antecedents generally occur more often.





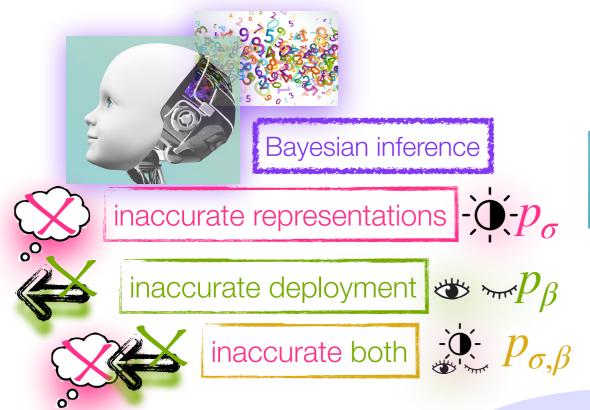


		prior						
		$p(\alpha)$	$p(\mathtt{FORM} lpha)$		$p(\operatorname{CON} \alpha)$		$p({ m MOR} lpha)$	
antecedent type		$p(\alpha)$	ø	overt	después	porque	SG	PL
SUBJ	SG	0.362	0.938	0.062	0.324	0.676	0.998	0.002
	PL	0.071	0.984	0.016	0.750	0.250	0.005	0.995
¬SUBJ	SG	0.438	0.817	0.183	0.132	0.868	0.998	0.002
	PL	0.129	0.959	0.041	0.394	0.606	0.005	0.995



The null pronoun form is generally used, though some antecedent types use it more often.

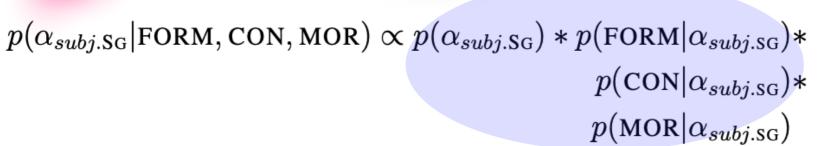








PRONOUN =



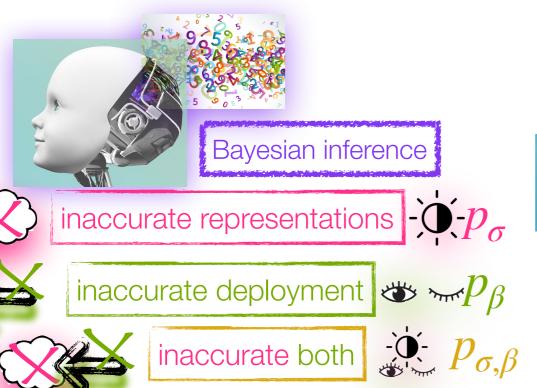
		prior	likelihoods					
		$p(\alpha)$	$p(extsf{FORM} lpha)$		$p(\operatorname{CON} lpha)$		$p(\text{MOR} \alpha)$	
antecedent type		$p(\alpha)$	ø	overt	después	porque	SG	PL
SUBJ	SG	0.362	0.938	0.062	0.324	0.676	0.998	0.002
	PL	0.071	0.984	0.016	0.750	0.250	0.005	0.995
¬SUBJ	SG	0.438	0.817	0.183	0.132	0.868	0.998	0.002
	PL	0.129	0.959	0.041	0.394	0.606	0.005	0.995



The connective *porque* is used more often for antecedents that aren't plural subjects, though how much more often varies.



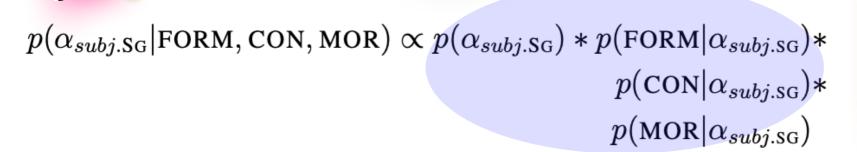








PRONOUN



		prior						
	$p(\alpha)$		$p(FORM \alpha)$		$p(\mathtt{CON} lpha)$		$p(\mathtt{MOR} lpha)$	
antecedent type		$P(\alpha)$	ø	overt	después	porque	SG	PL
SUBJ	SG	0.362	0.938	0.062	0.324	0.676	0.998	0.002
	PL	0.071	0.984	0.016	0.750	0.250	0.005	0.995
¬SUBJ	SG	0.438	0.817	0.183	0.132	0.868	0.998	0.002
	PL	0.129	0.959	0.041	0.394	0.606	0.005	0.995

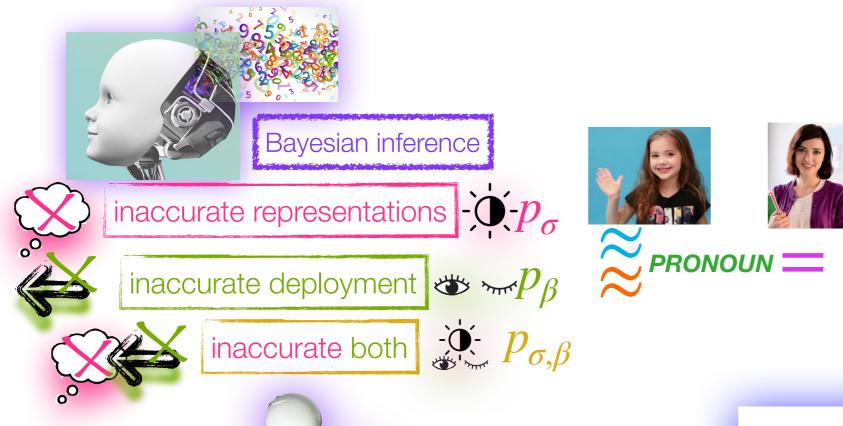


Agreement morphology is nearly categorical, with a very strong preference for matching morphology.

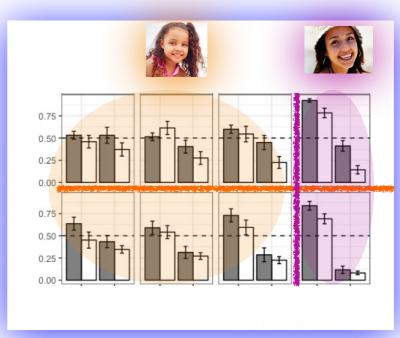


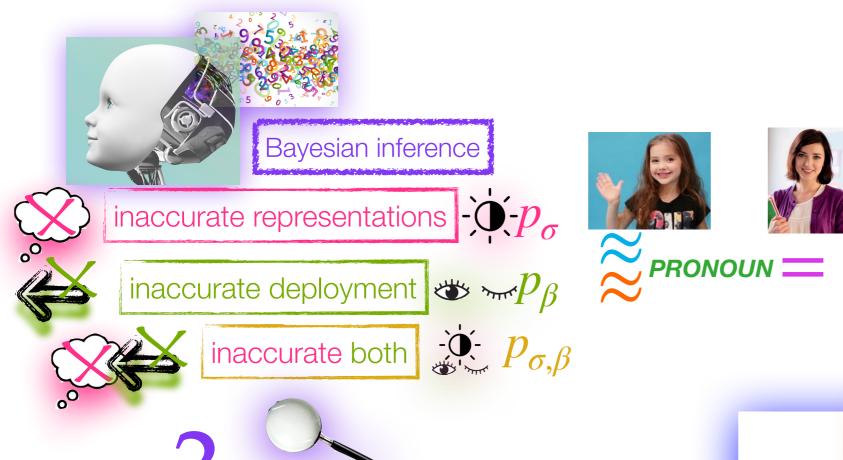




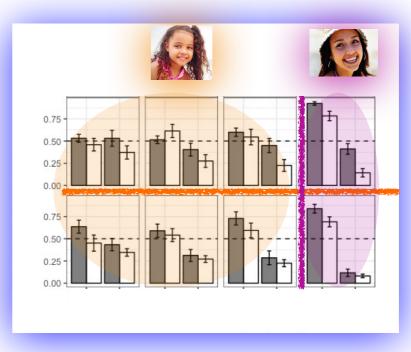


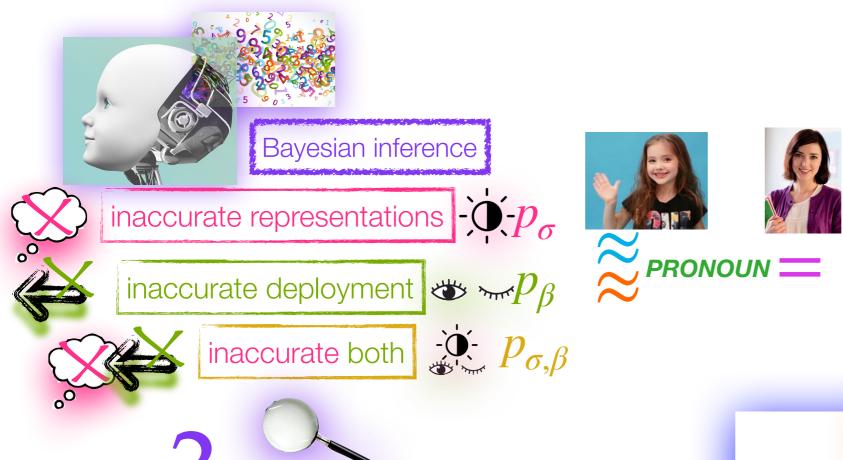
The plan, part 2: Use the computational cognitive model to identify the specific differences leading to child and adult pronoun interpretation in context.



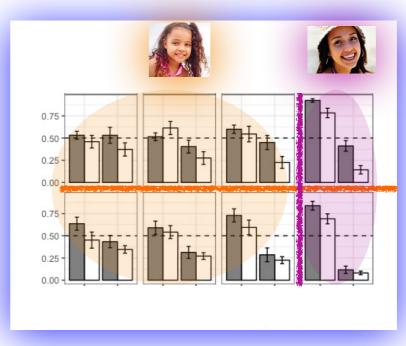


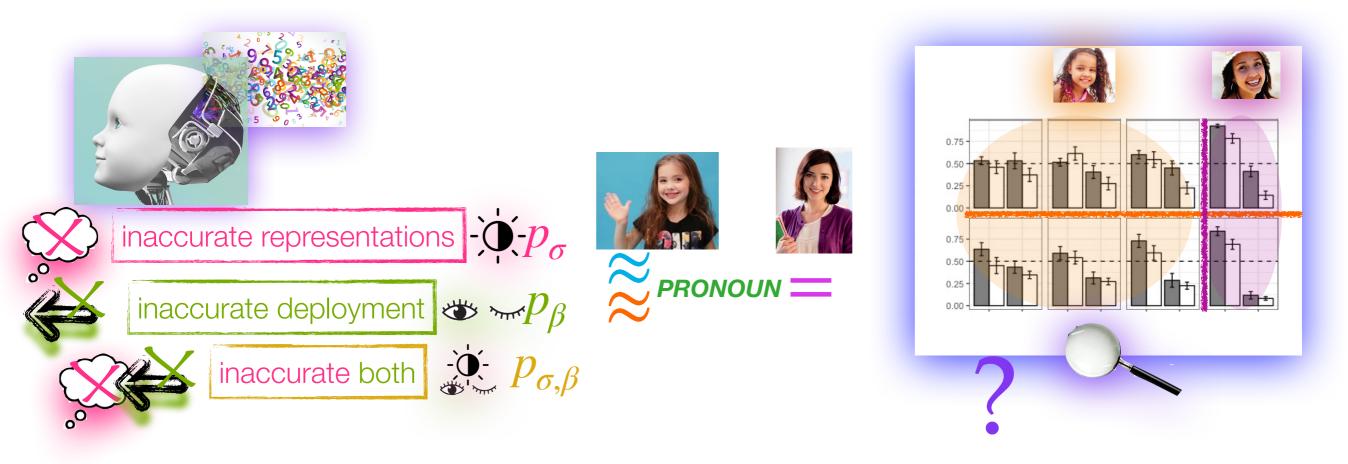
Which modeled listener variant best matches the observed pronoun interpretation behavior?



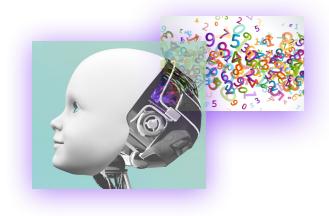


Important: Model variants with more parameters have an easier time fitting the data because they have more degrees of freedom.





```
baseline: accurate representations and deployment p(\alpha_{subj.SG}|\text{FORM}, \text{CON}, \text{MOR}) \propto p(\alpha_{subj.SG}) * p(\text{FORM}|\alpha_{subj.SG}) * 0 free parameters p(\text{CON}|\alpha_{subj.SG}) * p(\text{MOR}|\alpha_{subj.SG})
```







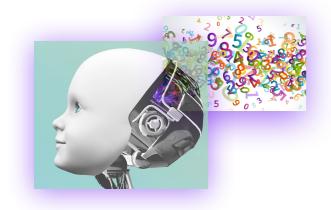


baseline: accurate representations and deployment

p o free parameters



4 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor}









inaccurate both $p_{\sigma,\beta}$



baseline: accurate representations and deployment

0 free parameters





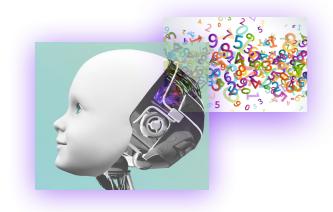
inaccurate representations - $\dot{\mathcal{D}}$ - $\dot{\mathcal{D}}$ 4 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor}



inaccurate deployment $woheadrightarrow \mathcal{P}_{eta}$



4 free parameters: β_{α} , β_{form} , β_{con} , β_{mor}







baseline: accurate representations and deployment

0 free parameters





inaccurate representations - p_{σ} 4 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor}



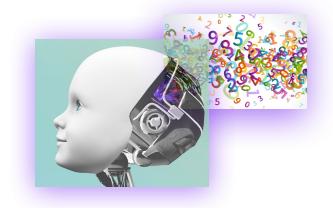
inaccurate deployment $\begin{cases} \begin{cases} \begin$



inaccurate both $p_{\sigma,\beta}$



8 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor} , β_{α} , β_{form} , β_{con} , β_{mor}





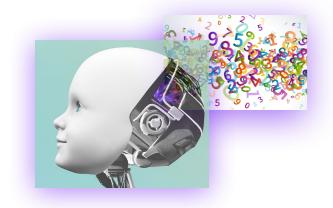


Forsythe & Pearl 2019, in prep



We want variants with more parameters to have a substantially better fit in order to favor them over variants with fewer parameters.

baseline: accurate representations and deployment P 0 free parameters inaccurate representations P_{σ} 4 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor} inaccurate deployment P_{σ} 4 free parameters: P_{σ} , P_{σ} ,

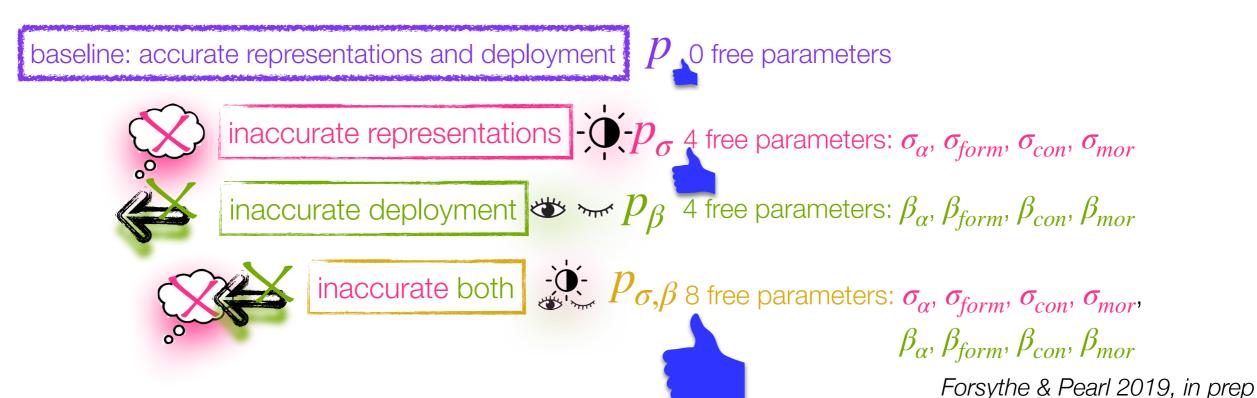


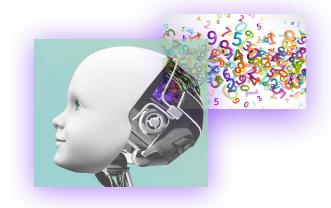






The Bayesian Information Criterion (BIC) is one way to quantify this preference (Schwarz 1978).









BIC = # parameters $\cdot \log(|data|) - 2 \cdot \log(model fit)$

 $0 \le BIC \le \infty$ (closer to 0 is better)

baseline: accurate representations and deployment

 p_{\searrow} 0 free parameters



inaccurate representations $-\dot{\mathcal{D}}-p_{\sigma}$ 4 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor}





inaccurate deployment $\begin{cases} \begin{cases} \begin$



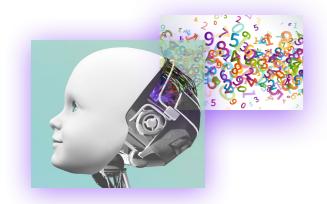




inaccurate both $p_{\sigma,\beta}$ 8 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor} ,

 β_{α} , β_{form} , β_{con} , β_{mor}

Forsythe & Pearl 2019, in prep



more parameters = higher score, closer to 0 is better





BIC = # parameters $\cdot \log(|data|) - 2 \cdot \log(model fit)$

 $0 \le BIC \le \infty$ (closer to 0 is better)

baseline: accurate representations and deployment

p of free parameters



inaccurate representations $-\dot{\mathcal{D}}-p_{\sigma}$ 4 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor}





inaccurate deployment $\begin{tabular}{c} \begin{tabular}{c} \begin{t$



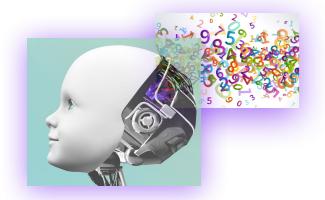




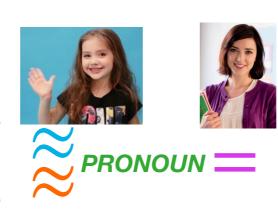
inaccurate both $p_{\sigma,\beta}$ 8 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor}

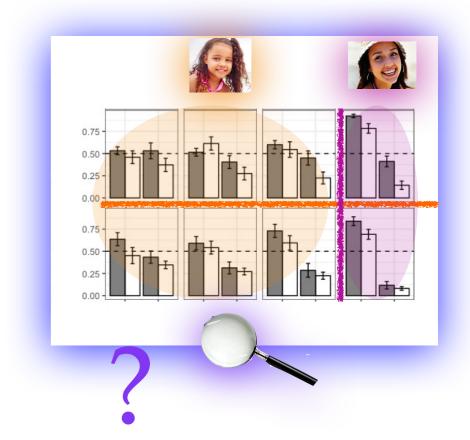
 β_{α} , β_{form} , β_{con} , β_{mor}

Forsythe & Pearl 2019, in prep



model fit = likelihood of data, given model with best-fitting parameter values





BIC = # parameters $\cdot \log(|data|) - 2 \cdot \log(model fit)$

 $0 \le BIC \le \infty$ (closer to 0 is better)

baseline: accurate representations and deployment

 $p_{
ightharpoonup}$ 0 free parameters



inaccurate representations $-\dot{\mathcal{D}}-p_{\sigma}$ 4 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor}







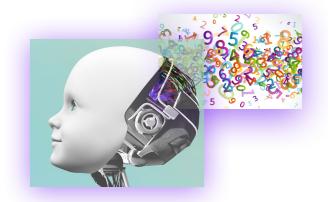




inaccurate both $p_{\sigma,\beta}$ 8 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor}

 β_{α} , β_{form} , β_{con} , β_{mor}

Forsythe & Pearl 2019, in prep



 $-\infty \le \log(\text{likelihood}) \le 0$, closer to 0 is better



BIC = # parameters $\cdot \log(|data|) - 2 \cdot \log(model fit)$

 $0 \le BIC \le \infty$ (closer to 0 is better)

baseline: accurate representations and deployment

p of free parameters



inaccurate representations $-\dot{\mathcal{D}}-p_{\sigma}$ 4 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor}





inaccurate deployment $\begin{tabular}{c} \begin{tabular}{c} \begin{t$



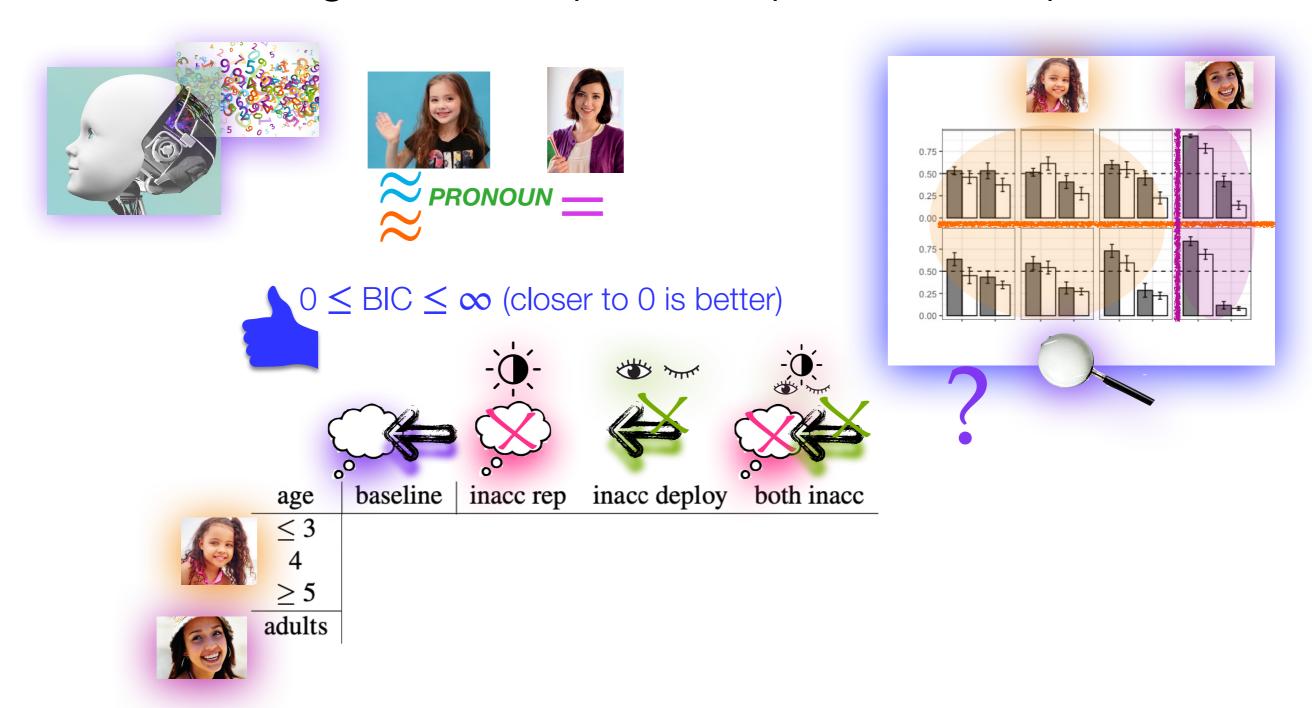


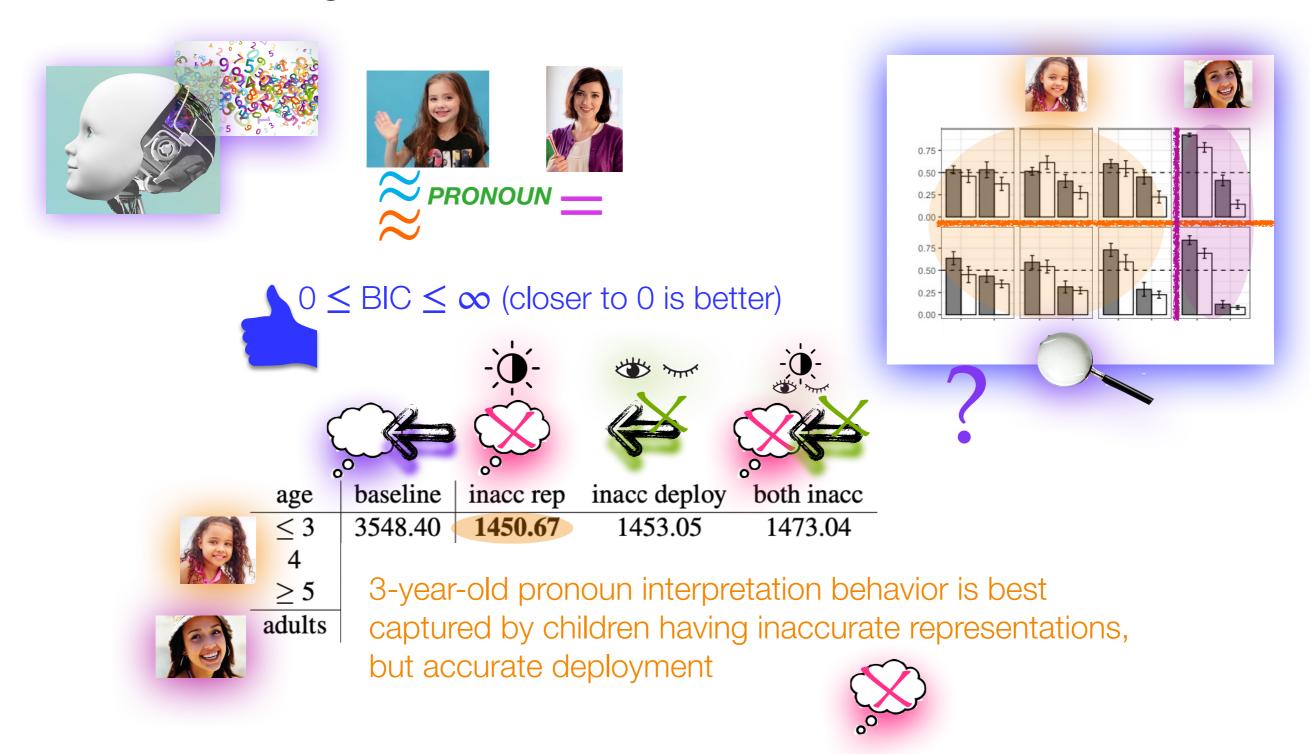


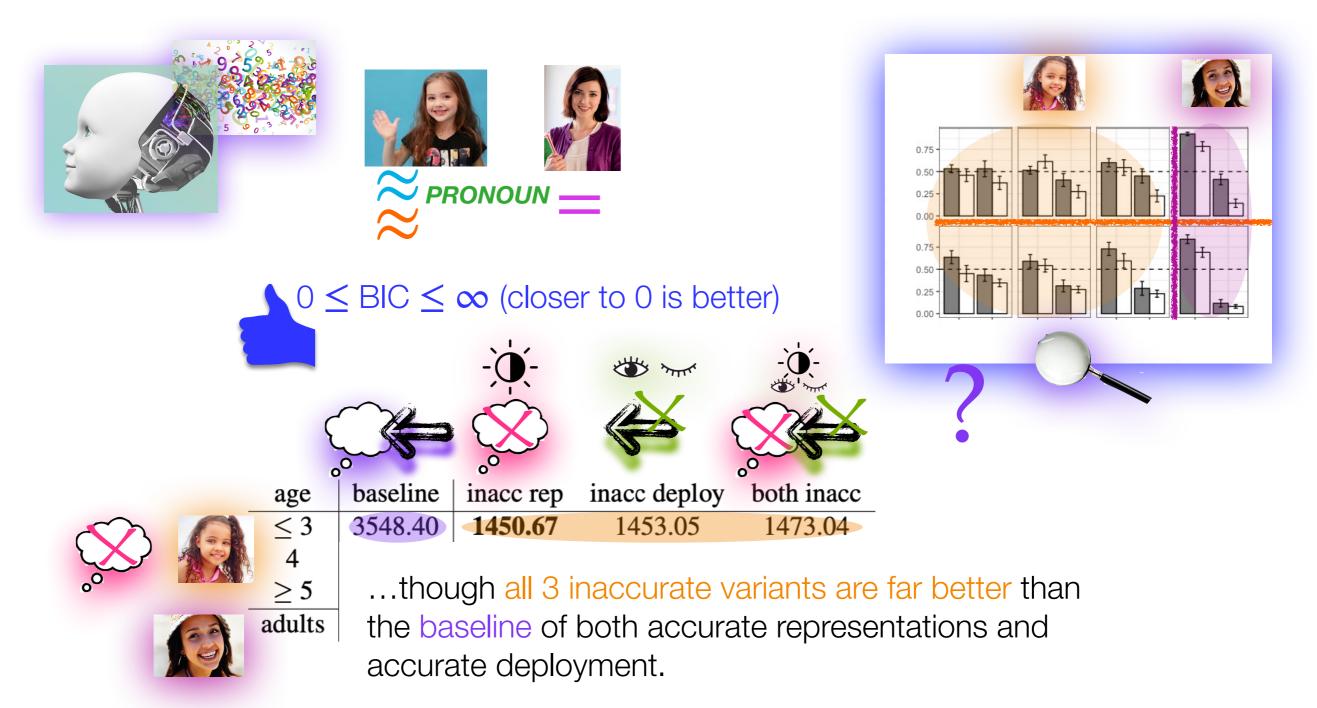
inaccurate both $p_{\sigma,\beta}$ 8 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor}

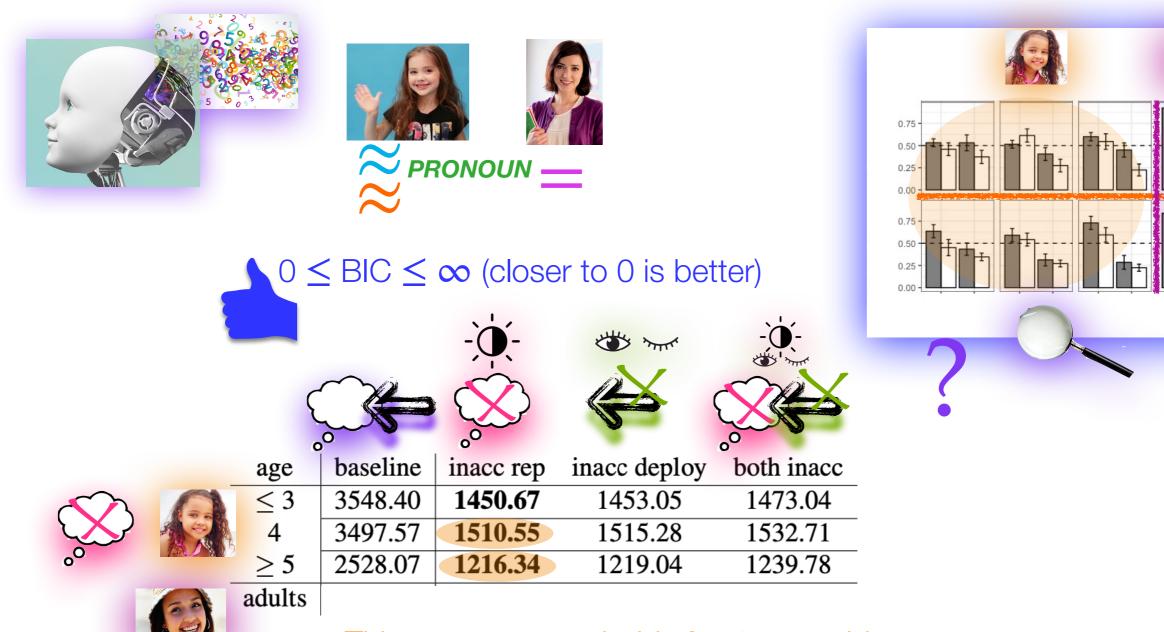
 β_{α} , β_{form} , β_{con} , β_{mor}

Forsythe & Pearl 2019, in prep

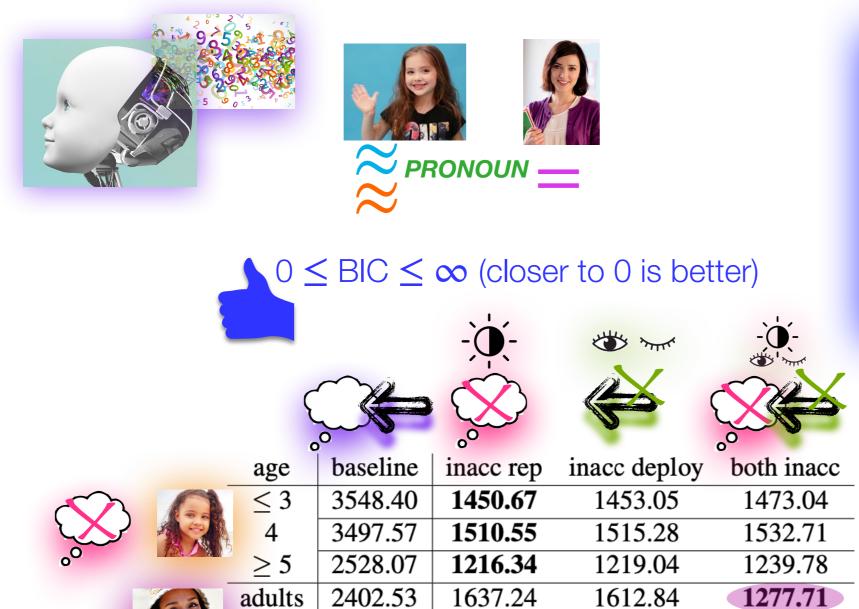








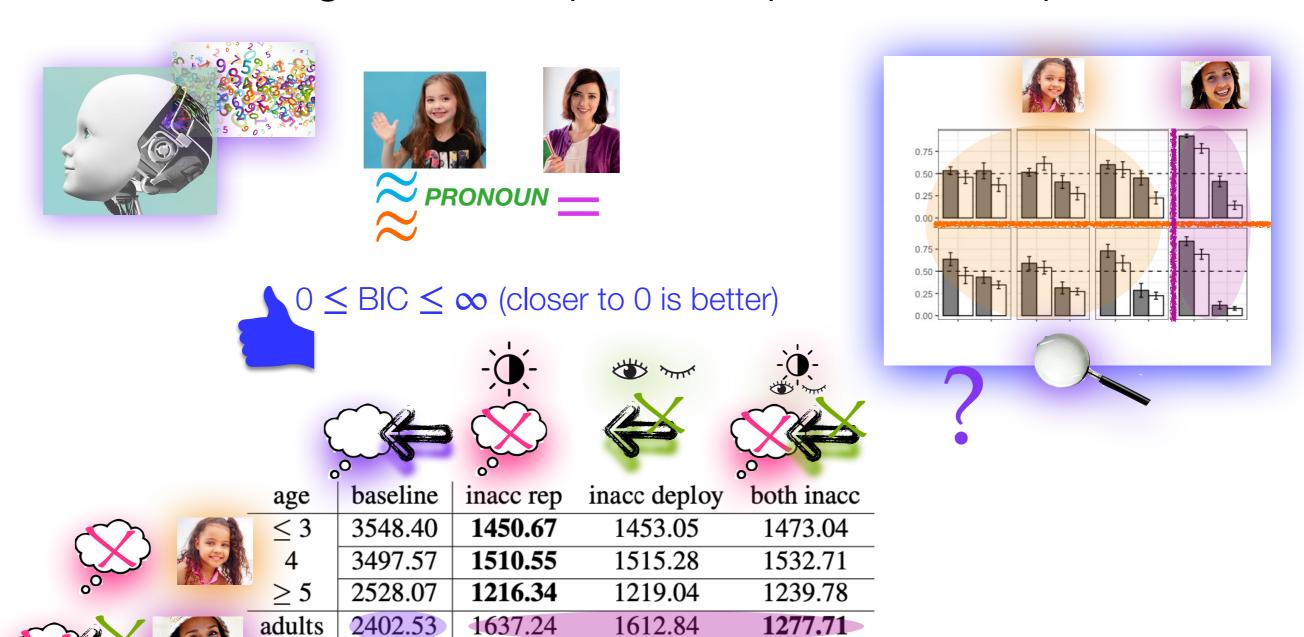
This same pattern holds for 4-year-olds and 5-year-olds, too.





Adult pronoun interpretation behavior is best captured by the model that has both inaccurate representations and inaccurate deployment.



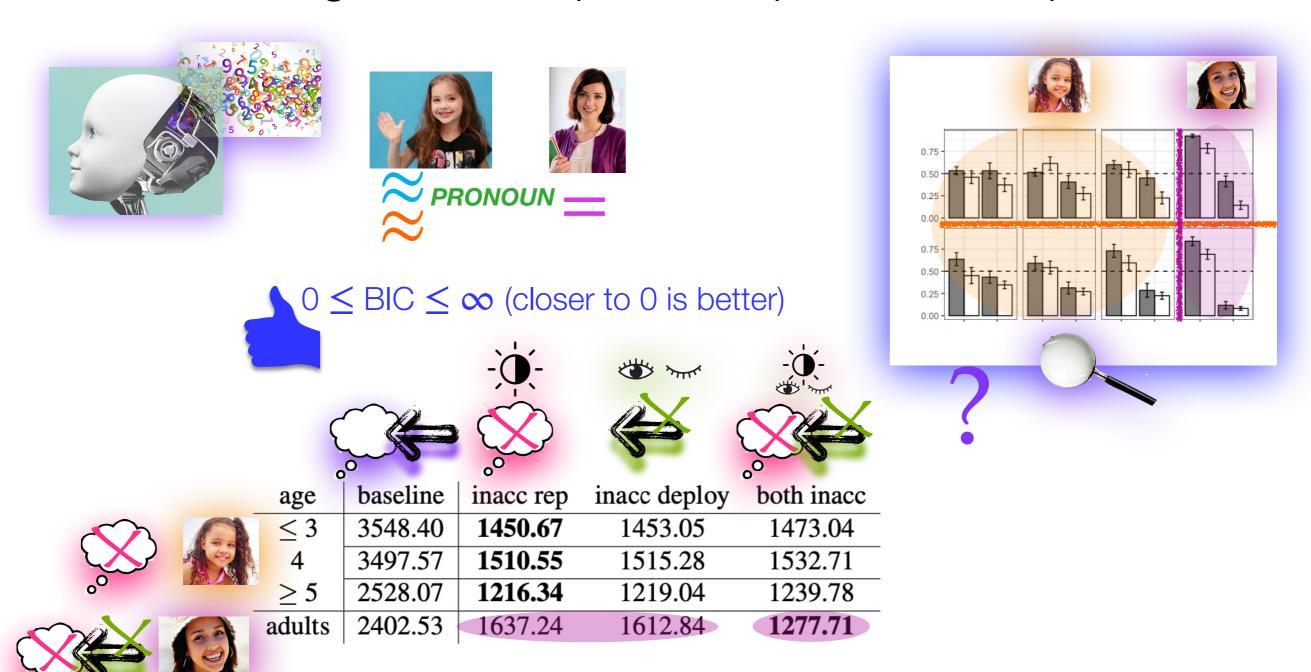


1612.84

1637.24

Like the child models, all inaccurate variants are far better than the baseline that has both accurate representations and accurate deployment.

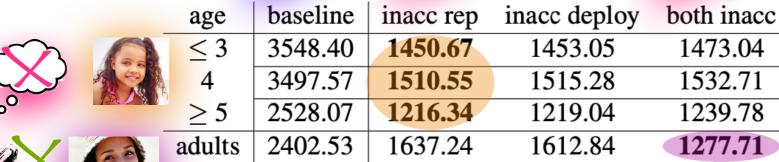
1277.71



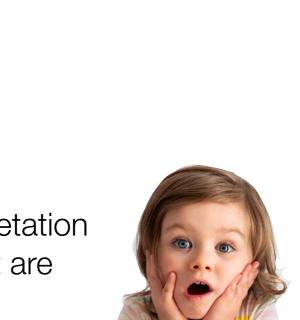
...but within the inaccurate variants, the both inaccurate variant is much better than the other two.

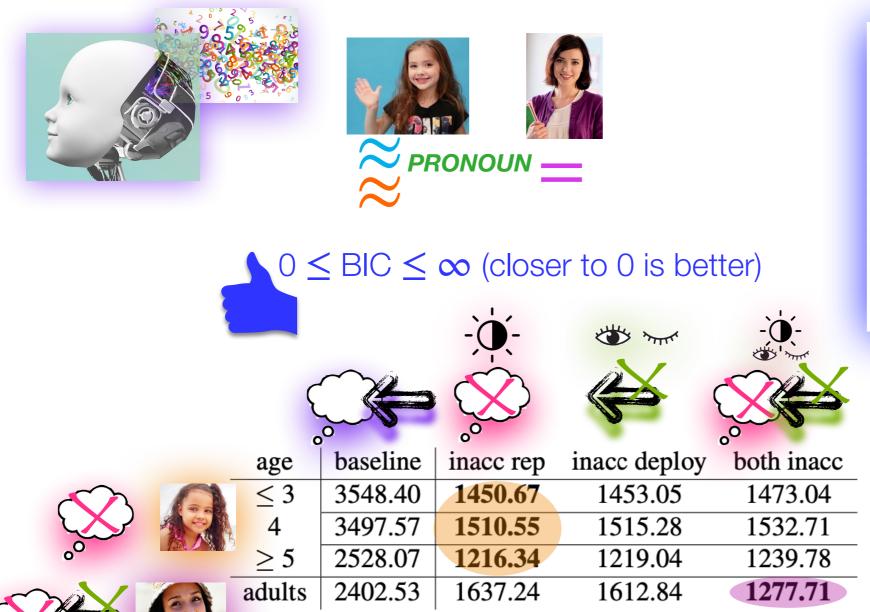






Takeaway: Both child and adult pronoun interpretation behavior are captured by modeled listeners that are inaccurate in some way.







Becoming adult-like doesn't mean becoming more accurate!

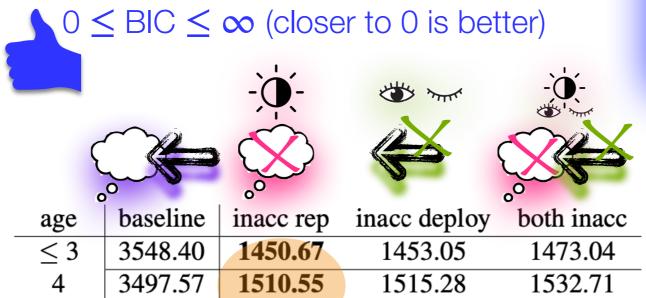




2528.07

2402.53

adults



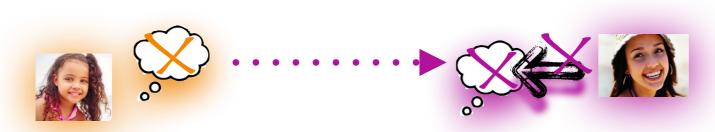
But what does it mean? It means learning to become inaccurate in adult-like ways.

1219.04

1612.84

1239.78

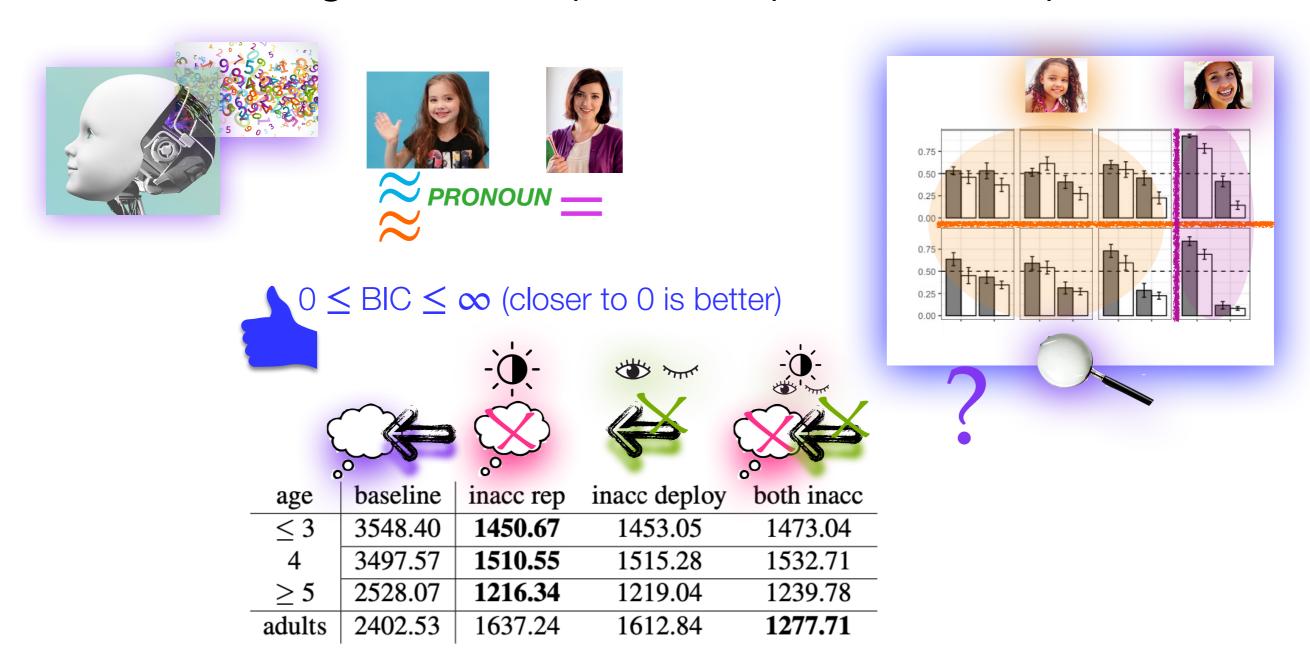
1277.71



1216.34

1637.24





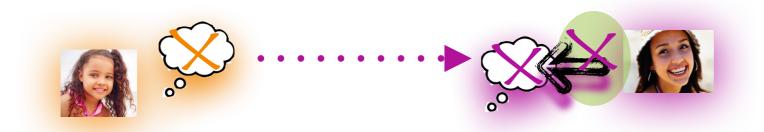
We know that adults inaccurately deploy their representations.

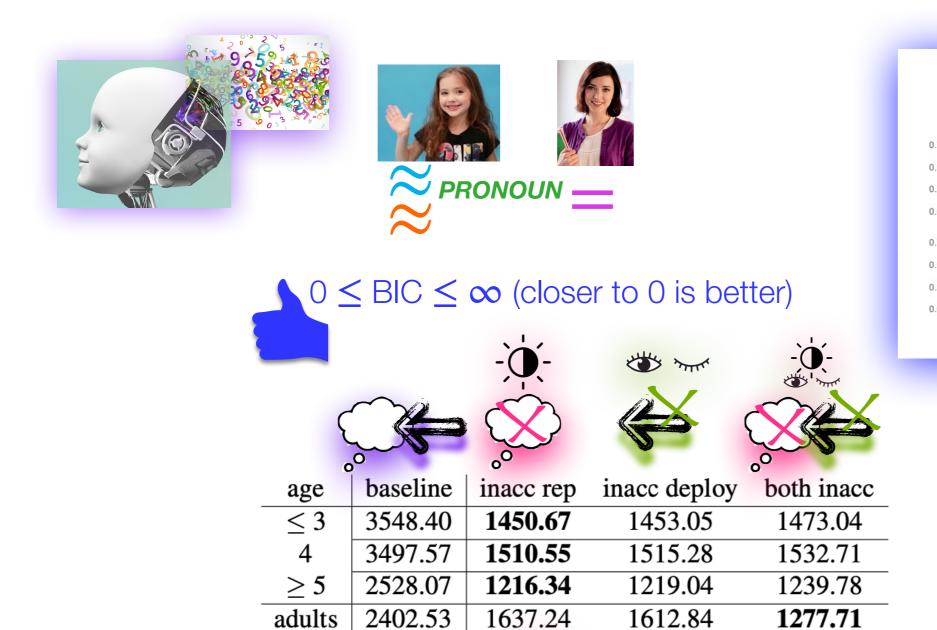




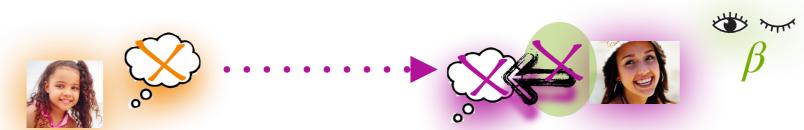
So we can look at the β values to see which information is being ignored and how often it's being ignored.

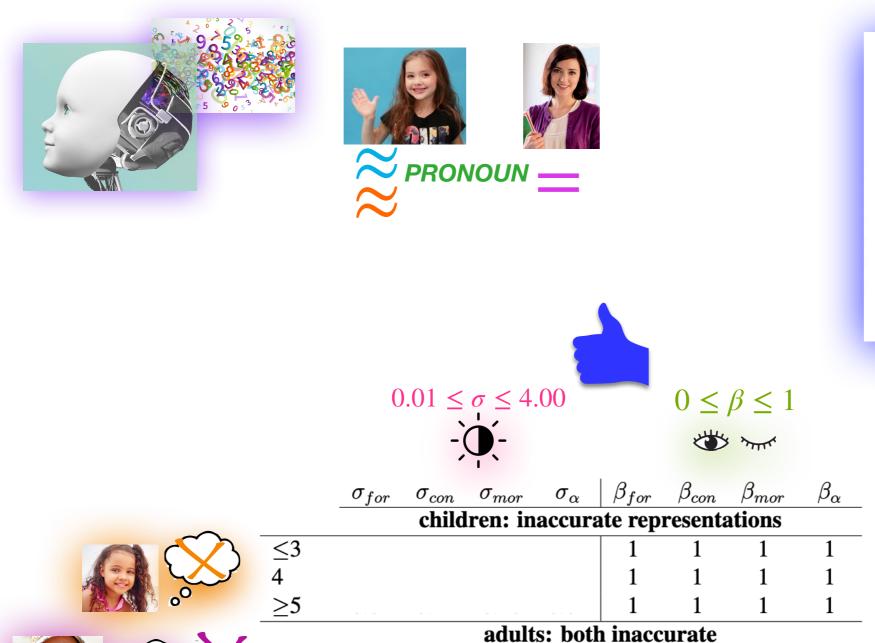








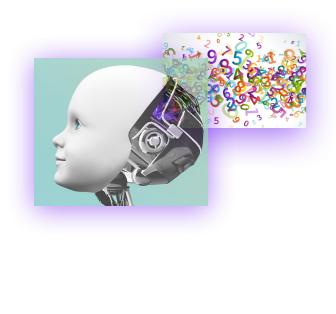




adults

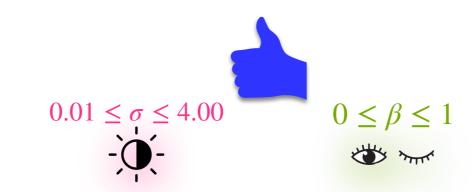


All children are best represented by accurate deployment, which is equivalent to always paying attention to all information (β =1).



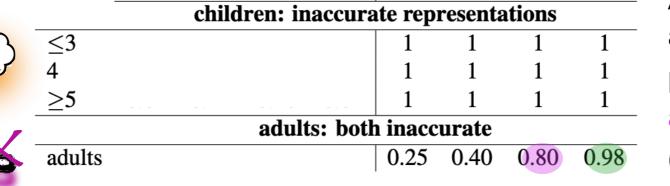


 σ_{con} σ_{mor}



 β_{for} β_{con} β_{mor}

 eta_{lpha}



 σ_{lpha}



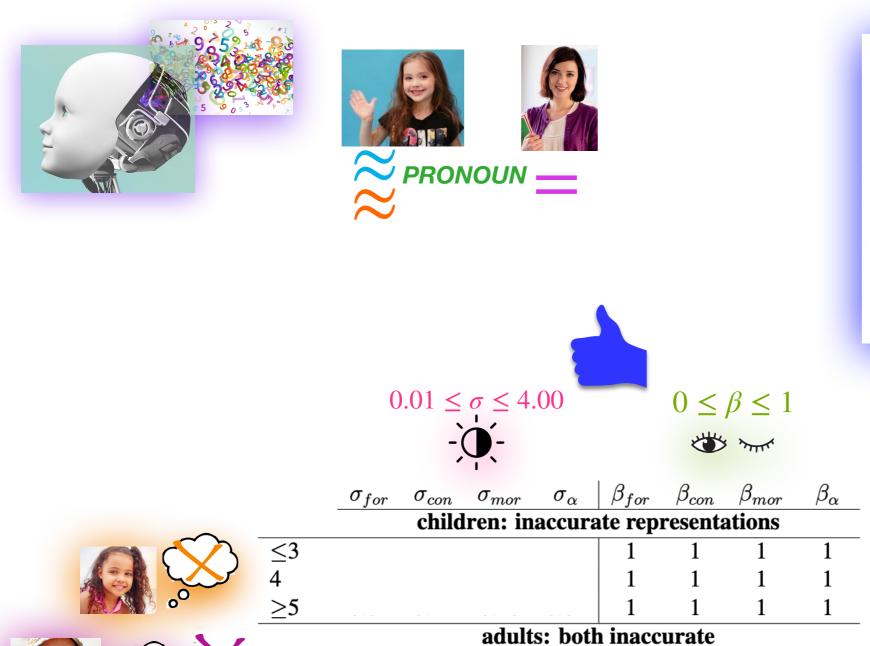
Adults pay much more attention to the prior over possible antecedents (β_{α}) and agreement morphology (β_{mor}) ...

0.80

0.25

0.40

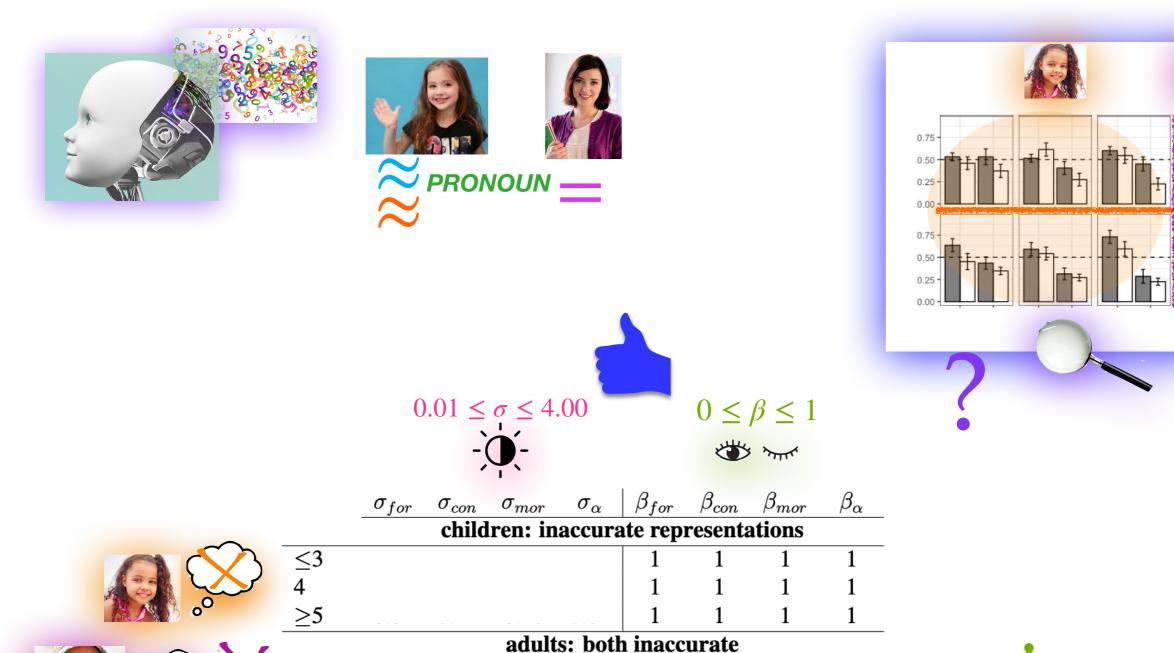
0.98



adults



...and much less attention to the pronoun form (β_{α}) and the connective (β_{con}) .



Becoming adult-like:

adults

Learn how much to ignore certain information

(the pronoun form and connective more often, sometimes the agreement morphology, and occasionally even the prior).

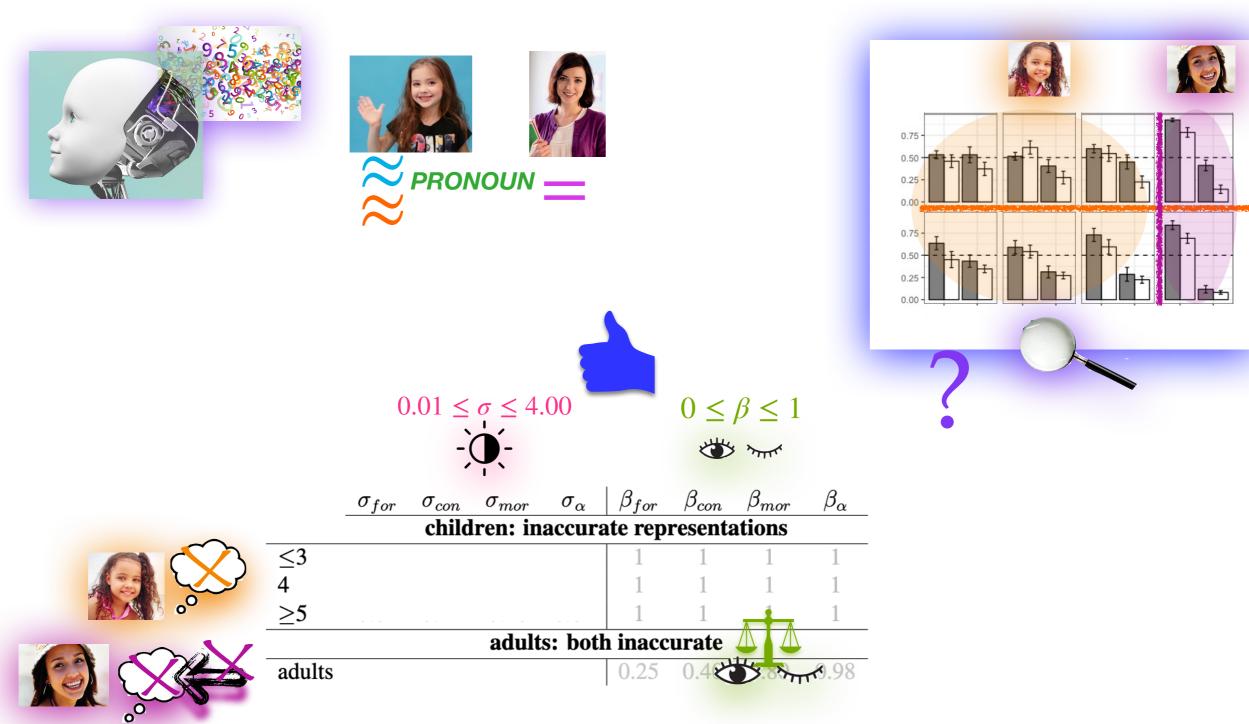
0.40

0.25

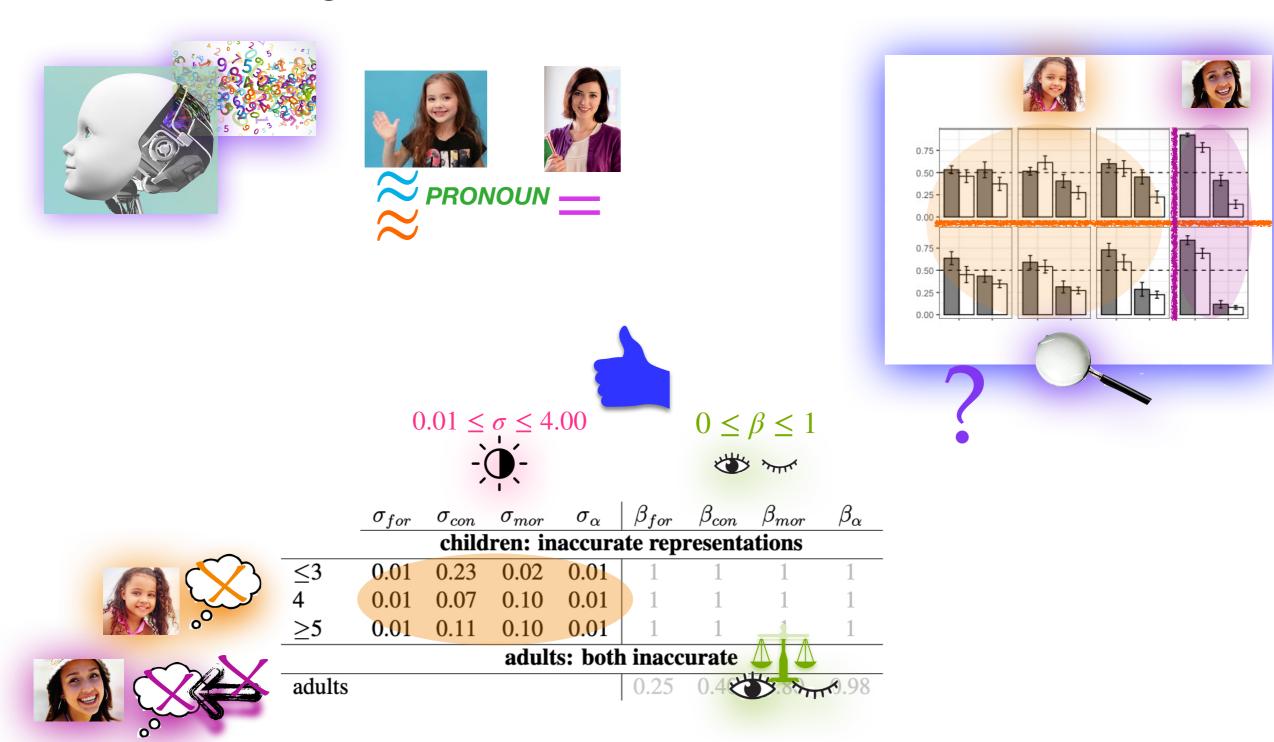
0.80

0.98



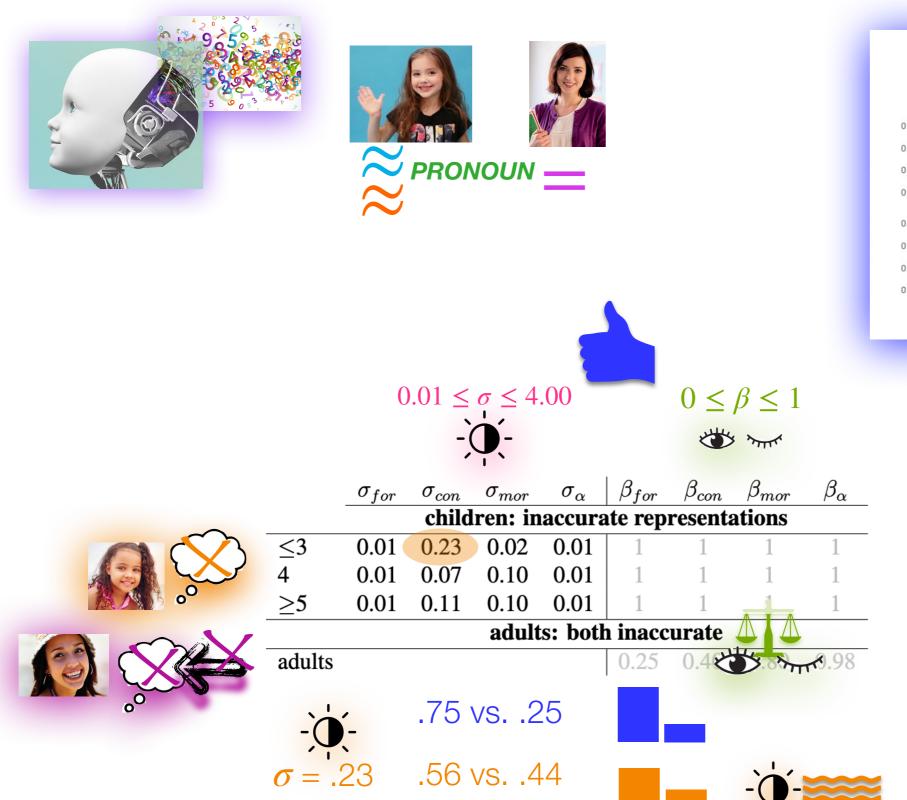


What about the representations? Both child and adult representations are inaccurate, but *how* are they inaccurate?

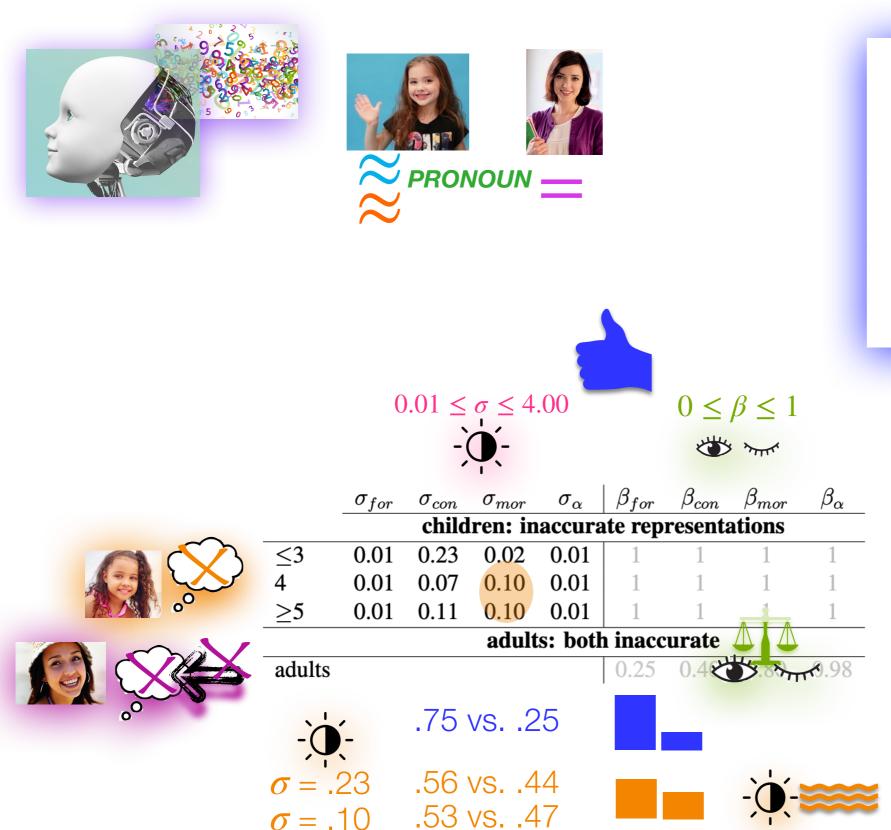


Child representations, with σ values often much < 1, seem to smooth away probability differences, effectively turning down the contrast.

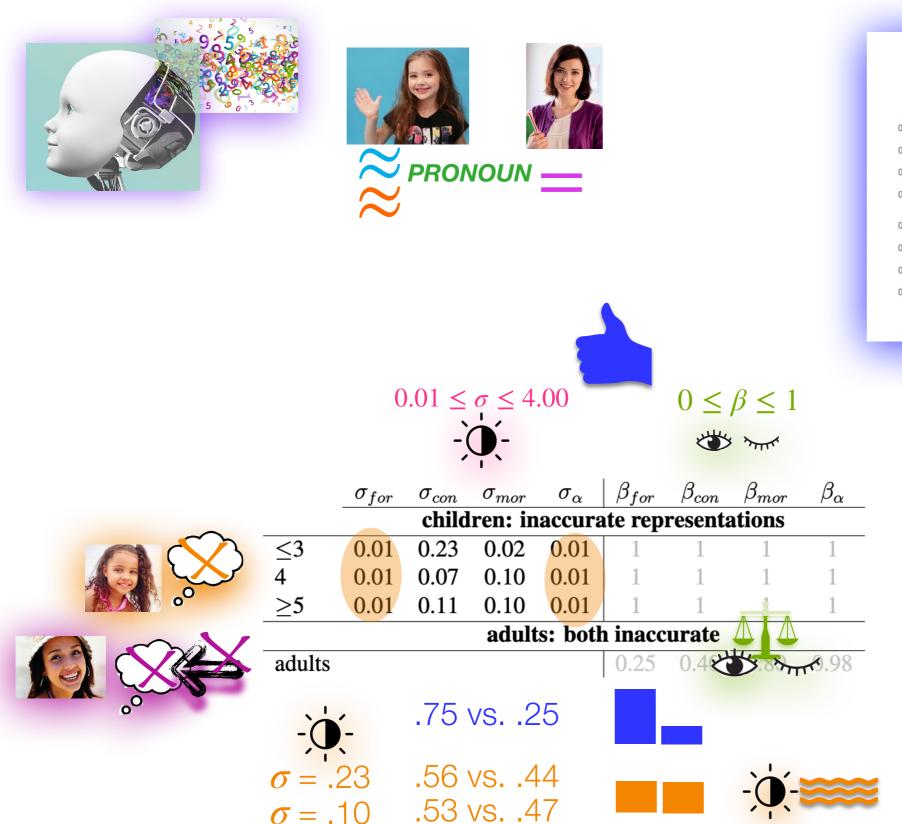












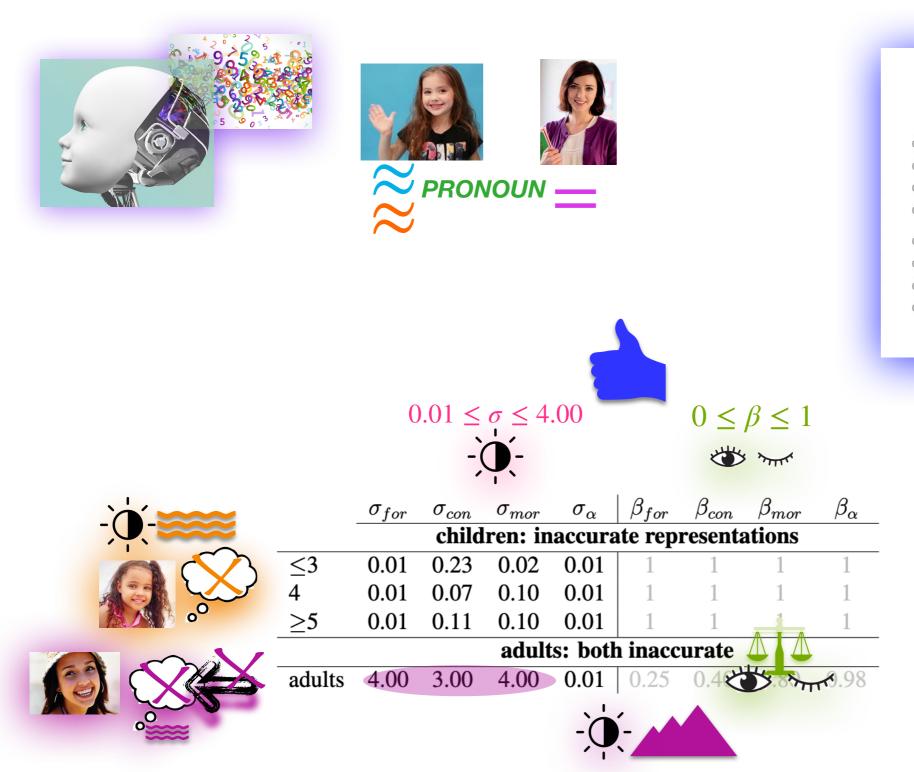
.502 vs. .498

 $\sigma = .01$

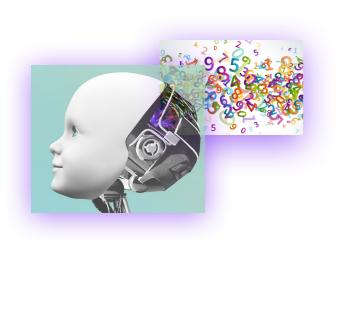




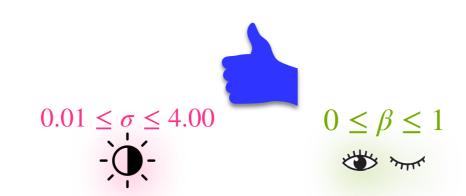
Adults do this same extreme smoothing for the prior information.



But adults substantially sharpen all the other information ($\sigma = 3.00, 4.00$)

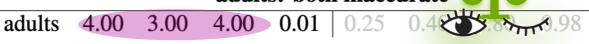






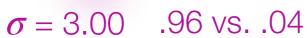


	σ_{for}	σ_{con}	σ_{mor}	σ_{lpha}	ρ_{for}	ρ_{con}	$ ho_{mor}$	$ ho_{lpha}$				
children: inaccurate representations												
<u>≤3</u>	0.01	0.23	0.02	0.01	1	1	1	1				
4	0.01	0.07	0.10	0.01	1	1	1	1				
≥5	0.01	0.11	0.10	0.01	1	1	1	1				
adults: both inaccurate 4												



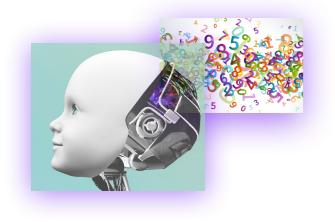


.75 vs. .25

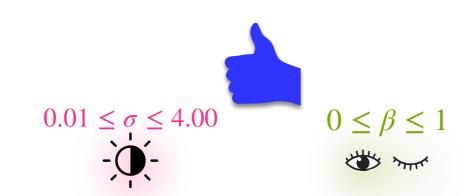














	o_{for}	o_{con}	o_{mor}	o_{α}	$\mid \rho_{for} \mid$	$ u_{con}$	$ ho_{mor}$	$ ho_{lpha}$					
children: inaccurate representations													
<u>≤3</u>	0.01	0.23	0.02	0.01	1	1	1	1					
4	0.01	0.07	0.10	0.01	1	1	1	1					
≥5	0.01	0.11	0.10	0.01	1	1	1	1					
adults: both inaccurate \triangle													



adults 4.00 3.00 4.00 0.01 0.25 0



.75 vs. .25

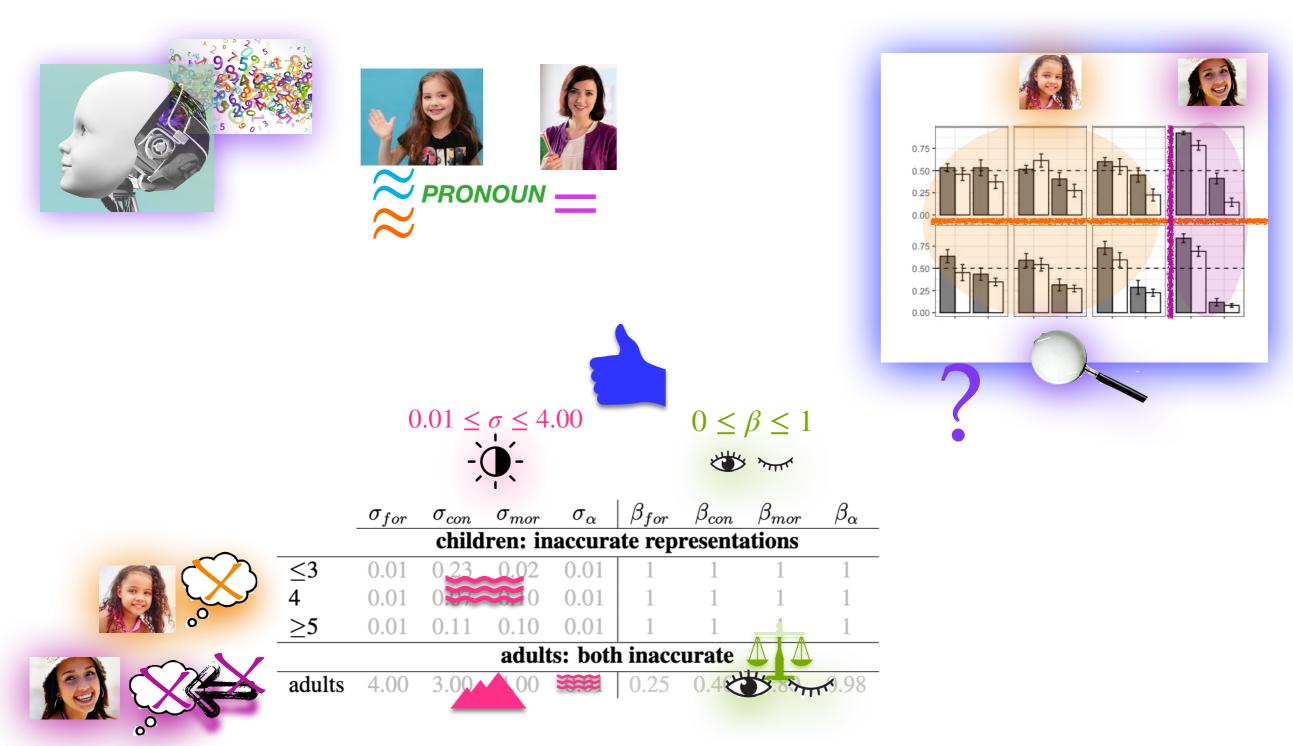


.96 vs. .04

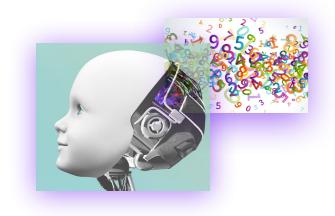








Becoming adult-like: Children need to learn to sharpen certain perceived representations, rather than smooth away probability differences.

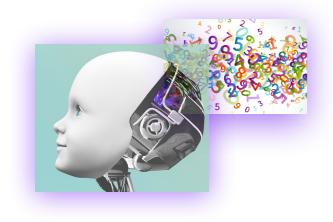




Takeaway:

To generate adult-like pronoun interpretation behavior in context, children need to change both how they represent relevant information from their input and how they deploy those representations.

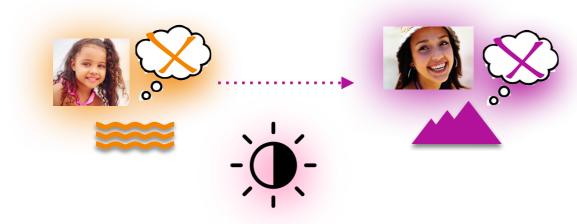




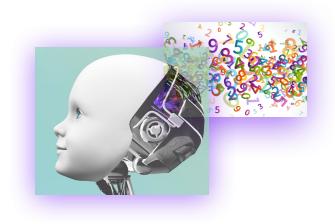


Takeaway:

Child representations are overly smooth — to become adult-like, many representations must sharpen perceived probability differences.



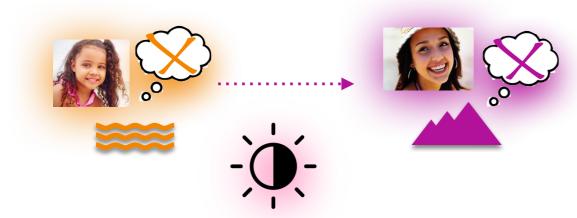






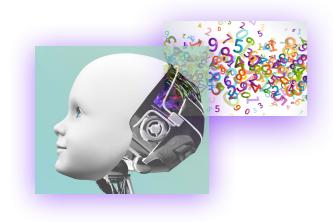
Takeaway:

Child representations are overly smooth — to become adult-like, many representations must sharpen perceived probability differences.





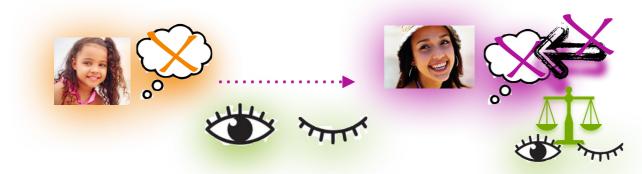
Important: Being adult-like doesn't mean being accurate!

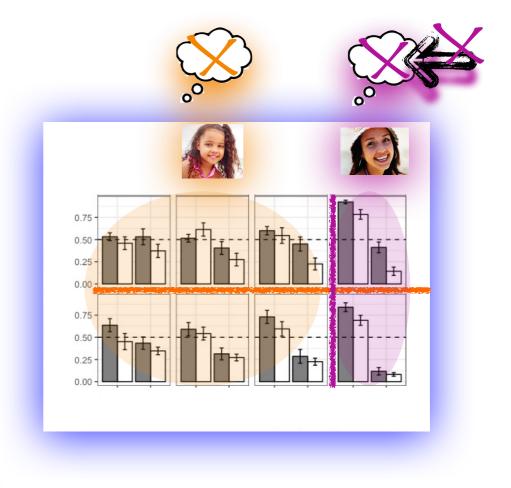




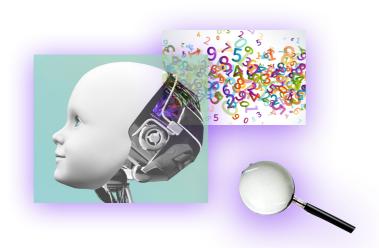
Takeaway:

Adult deployment is selective — to become adult-like, children must learn how to selectively ignore information like adults.

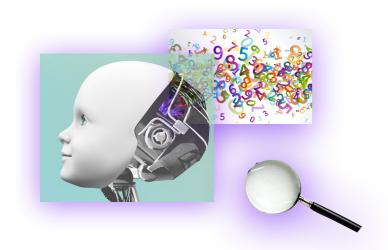


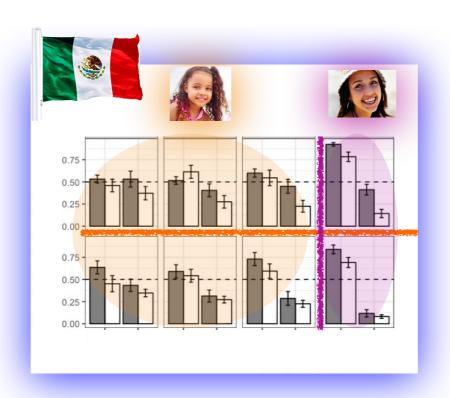


Important: Being adult-like doesn't mean being accurate!

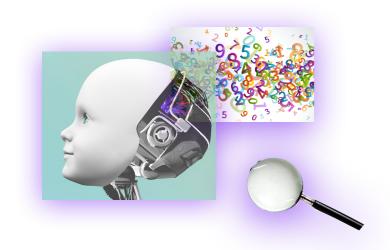


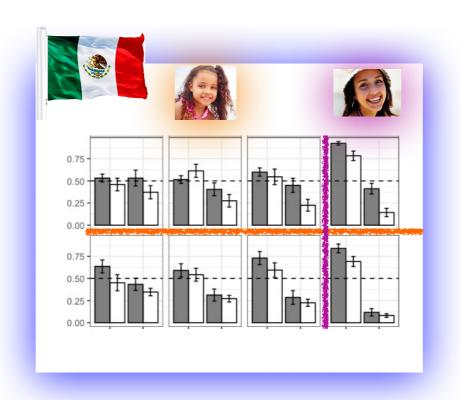
More generally, this case study demonstrates how we can use computational cognitive modeling...



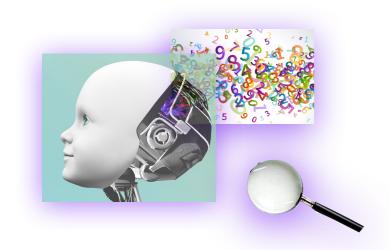


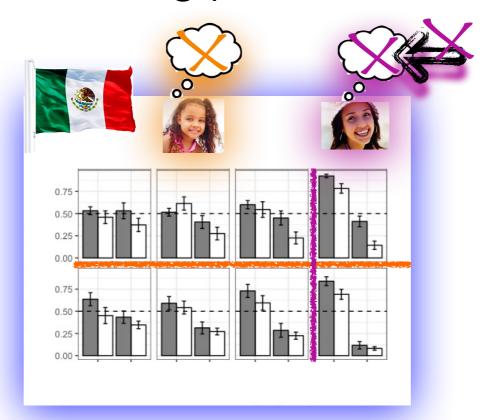
More generally, this case study demonstrates how we can use computational cognitive modeling, grounded in empirical data...





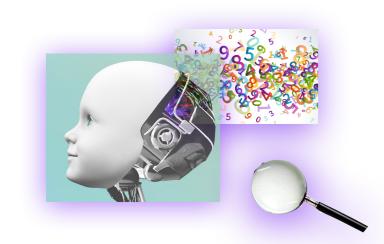
More generally, this case study demonstrates how we can use computational cognitive modeling, grounded in empirical data, to better understand how children and adults...





More generally, this case study demonstrates how we can use computational cognitive modeling, grounded in empirical data, to better understand how children and adults can solve complex linguistic tasks (like interpreting pronouns in a context with multiple, potentially conflicting, cues).

PRONOUN —









This helps us better understand what children need to do to become adult-like (and it seems to be about learning the right helpful biases).

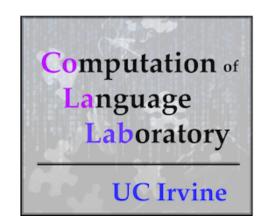


Thank you!

Hannah Forsythe



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