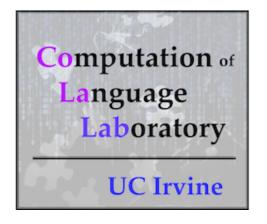
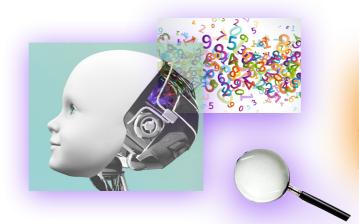
How children are and aren't like adults when interpreting pronouns: A computational cognitive modeling investigation

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Oct 8, 2021 **UCLA**

Linguistics Colloquium





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 in this context.





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.



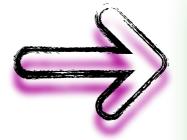


PRONOUN

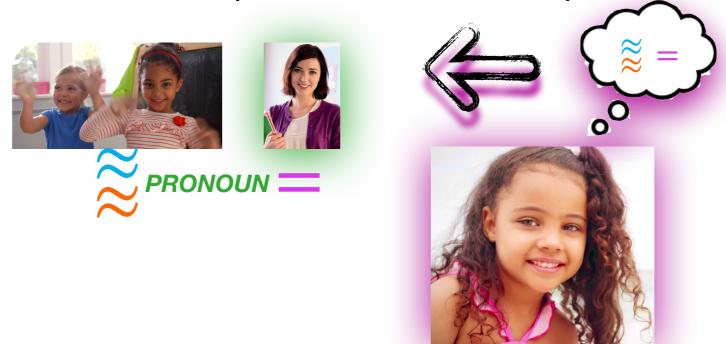


Second, children need adult-like ability to deploy that knowledge in real time.









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!

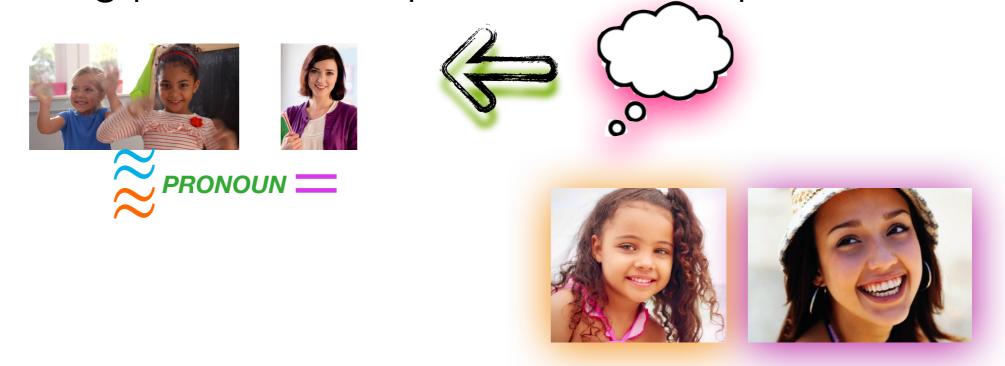




How do we tell what the differences are between child and adult pronoun interpretation? When we understand this better, we'll understand what children need to do to become adults.





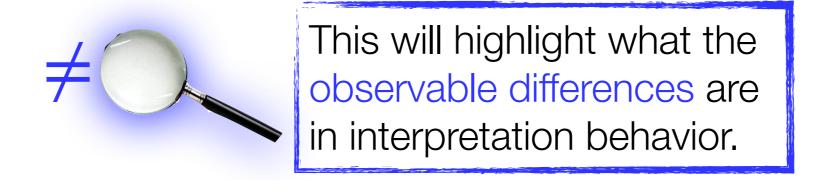


The plan, part 1: Get some empirical data on how children and adults interpret the same pronoun in a context where multiple cues are available.

Case study: Mexican Spanish



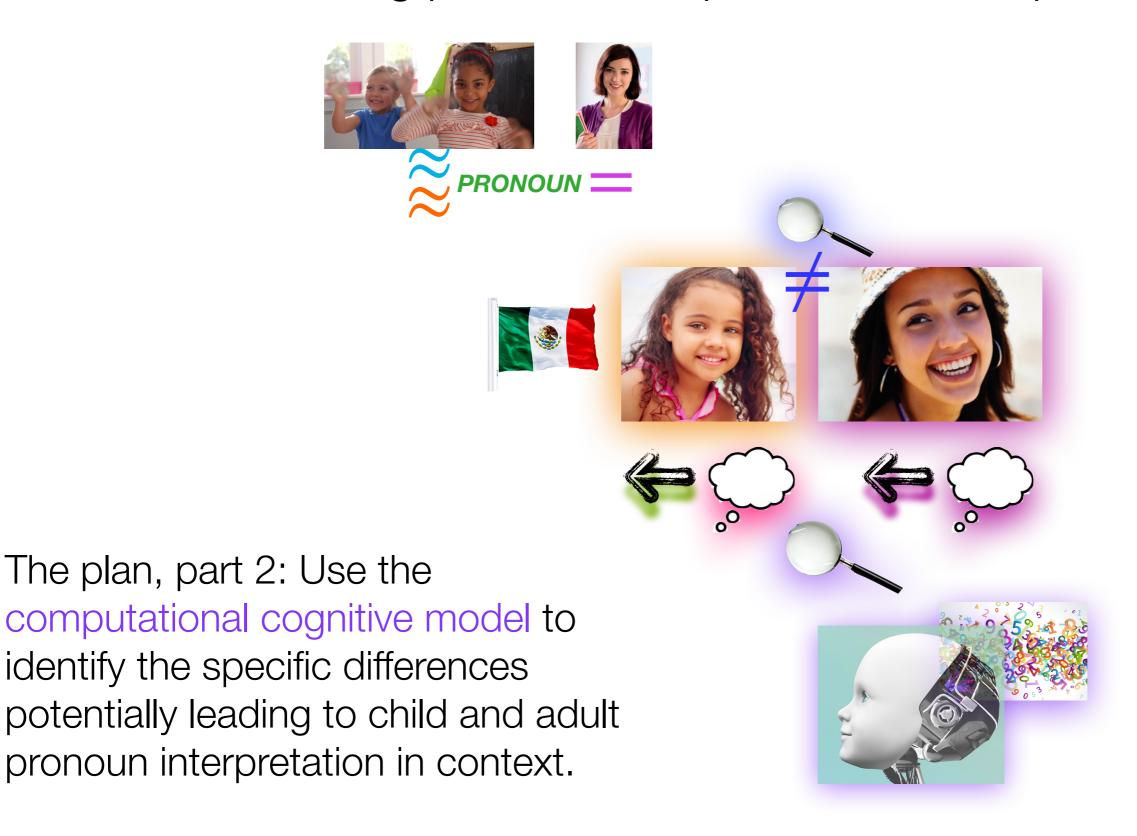
The plan, part 1: Get some empirical data on how children and adults interpret the same pronoun in a context where multiple cues are available.





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

Forsythe & Pearl 2019, Pearl & Forsythe under review





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.

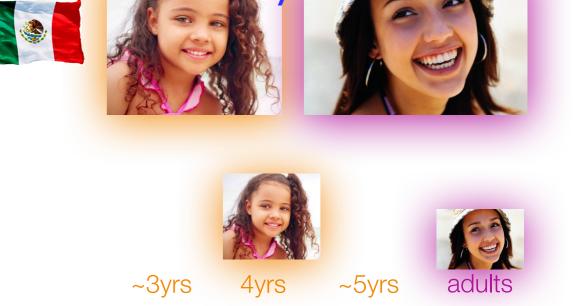
sg

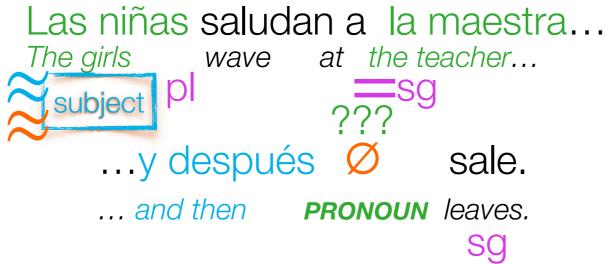
Children (~3, 4, and ~5) and adults are asked to interpret pronouns in the kind of contexts we saw before.

















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



...y después

sale.

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

sg



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









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

...y después Ø

... and then

PRONOUN leaves.

sale.



Context: Does agreement favor the subject or the object?

subject object

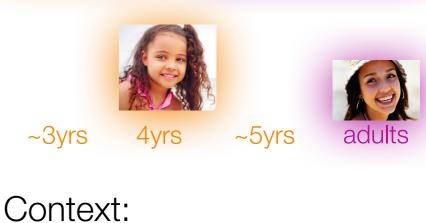






Rate of subject responses

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



Does agreement favor the subject or the object?

subject object Favored by agreement







...y después

... and then

PRONOUN leaves.

sale.

sg





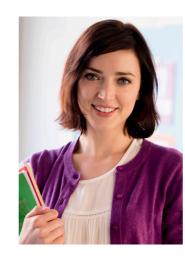


Context:

Does agreement favor the subject or the object?

subject object







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



Rate of subject responses

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

subject object







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

...y después ... and then

leaves. **PRONOUN**

sg



Context:

Rate of subject responses

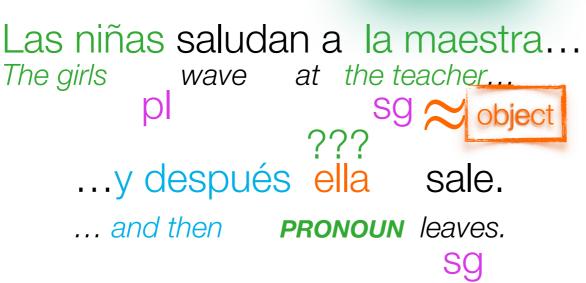
Does the pronoun form favor the subject or the object?

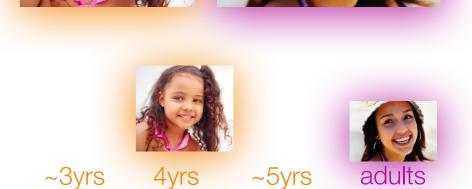
Ø (favors subject)

subject object









Rate of subject responses

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

overt (favors object)

subject object



Rate of subject responses











Las niñas saludan a la maestra...

The girls wave at the teacher...

Sg

...y después

sale.

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

Sg

Context:

Does the connective favor the subject or the object?

subject object













Las niñas saludan a la maestra...

The girls

Subject | D

wave at the teacher...

???

...y después

sale.

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

sg

Context:

Does the connective favor the subject or the object?

y después (favors subject)

subject object



Rate of subject responses











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

...porque

... because

Sg \propto object

sale.

PRONOUN leaves.

Sg

Context:

Does the connective favor the subject or the object?

porque (favors object)

subject object

Empirical data on pronoun interpretation Favored by connective y después (favors subject) porque (favors object) adults ~5yrs ~3yrs 4yrs Ø (favors subject) overt (favors object) Rate of subject responses Favored by form

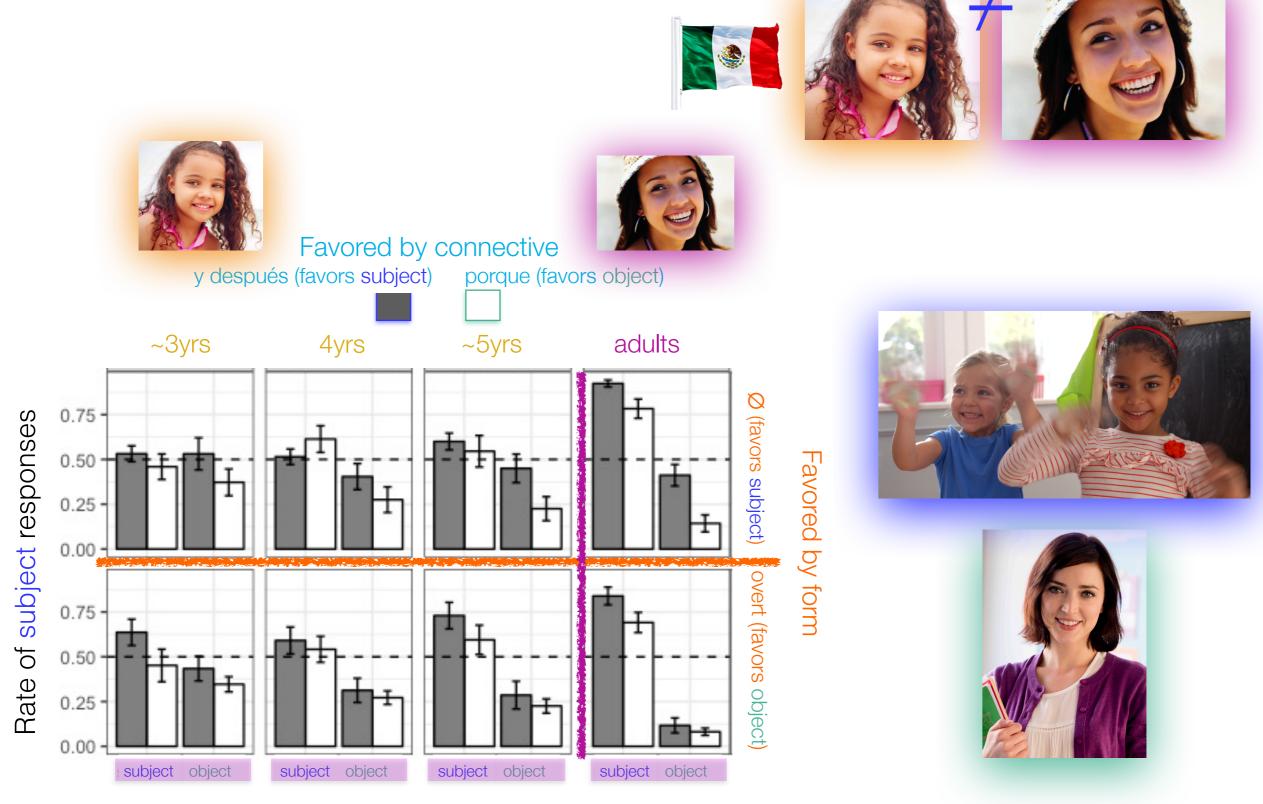
subject object

Favored by agreement

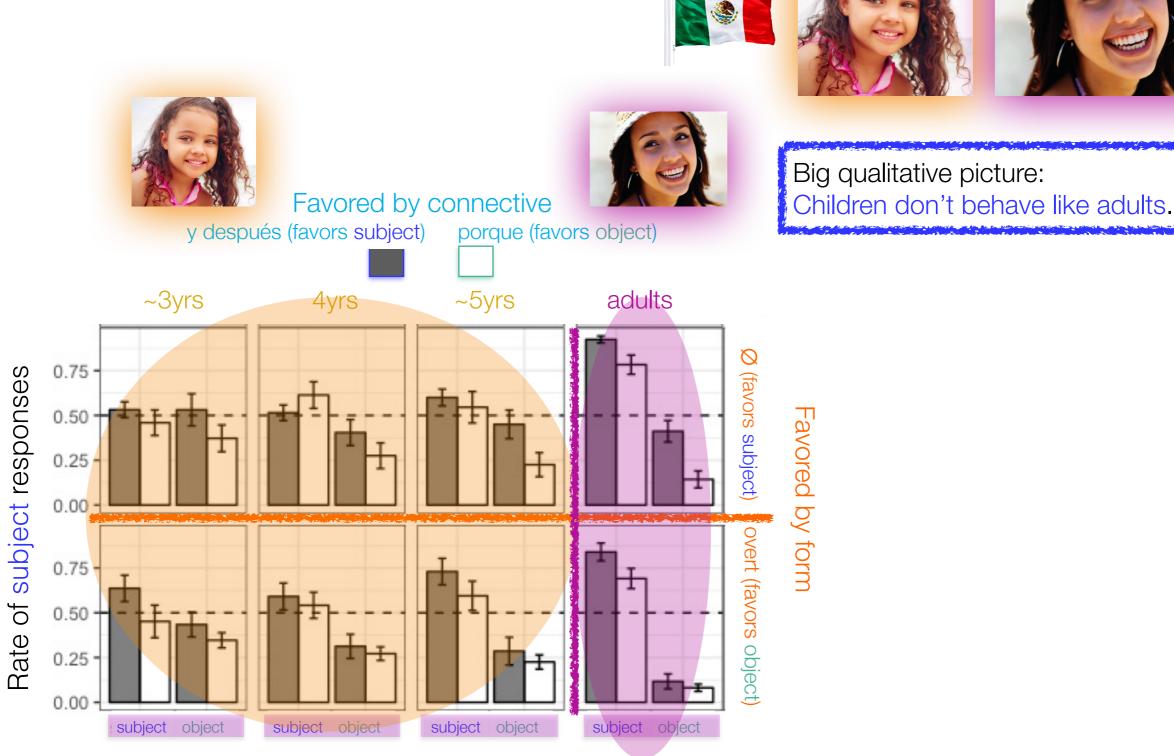
subject object

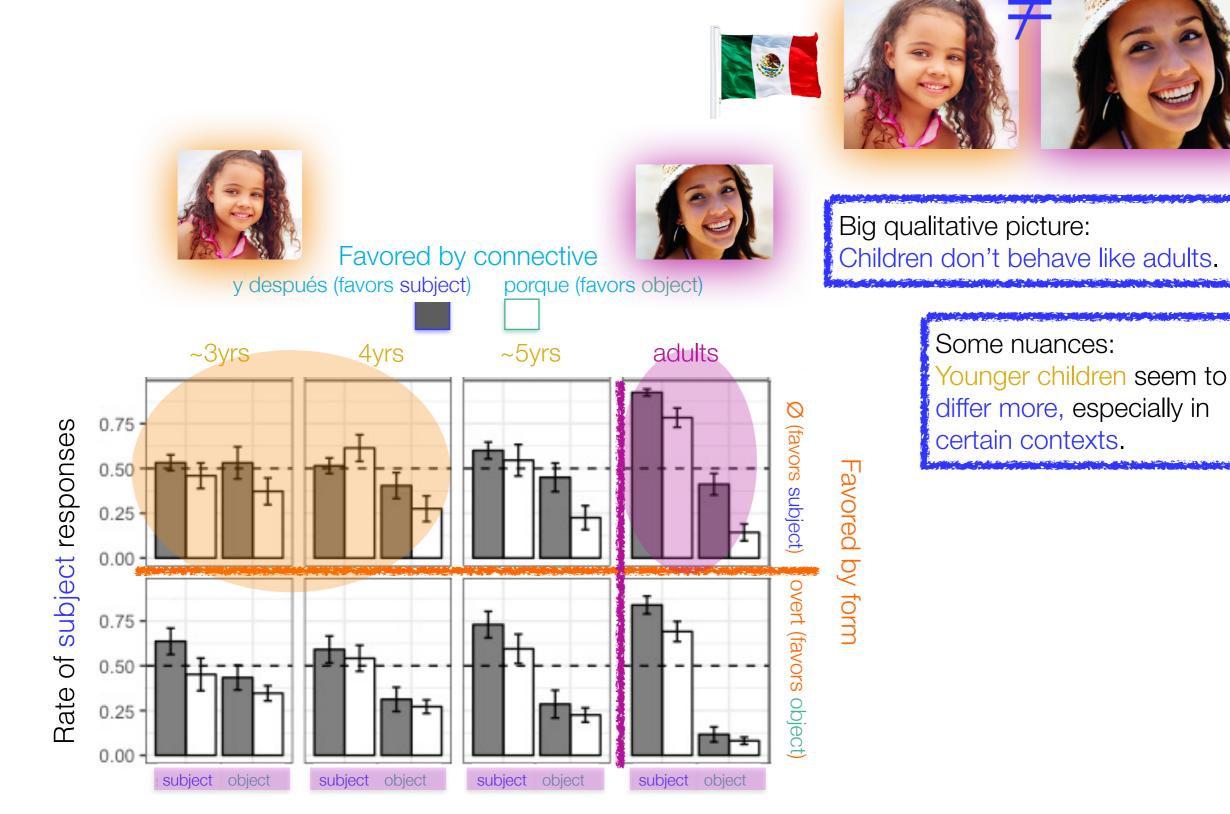
subject object

subject object

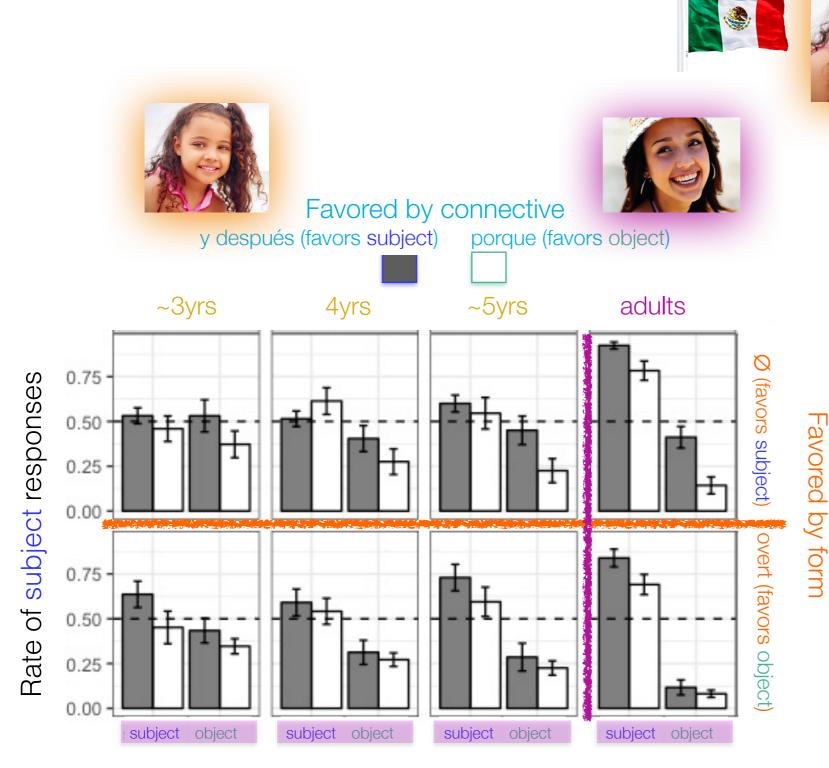


Favored by agreement



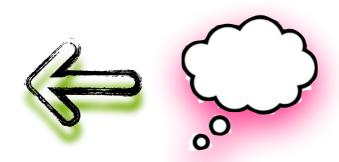


Favored by agreement



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

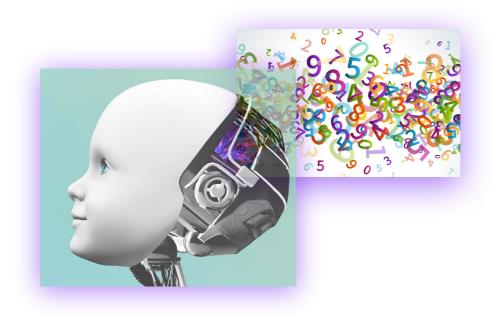




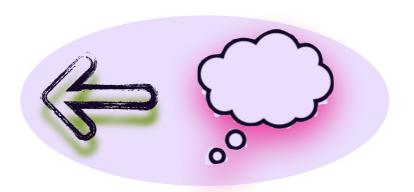
The plan, part 2:

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



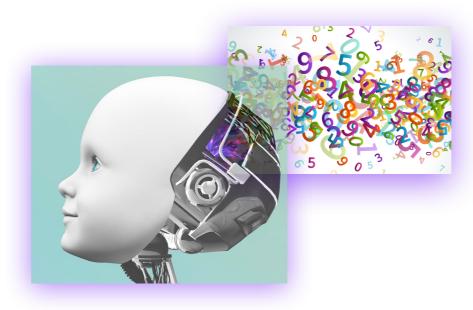




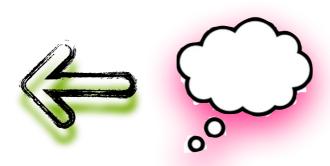


The computational cognitive model formally articulates and implements (what we think are) relevant aspects of pronoun interpretation in context.

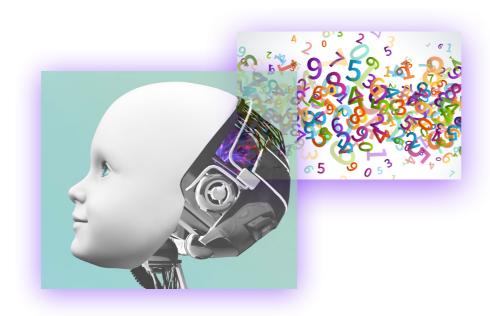




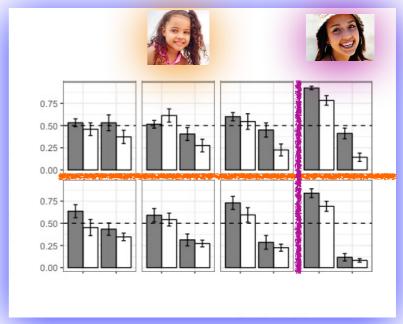




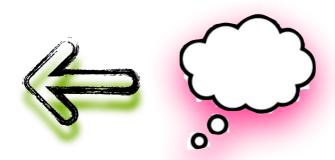
Here: Implement how a modeled listener represents pronoun information and deploys that information in order to predict the probability of a particular interpretation in context.





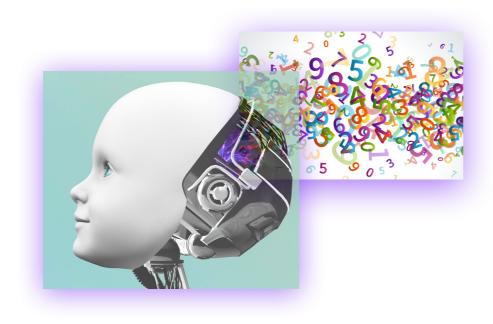


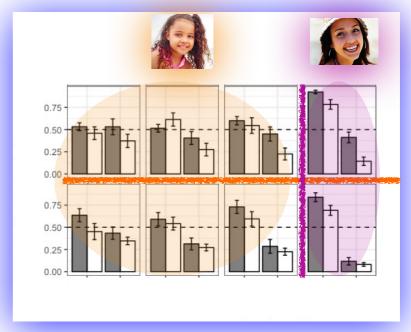




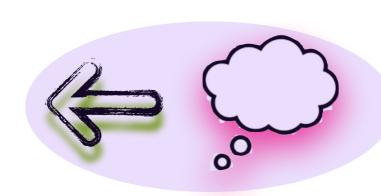


Then, see which options for representation and deployment best match child vs. adult pronoun interpretation behavior.



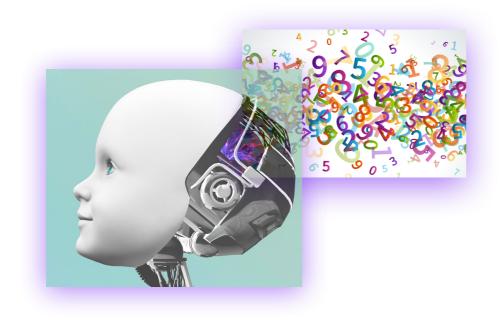


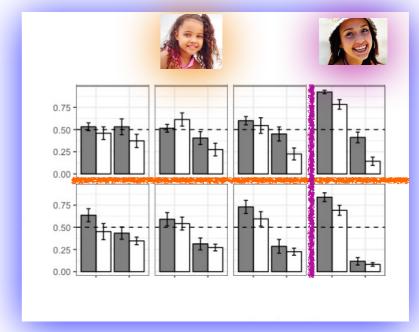


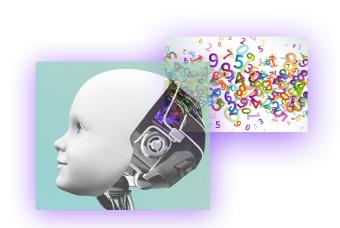


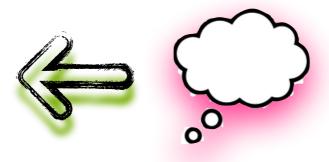


This will articulate how children (potentially) differ from adults, and what needs to develop in children for them to become adult-like.



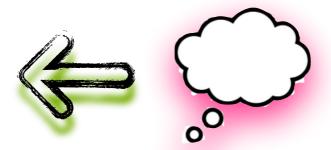




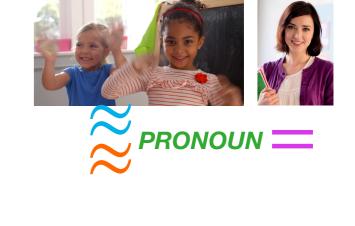


The model uses Bayesian inference to implement the cognitive mechanism that combines information to generate a particular interpretation in context.

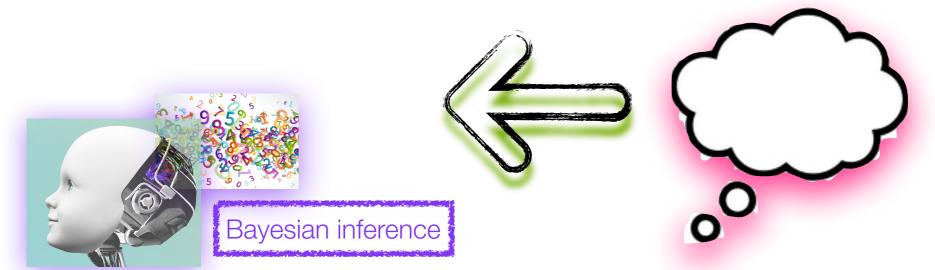




Bayesian inference is commonly used to model human cognition in general and language development in particular, since it matches human behavior quite well (see Pearl in press and b for recent reviews).



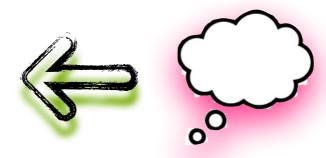




The particular Bayesian model we use is adapted from Gagliardi, Feldman & Lidz (2017), and offers one way to separate out the contributions of information representation vs. information deployment in the moment.







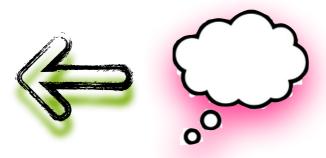


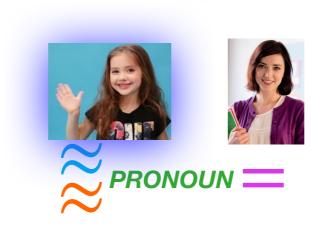
interpretation

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

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



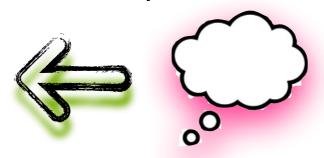


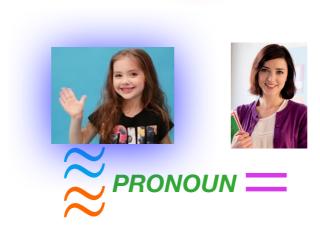


interpretation CONTEXT
$$p(\alpha_{subj.\rm SG}|{\rm FORM,CON,MOR}) \propto p(\alpha_{subj.\rm SG}) * p({\rm FORM,CON,MOR}|\alpha_{subj.\rm SG})$$

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





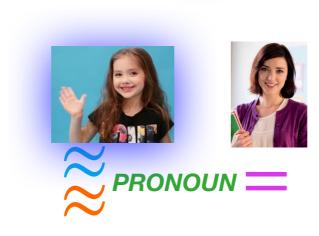


```
interpretation COntext p(\alpha_{subj.\text{SG}}|\text{FORM},\text{CON},\text{MOR}) \propto p(\alpha_{subj.\text{SG}}) * p(\text{FORM},\text{CON},\text{MOR}|\alpha_{subj.\text{SG}}) y \ \textit{despu\'es} \quad \textit{porque}
```

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



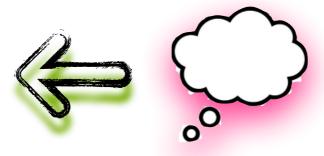




```
interpretation CONTEXT p(\alpha_{subj.\rm SG}|{\sf FORM,CON,MOR}) \propto p(\alpha_{subj.\rm SG}) * p({\sf FORM,CON,MOR}|\alpha_{subj.\rm SG})
```

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



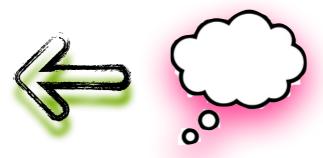




```
interpretation context p(\alpha_{subj.SG}|	extbf{FORM},	extbf{CON},	extbf{MOR}) \propto p(\alpha_{subj.SG}) * p(	extbf{FORM},	extbf{CON},	extbf{MOR}|\alpha_{subj.SG})
```

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).



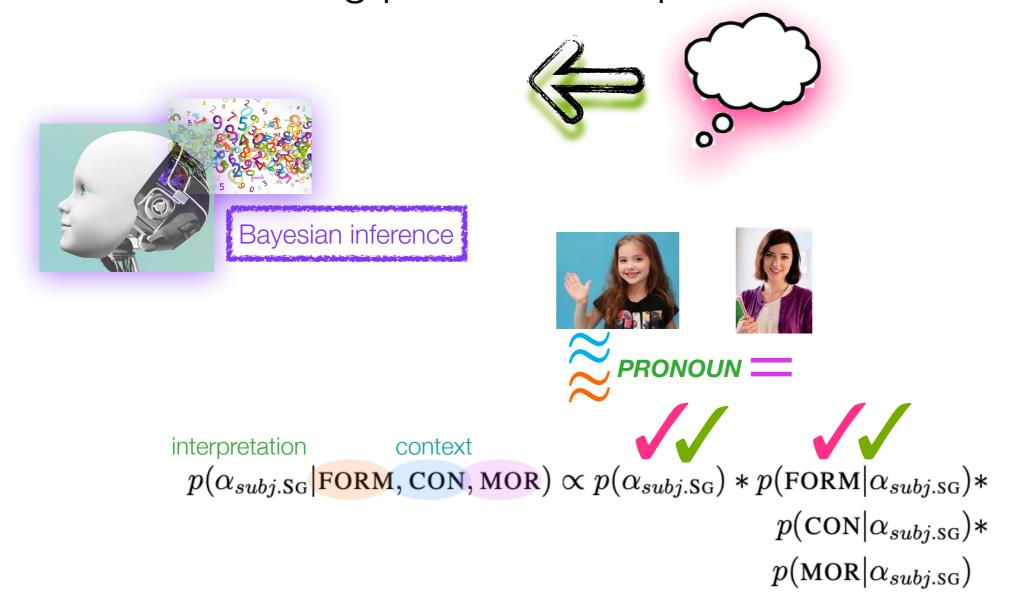




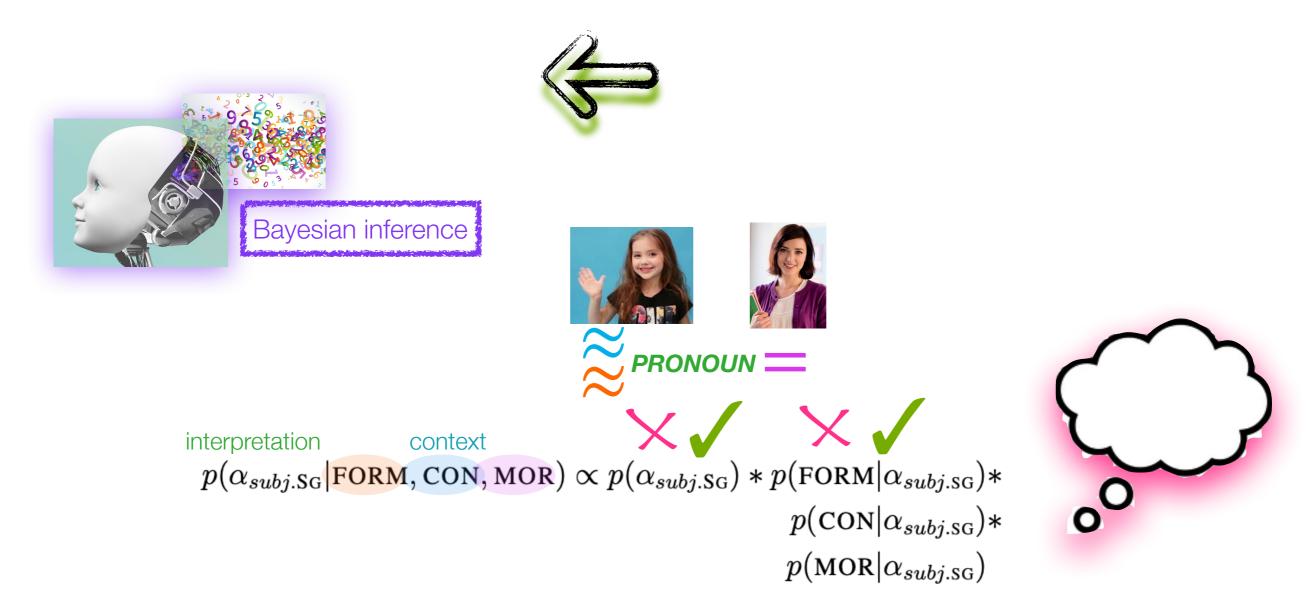
```
interpretation context p(lpha_{subj.\mathrm{SG}}|	extbf{FORM},	extbf{CON},	extbf{MOR}) \propto p(lpha_{subj.\mathrm{SG}}) * p(	extbf{FORM}|lpha_{subj.\mathrm{SG}}) * p(	extbf{CON}|lpha_{subj.\mathrm{SG}}) * p(	extbf{MOR}|lpha_{subj.\mathrm{SG}})
```

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

Not implausible: In visual perception, human behavior is best captured by models assuming features are independent (Vul & Rich 2010).



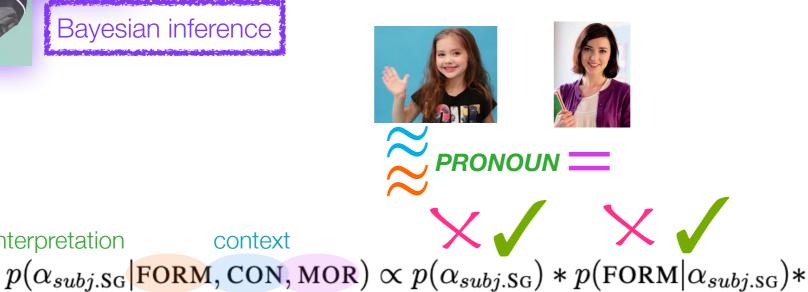
This is the baseline model, which has accurate representations of information and accurately deploys those representations in the moment.



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









inaccurate representations

interpretation

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

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

context

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

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













context

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











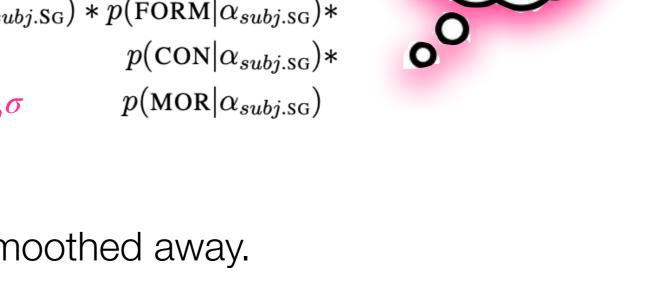


 σ <1: probability differences are smoothed away.

 σ =1: probabilities remain accurate.

0.324 vs. 0.676

 σ >1: probability differences are sharpened.





















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



inaccurate representations

interpretation

probability σ







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

 σ =1: probabilities remain accurate.

 σ >1: probability differences are sharpened.

context

0.409 vs. 0.591

0.324 vs. 0.676







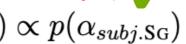






context

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





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

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



probability $^{\sigma}$

About σ : - \mathfrak{O} -



 σ <1: probability differences are smoothed away.

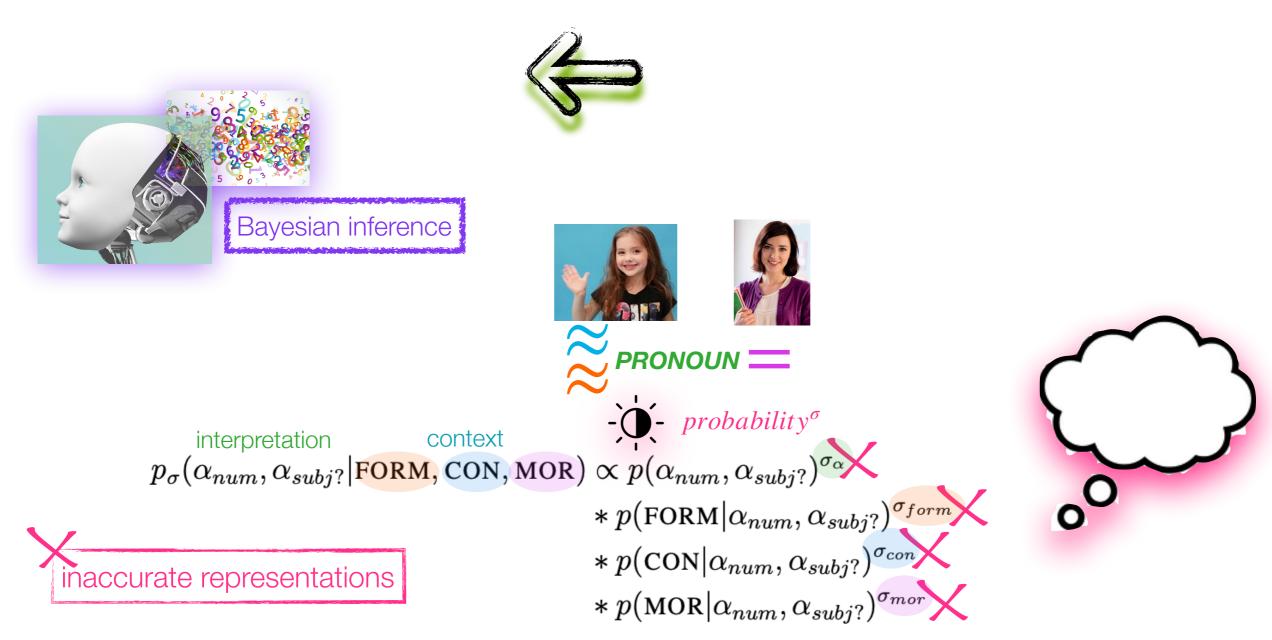
 σ =1: probabilities remain accurate.

0.324 vs. 0.676

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

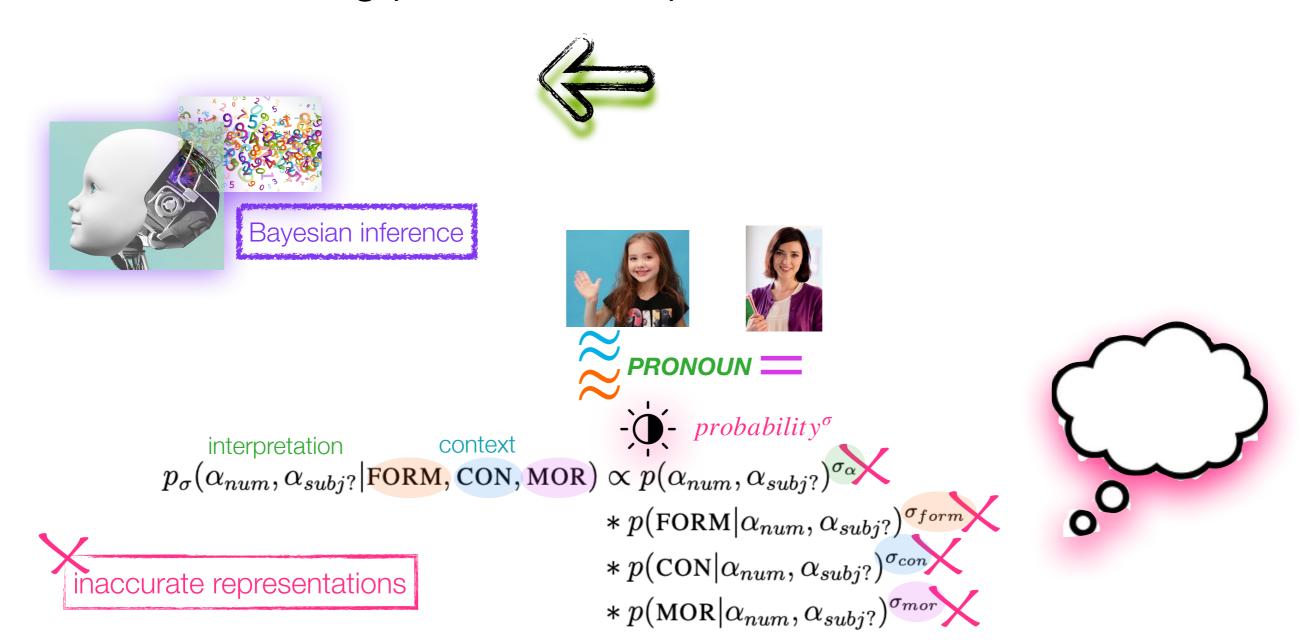
0.187 vs. 0.813





One σ for each information type:

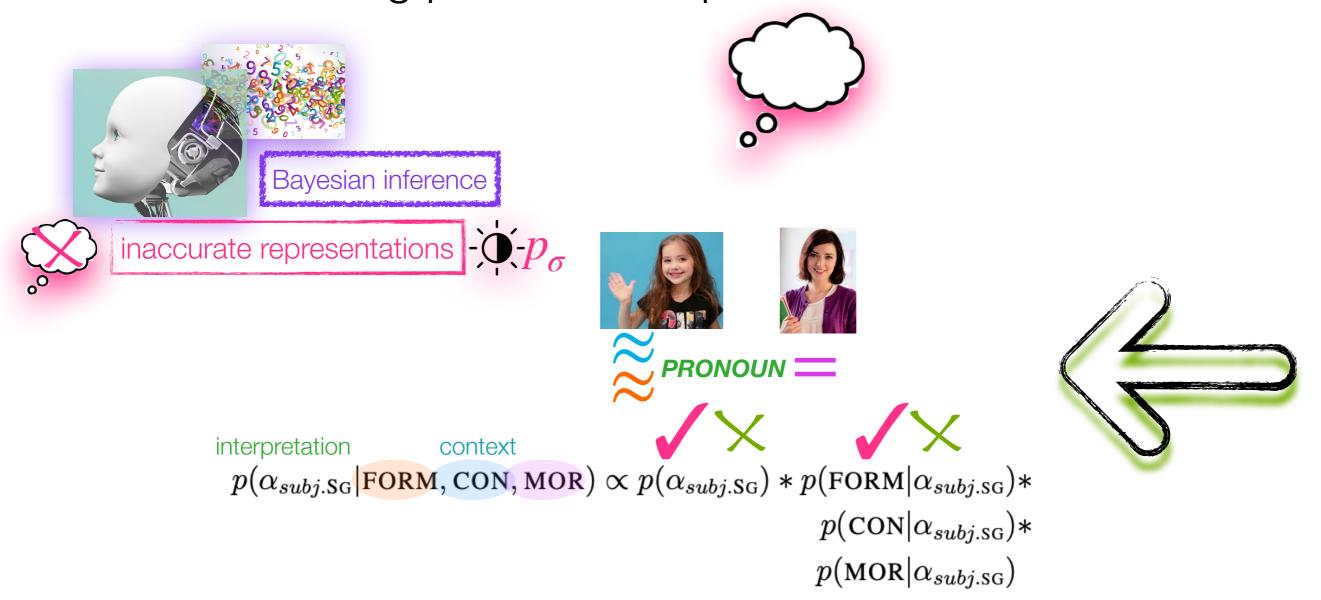
(in the prior) σ_{α} (in the likelihood) σ_{form} , σ_{con} , σ_{mor}



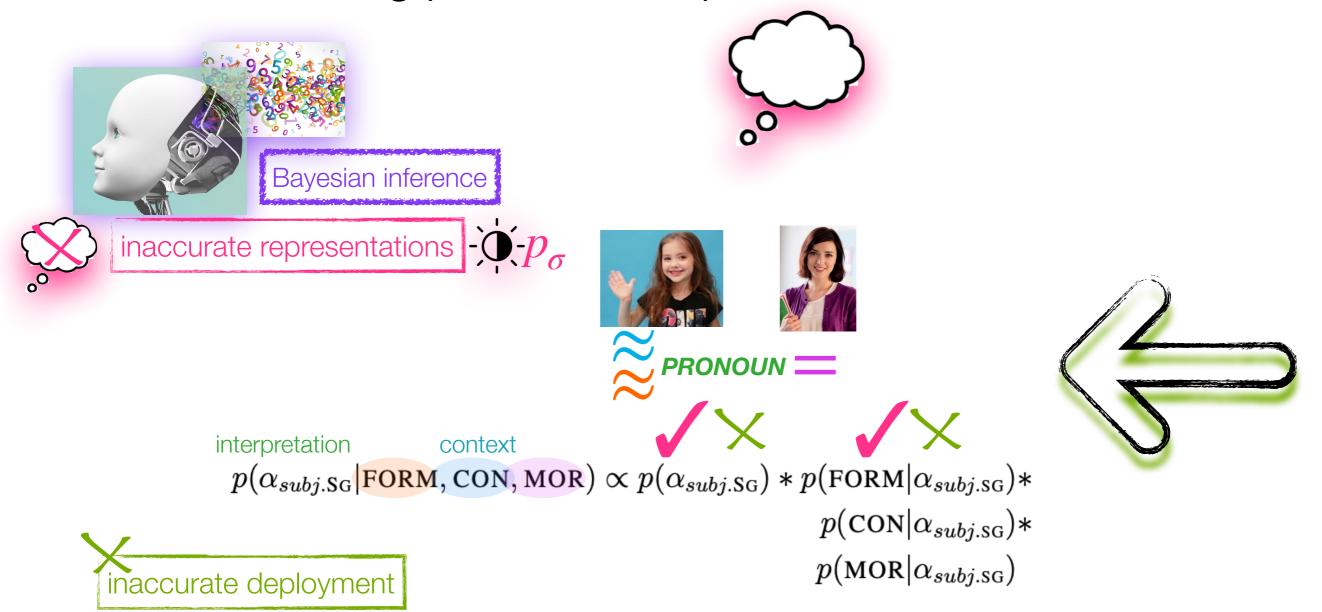
We allow $0.00 \le \sigma \le 4.00$, and see which σ value combinations best predict child and adult pronoun interpretation behavior.



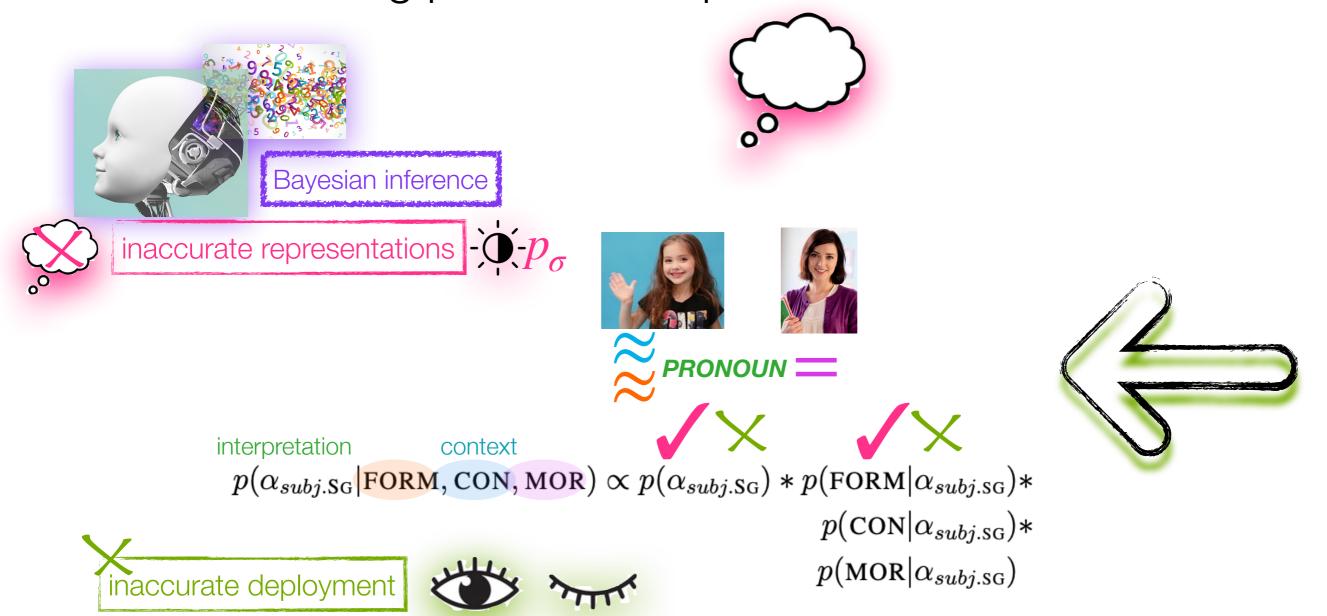




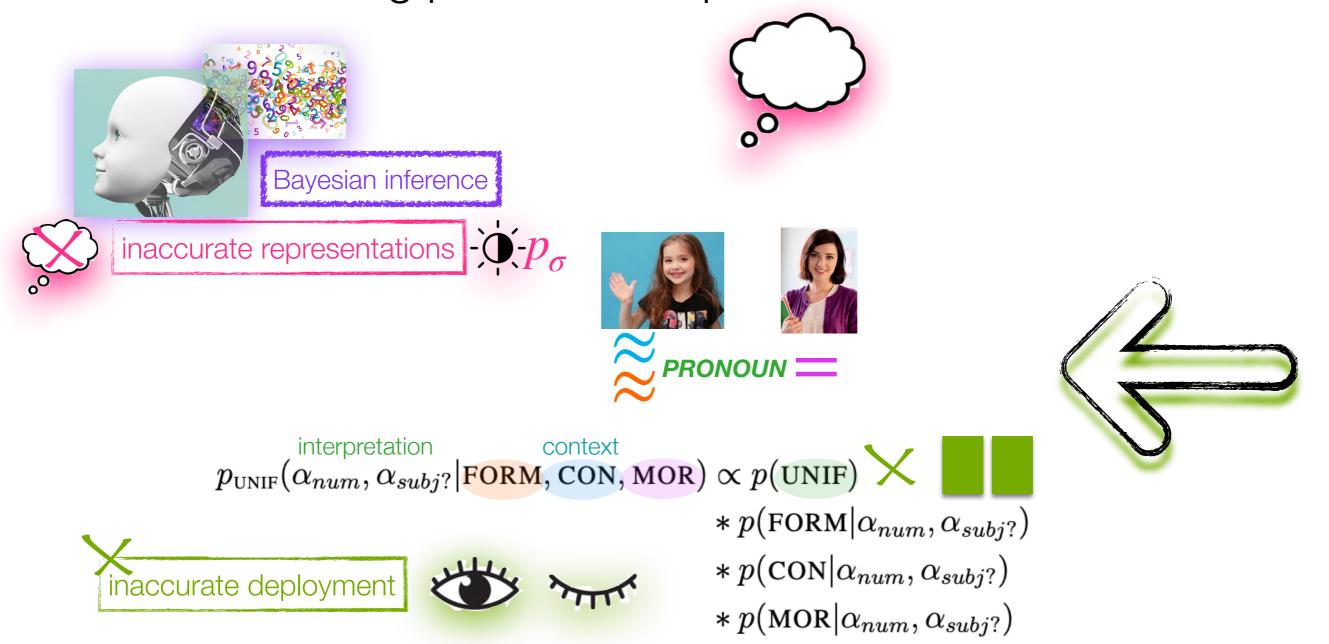
What about a modeled listener who has inaccurate deployment of information in the representations? This could involve inaccurately deploying the prior or the likelihood information, or both.



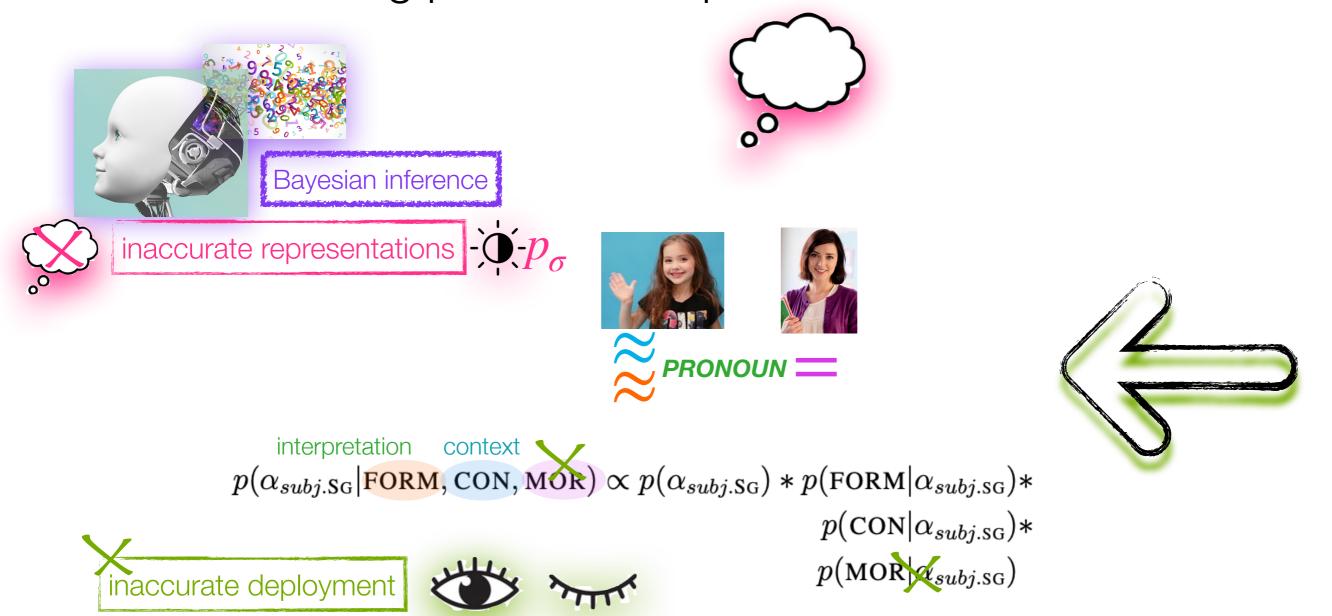
We implement this as ignoring that information. So, for any piece of information, the modeled listener either pays attention to it (and so uses it) or ignores it in the moment.



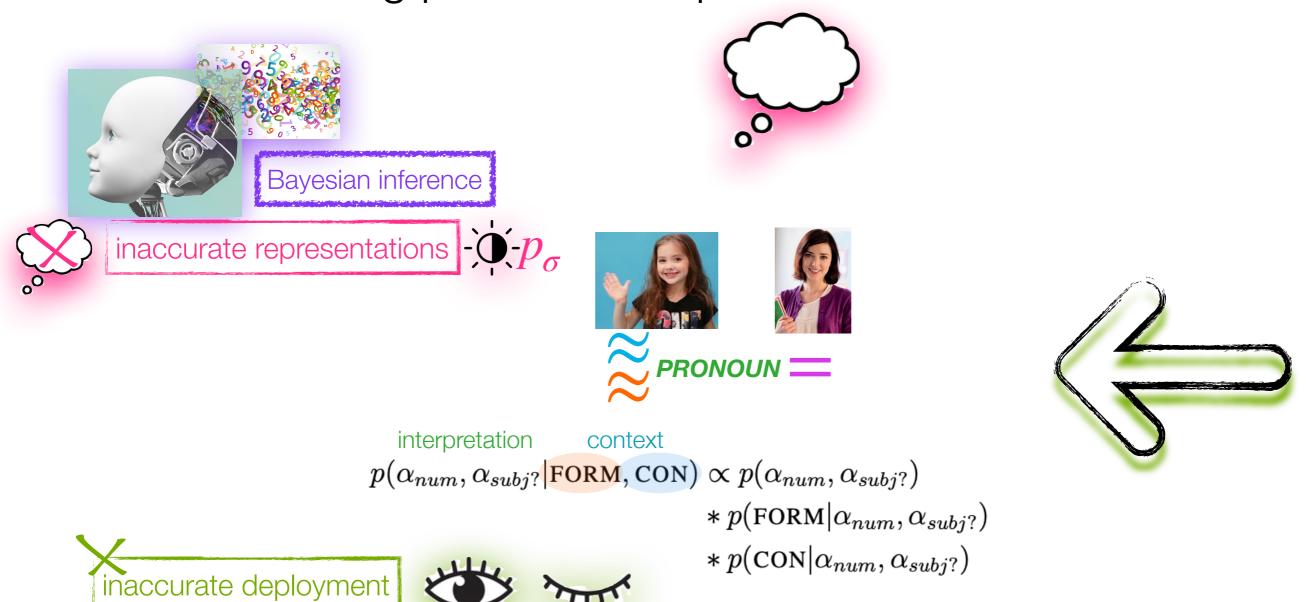
Not using information means not incorporating it into the inference.



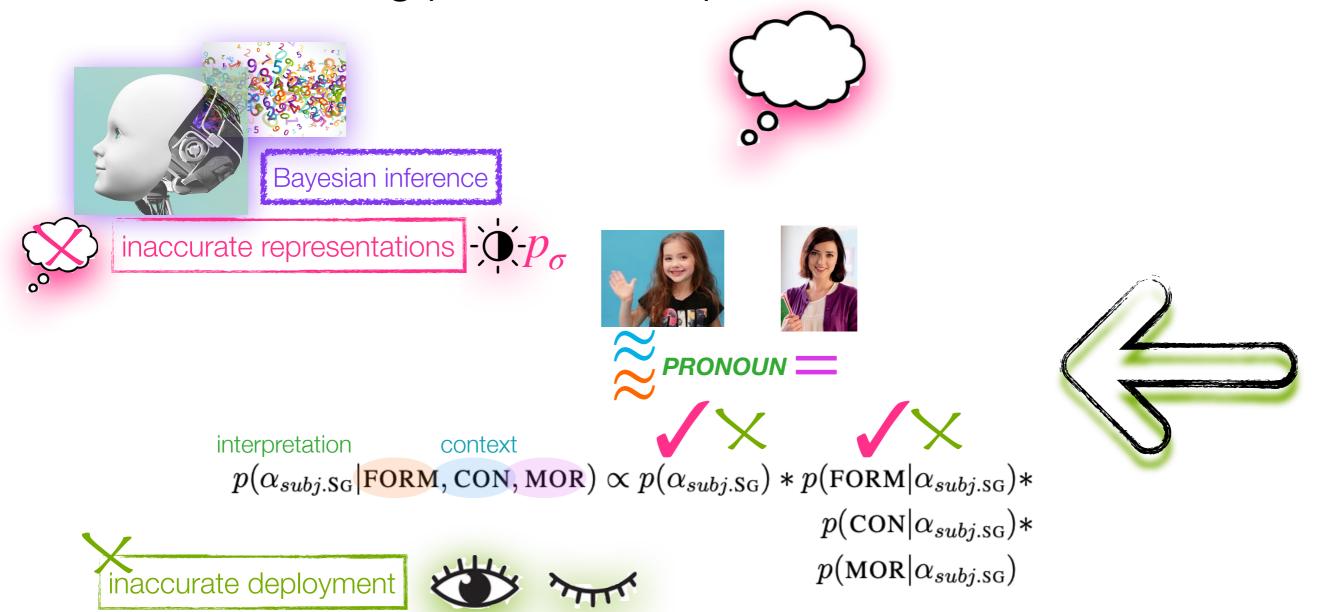
Not using the prior means relying on a uniform (uninformative) prior.



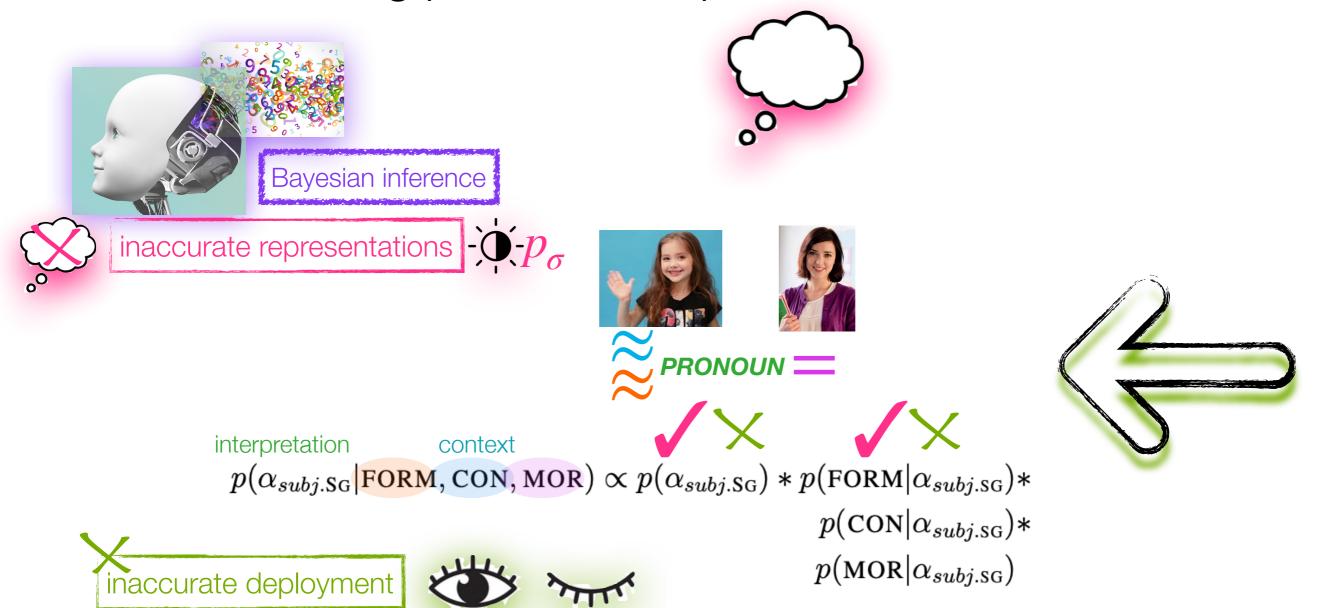
Not using likelihood information for a cue means not using that cue's information. For example, ignoring morphology information means not using the morphology likelihood.



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.



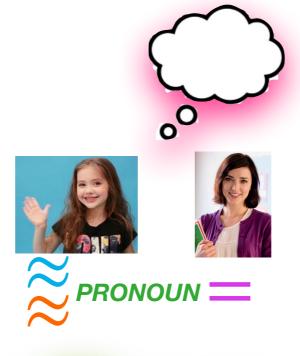
Each of the four information types has its own β : (prior) β_{α}

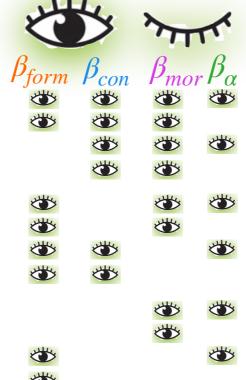
(likelihood) β_{form} , β_{con} , β_{mor}





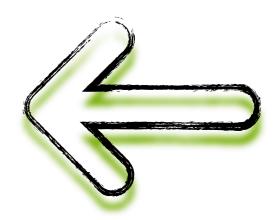
interpretation context $p_{\beta}(\alpha|\text{FORM}, \text{CON}, \text{MOR}, \alpha_{num}, \alpha_{subj?}) =$ $(\beta_{form})(\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|\text{FORM, CON, MOR, }\alpha_{num},\alpha_{subj?}) +$ $(\beta_{form})(\beta_{con})(\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM,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





*

5

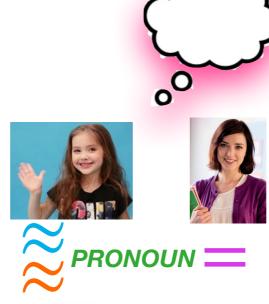


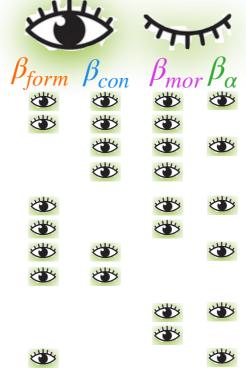
This yields 16 possible use combinations for any particular moment, implemented with a mixture model p_{β} .

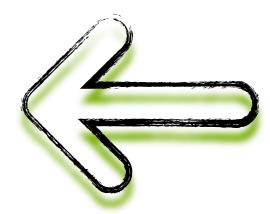




interpretation context $p_{\beta}(\alpha|\text{FORM}, \text{CON}, \text{MOR}, \alpha_{num}, \alpha_{subj?}) =$ $(\beta_{form})(\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|\text{FORM, CON, MOR, }\alpha_{num},\alpha_{subj?}) +$ $(\beta_{form})(\beta_{con})(\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM,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





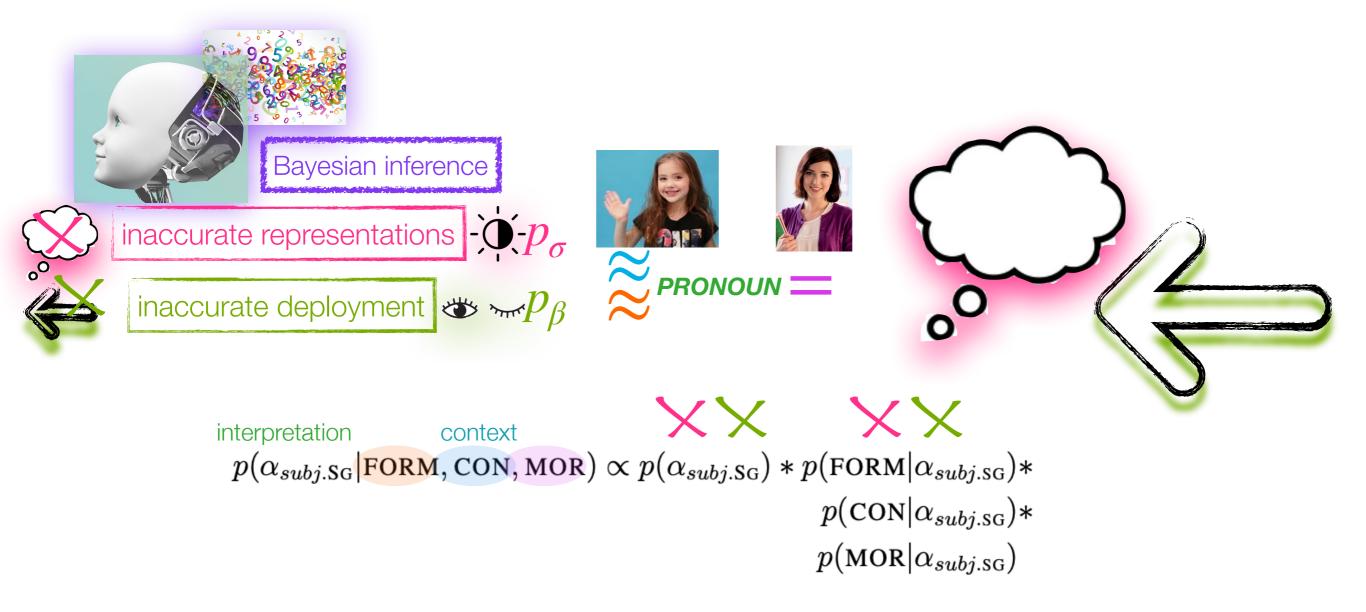


We allow $0 \le \beta \le 1$, and see which β value combinations best predict child and adult pronoun interpretation behavior.

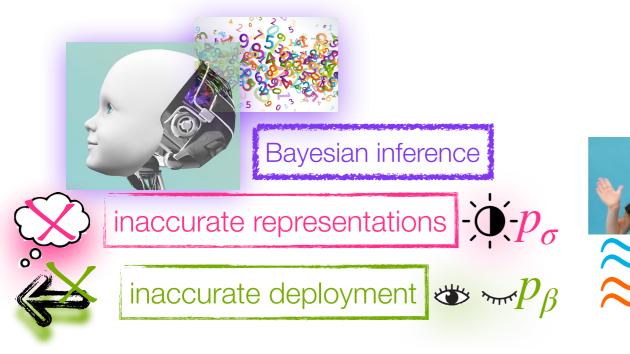




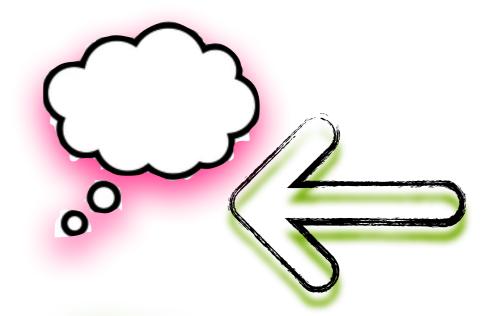
Forsythe & Pearl 2019, Pearl & Forsythe under review



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









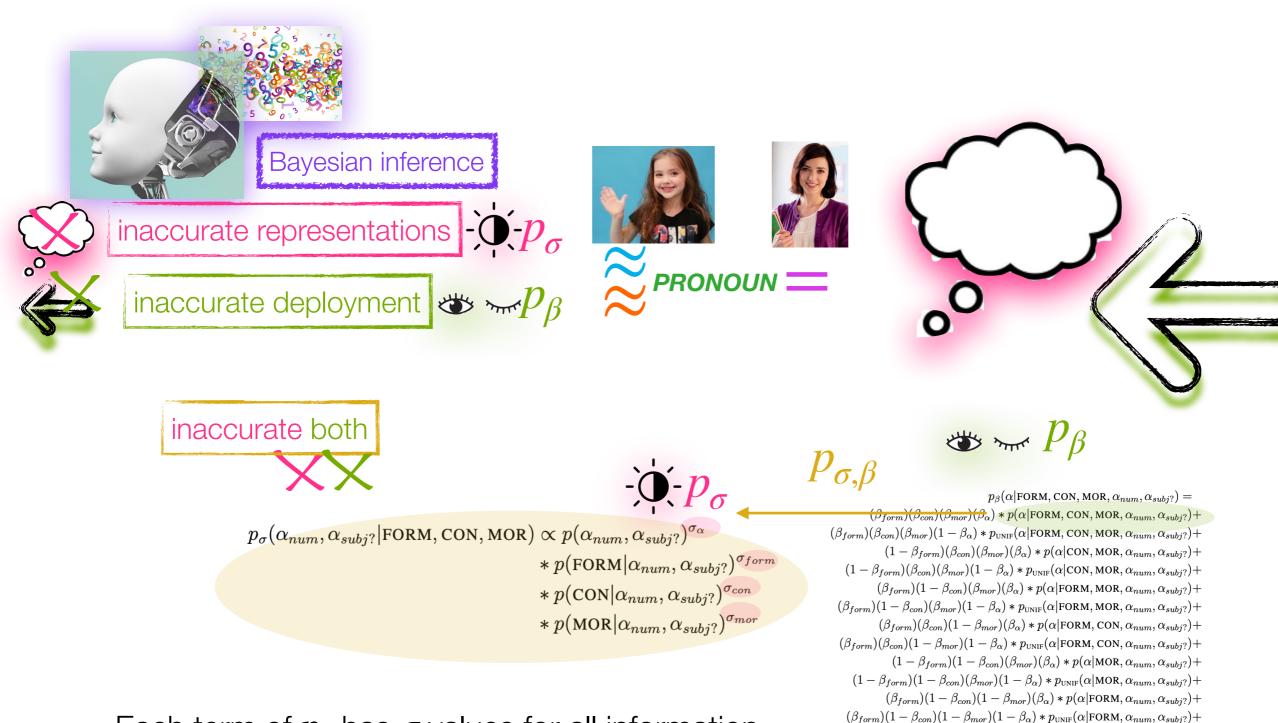
$$p_{\sigma}(\alpha_{num}, \alpha_{subj?}| ext{FORM, CON, MOR}) \propto p(\alpha_{num}, \alpha_{subj?})^{\sigma_{\alpha}} \\ * p(ext{FORM}|\alpha_{num}, \alpha_{subj?})^{\sigma_{form}} \\ * p(ext{CON}|\alpha_{num}, \alpha_{subj?})^{\sigma_{con}} \\ * p(ext{MOR}|\alpha_{num}, \alpha_{subj?})^{\sigma_{mor}}$$

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_{\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?}) +
```

 P_{β}

 $(1 - \beta_{form})(1 - \beta_{con})(1 - \beta_{mor})(1 - \beta_{\alpha}) * p(UNIF)$



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

 $(1 - \beta_{form})(\beta_{con})(1 - \beta_{mor})(\beta_{\alpha}) * p(\alpha | CON, \alpha_{num}, \alpha_{subj?}) +$

 $(1 - \beta_{form})(1 - \beta_{con})(1 - \beta_{mor})(\beta_{\alpha}) * p(\alpha_{num}, \alpha_{subj?}) + (1 - \beta_{form})(1 - \beta_{con})(1 - \beta_{mor})(1 - \beta_{\alpha}) * p(\text{UNIF})$

 $(1 - \beta_{form})(\beta_{con})(1 - \beta_{mor})(1 - \beta_{\alpha}) * p_{\text{UNIF}}(\alpha|\text{CON}, \alpha_{num}, \alpha_{subj?}) +$





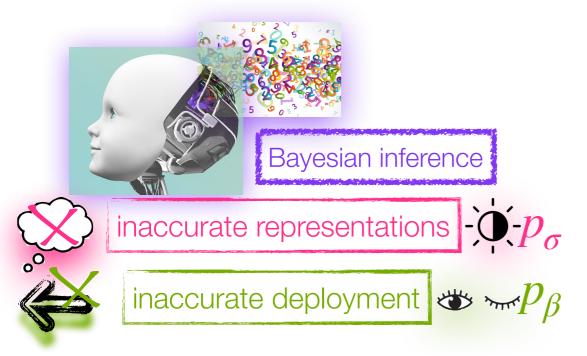


 $p_{\sigma}(\alpha_{num}, \alpha_{subi?}|\text{FORM, CON, MOR}) \propto p(\alpha_{num}, \alpha_{subi?})^{\sigma_{\alpha}}$ * $p(\text{FORM}|\alpha_{num},\alpha_{subj?})^{\sigma_{form}}$ * $p(\text{CON}|\alpha_{num}, \alpha_{subj?})^{\sigma_{con}}$ * $p(MOR|\alpha_{num}, \alpha_{subj?})^{\sigma_{mor}}$

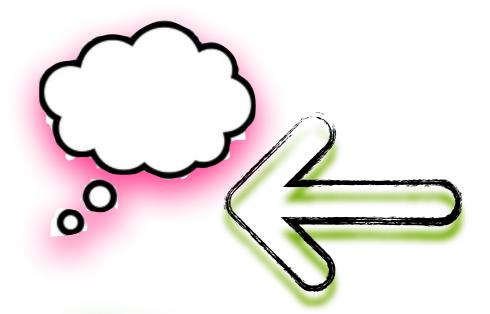
So, $p_{\sigma,\beta}$ has 8 parameter values: (in the prior) σ_{α} , β_{α} (in the likelihood) σ_{form} , σ_{con} , σ_{mor} , ρ_{form} , ρ_{con} , ρ_{mor}

 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(\text{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 $0.00 \le \sigma \le 4.00$ and $0 \le \beta \le 1$, and see which σ and β value combinations best predict child and adult pronoun interpretation behavior.

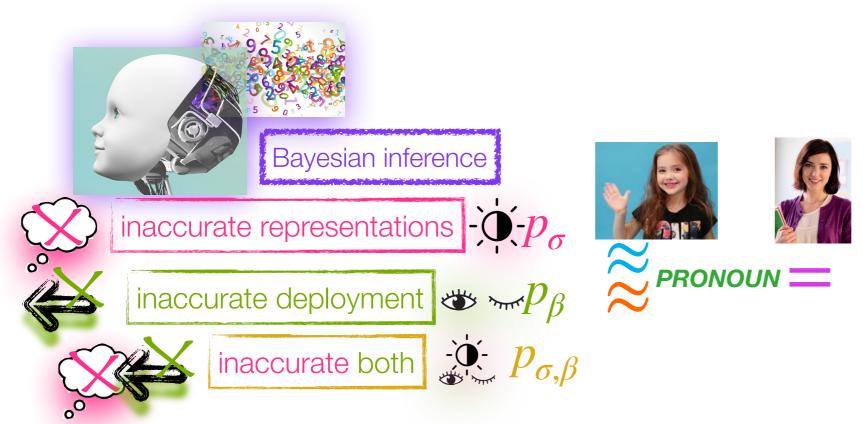






```
p_{\beta}(\alpha|\text{FORM}, \text{CON}, \text{MOR}, \alpha_{num}, \alpha_{subj?}) =
               (\beta_{form})(\beta_{con})(\beta_{mor})(\beta_{\alpha}) * p(\alpha|FORM, CON, MOR, \alpha_{num}, \alpha_{subi?}) +
(\beta_{form})(\beta_{con})(\beta_{mor})(1-\beta_{\alpha})*p_{\text{UNIF}}(\alpha|\text{FORM,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

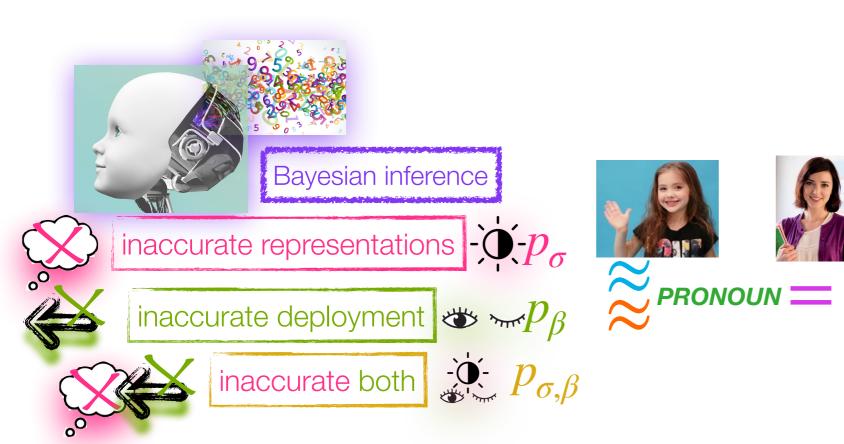
 P_{β}

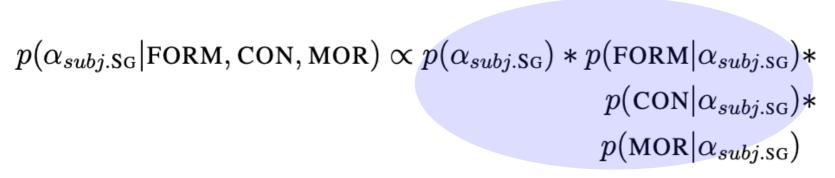




```
p(\alpha_{subj.SG}|{
m FORM,CON,MOR}) \propto p(\alpha_{subj.SG}) * p({
m FORM}|\alpha_{subj.SG}) * p({
m CON}|\alpha_{subj.SG}) * p({
m MOR}|\alpha_{subj.SG})
```

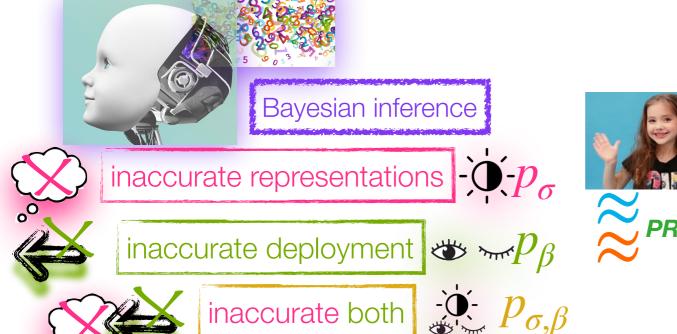
What input is the modeled listener using to represent the various information types?



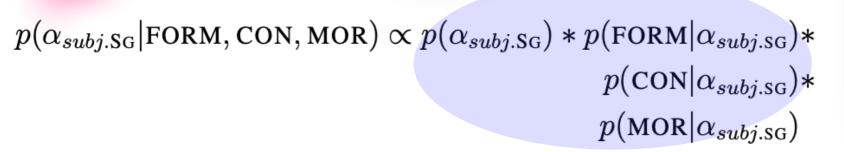




54,757 utterances of Mexico City spontaneous child-directed speech to children 1;6-5;11 from the Schmitt-Miller corpus (Miller & Schmitt 2012).



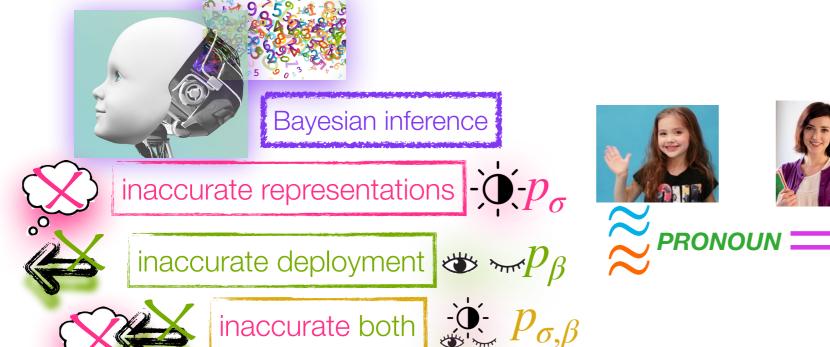




		prior	likelihoods					
		$p(\alpha)$	$p(\mathtt{FORM} lpha)$		$p(\mathtt{CON} lpha)$		$p(\mathtt{MOR} lpha)$	
antecedent type		$p(\alpha)$	ø	overt	después	porque	SG	PL
SUBJ	SG	0.362	0.938	0.062	0.324	0.676	0.998	0.002
	PL	0.071	0.984	0.016	0.750	0.250	0.005	0.995
¬SUBJ	SG	0.438	0.817	0.183	0.132	0.868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



From this, we estimate the relevant priors and likelihoods.

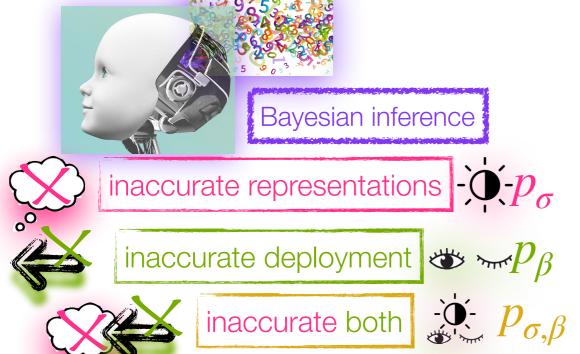


$$p(\alpha_{subj.SG}|{
m FORM,CON,MOR}) \propto p(\alpha_{subj.SG}) * p({
m FORM}|\alpha_{subj.SG}) * p({
m CON}|\alpha_{subj.SG}) * p({
m MOR}|\alpha_{subj.SG})$$

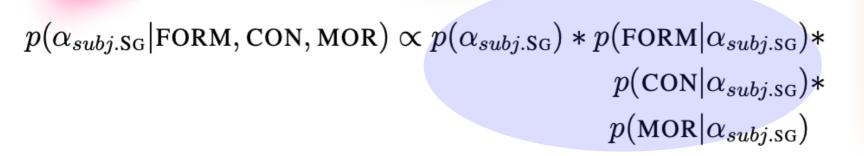
		prior	likelihoods						
		$p(\alpha)$	p(FOF)	$p(\mathtt{FORM} lpha)$		$p(\operatorname{CON} \alpha)$		p(MOR lpha)	
antecedent type		$p(\alpha)$	ø	overt	después	porque	SG	PL	
SUBJ	SG	0.362	0.938	0.062	0.324	0.676	0.998	0.002	
	PL	0.071	0.984	0.016	0.750	0.250	0.005	0.995	
¬SUBJ	SG	0.438	0.817	0.183	0.132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



Singular antecedents generally occur more often.



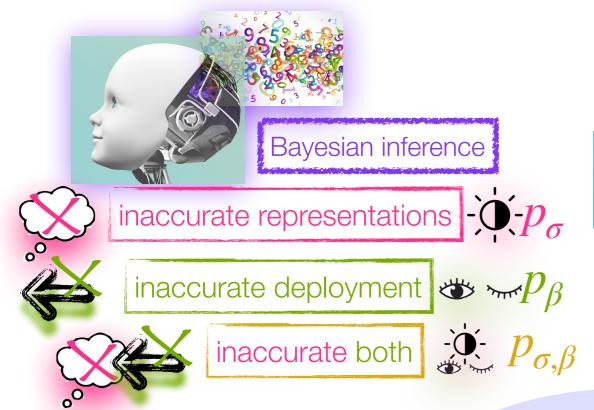




		prior	likelihoods					
		$p(\alpha)$	$p(\mathtt{FORM} lpha)$		$p(\operatorname{CON} \alpha)$		$p({ m MOR} lpha)$	
antecedent type		$p(\alpha)$	ø	overt	después	porque	SG	PL
SUBJ	SG	0.362	0.938	0.062	0.324	0.676	0.998	0.002
	PL	0.071	0.984	0.016	0.750	0.250	0.005	0.995
¬SUBJ	SG	0.438	0.817	0.183	0.132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



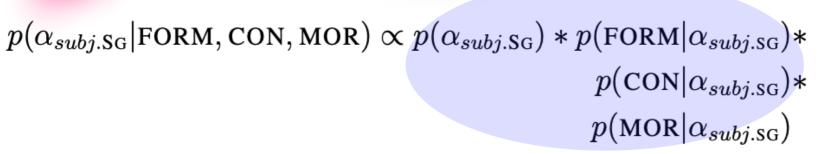








PRONOUN =

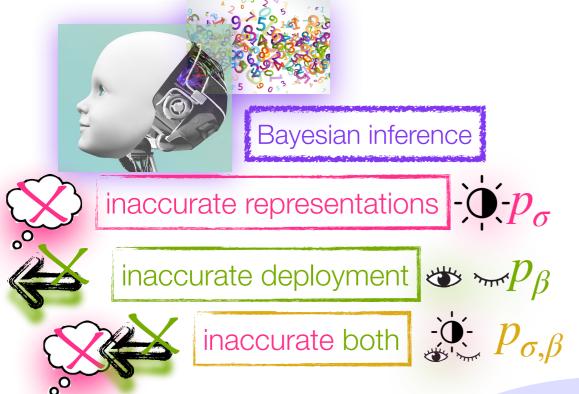


		prior	likelihoods						
		$p(\alpha)$	p(FOF)	$p(\mathtt{FORM} lpha)$		$p(\operatorname{CON} lpha)$		$p(\mathtt{MOR} lpha)$	
antecedent type		$p(\alpha)$	ø	overt	después	porque	SG	PL	
SUBJ	SG	0.362	0.938	0.062	0.324	0.676	0.998	0.002	
	PL	0.071	0.984	0.016	0.750	0.250	0.005	0.995	
¬SUBJ	SG	0.438	0.817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





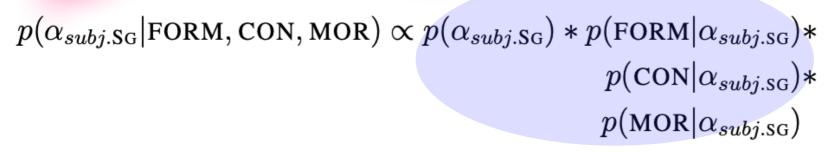








PRONOUN



		prior	likelihoods						
		$p(\alpha)$	p(FOF)	$p(\mathtt{FORM} lpha)$		$p(\mathtt{CON} lpha)$		$p(\mathtt{MOR} lpha)$	
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	

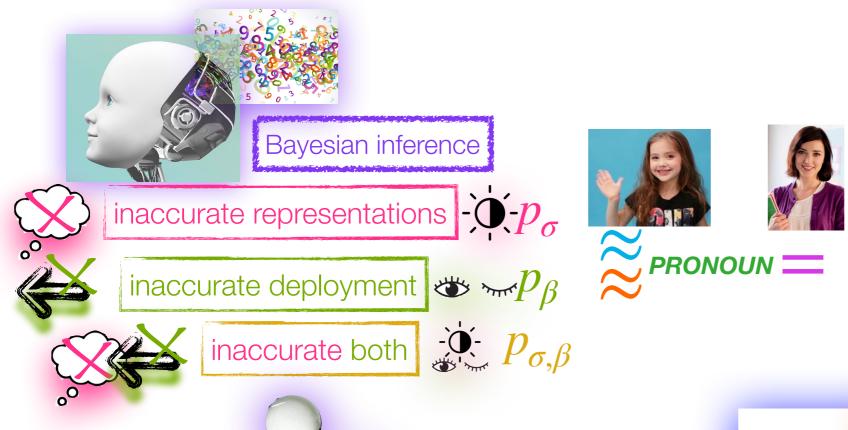


Agreement morphology is nearly categorical, with a very strong preference for matching morphology.

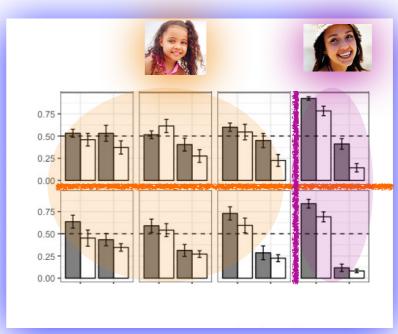


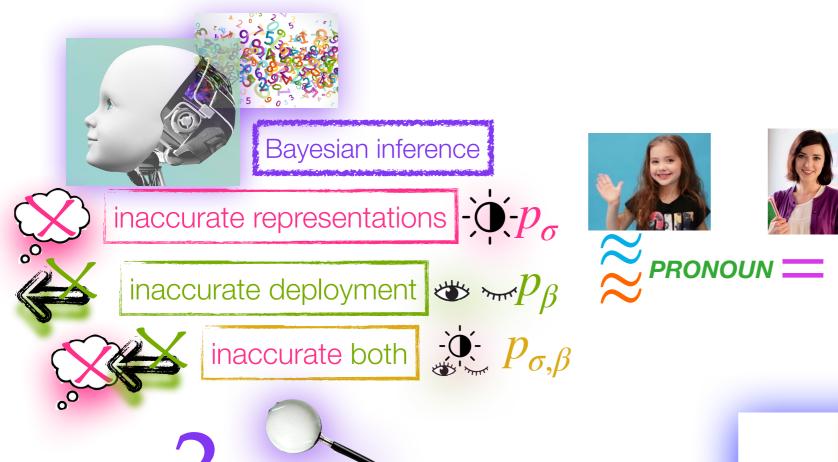




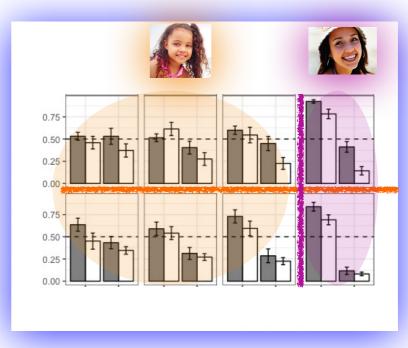


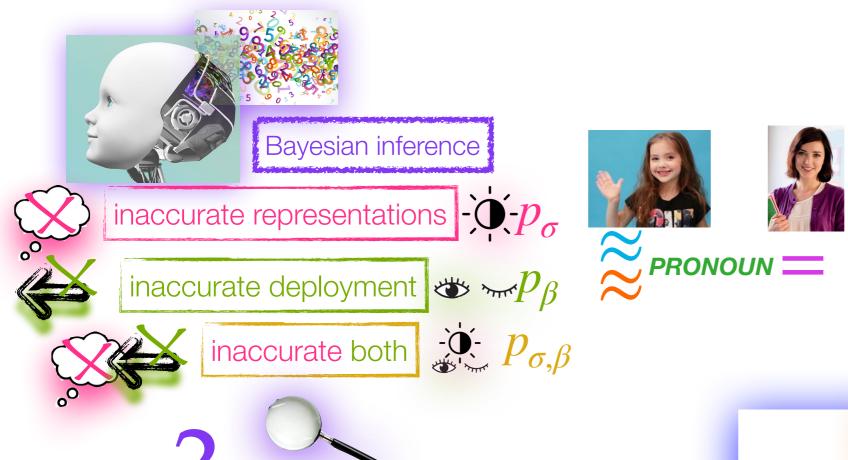
The plan, part 2: Use the computational cognitive model to identify the specific potential differences leading to child and adult pronoun interpretation in context.



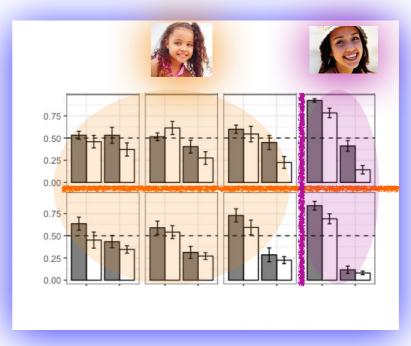


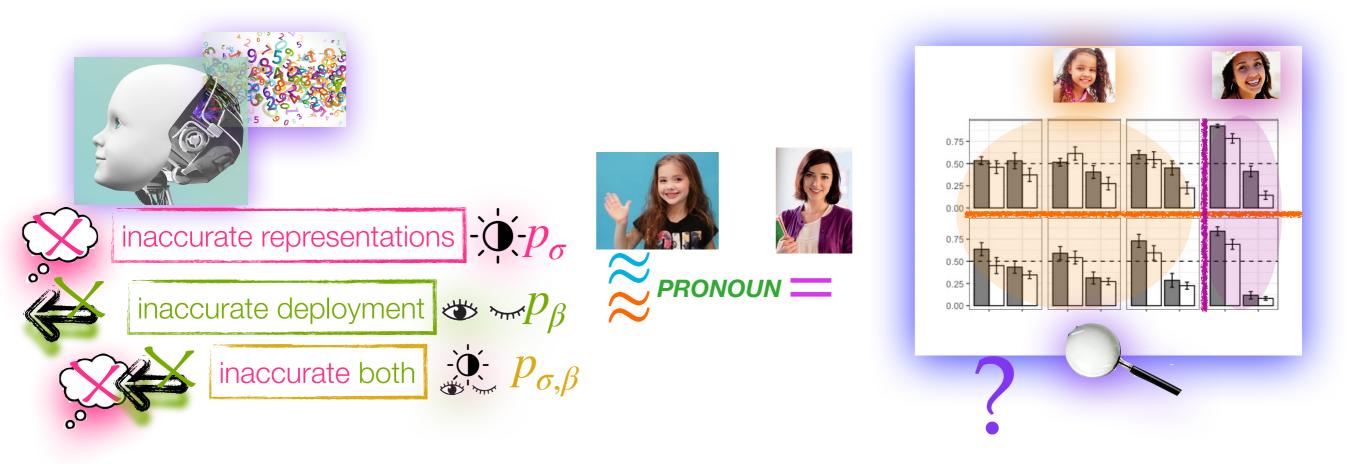
Which modeled listener variant best matches the observed pronoun interpretation behavior?



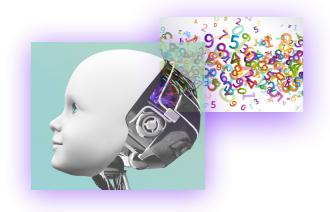


Important: Model variants with more parameters have an easier time fitting the data because they have more degrees of freedom.

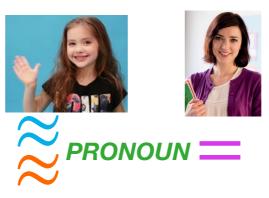




baseline: accurate representations and deployment p $p(\alpha_{subj.SG}|{
m FORM},{
m CON},{
m MOR}) \propto p(\alpha_{subj.SG}) * p({
m FORM}|\alpha_{subj.SG}) * O$ free parameters $p({
m CON}|\alpha_{subj.SG}) * p({
m MOR}|\alpha_{subj.SG})$







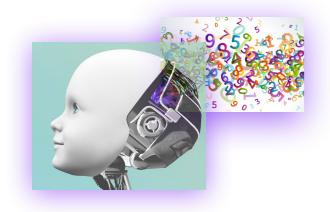


baseline: accurate representations and deployment

p 0 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



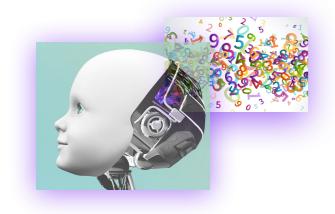


inaccurate representations - $\dot{\mathcal{D}}$ - $\dot{\mathcal{D}}$ 4 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor}





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







baseline: accurate representations and deployment

0 free parameters



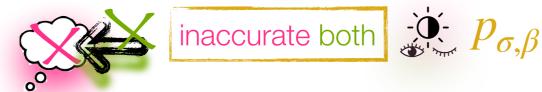


inaccurate representations - $\dot{\mathcal{D}}$ - $\dot{\mathcal{D}}_{\sigma}$ 4 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor}



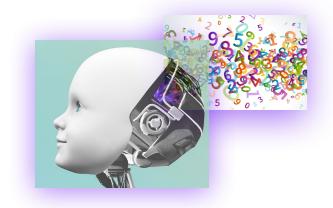


inaccurate deployment $\begin{cases} \begin{cases} \begin$

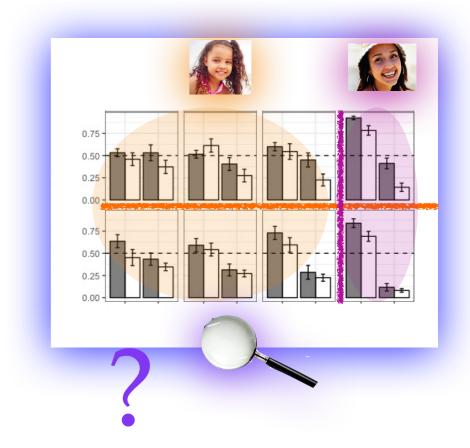




8 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor} , β_{α} , β_{form} , β_{con} , β_{mor}



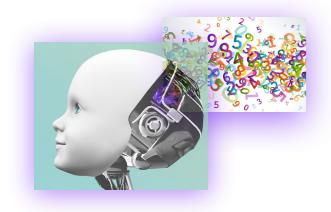






We want variants with more parameters to have a substantially better fit in order to favor them over variants with fewer parameters.

baseline: accurate representations and deployment P , 0 free parameters inaccurate representations P_{σ} 4 free parameters: P_{σ} 4 free parameters: P_{σ} 4 free parameters: P_{σ} 4 free parameters: P_{σ} 5 free parameters: P_{σ} 8 free parameters: P_{σ} 9 from $P_$

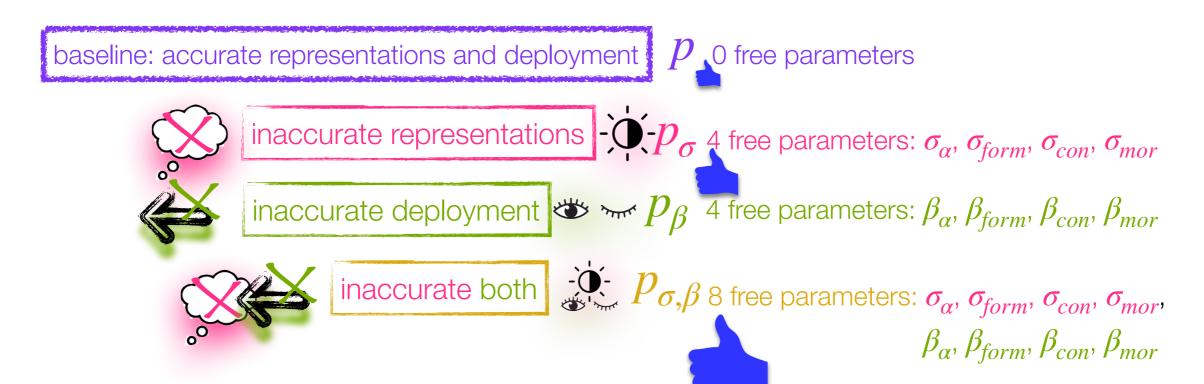


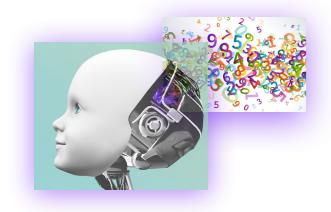






The Bayesian Information Criterion (BIC) is one way to quantify this preference (Schwarz 1978).









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

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

baseline: accurate representations and deployment

 p_{\bullet} 0 free parameters







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

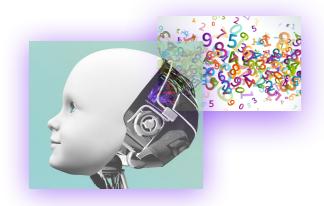




inaccurate both $p_{\sigma,\beta}$ 8 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor} ,



 $\beta_{\alpha}, \beta_{form}, \beta_{con}, \beta_{mor}$



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

 $p_{
ightharpoonup}$ 0 free parameters



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





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







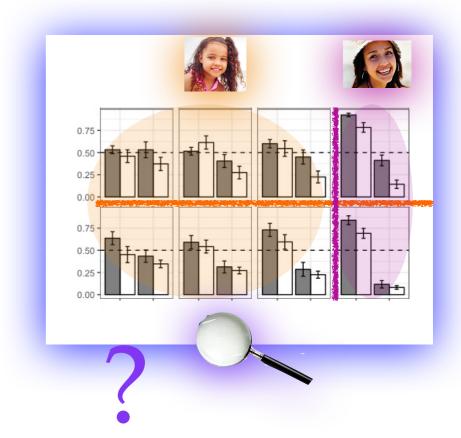
inaccurate both $p_{\sigma,\beta}$ 8 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor}

 $\beta_{\alpha}, \beta_{form}, \beta_{con}, \beta_{mor}$



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_{\searrow} 0 free parameters





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



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

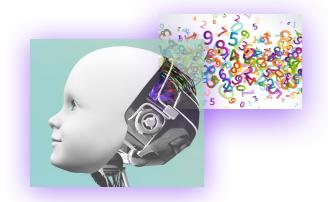






inaccurate both $p_{\sigma,\beta}$ 8 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor} ,

 $\beta_{\alpha}, \beta_{form}, \beta_{con}, \beta_{mor}$



 $-\infty \le \log(\text{likelihood}) \le 0$, 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

 $p_{
ightharpoonup}$ 0 free parameters



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





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

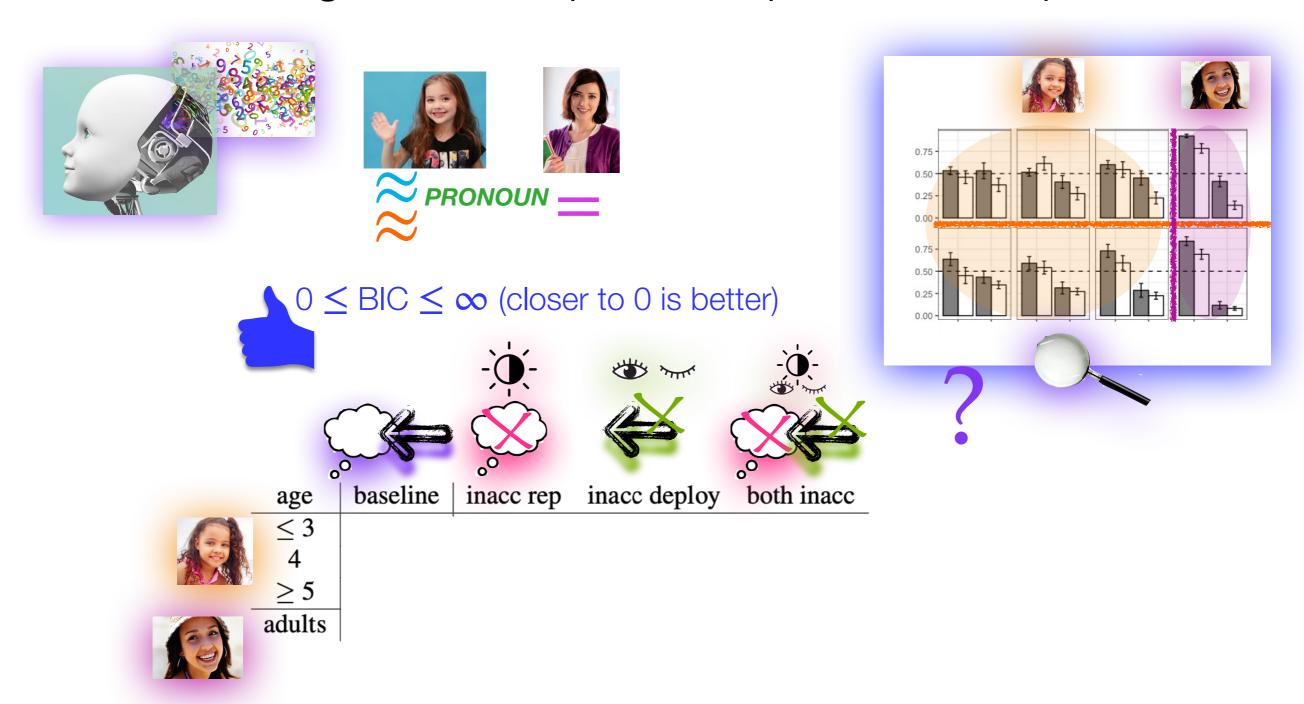


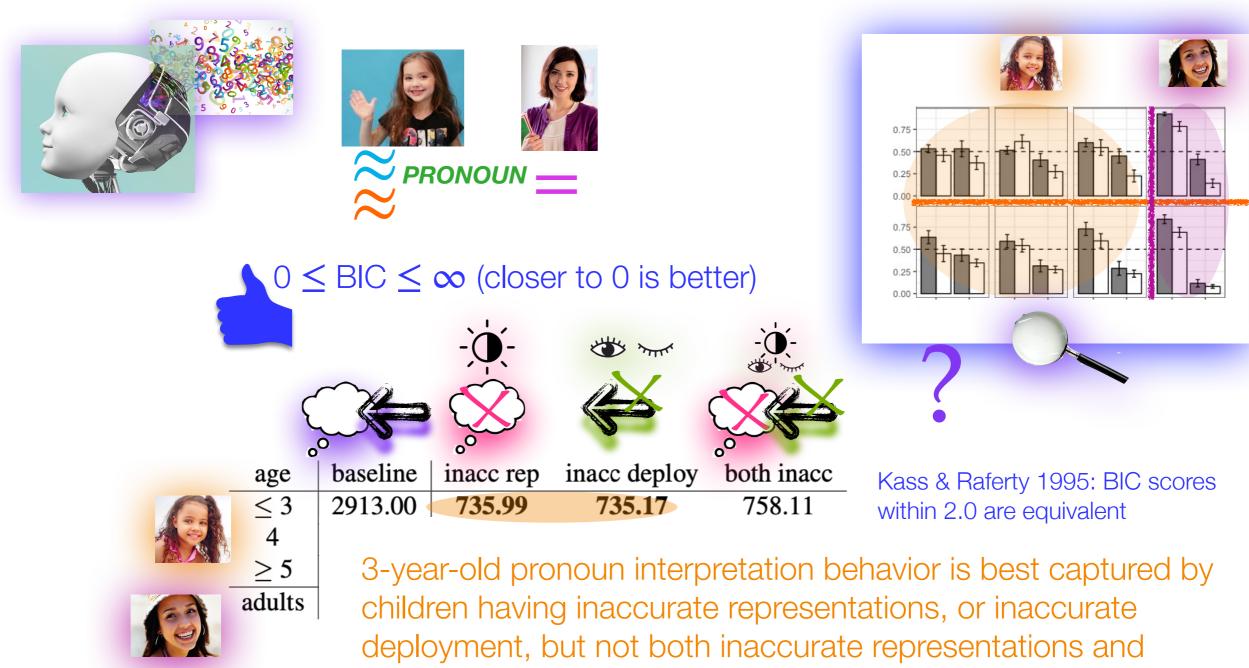




inaccurate both $p_{\sigma,\beta}$ 8 free parameters: σ_{α} , σ_{form} , σ_{con} , σ_{mor} ,

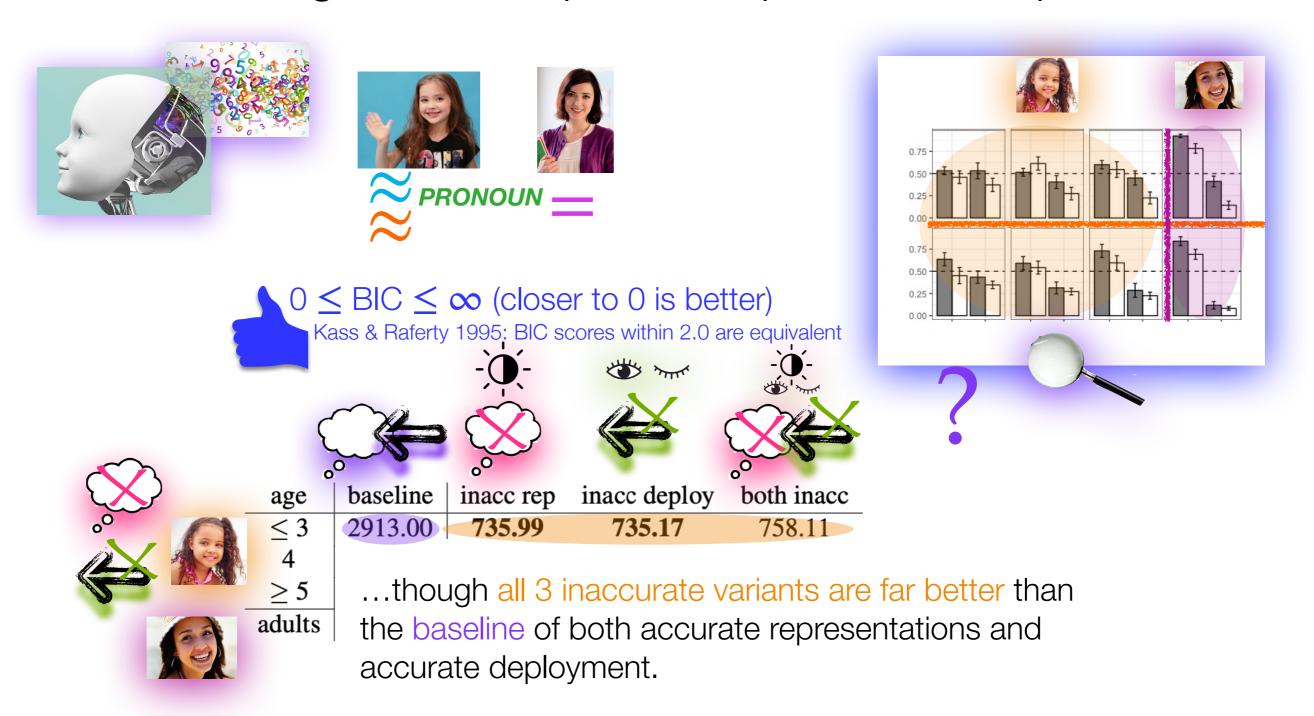
 $\beta_{\alpha}, \beta_{form}, \beta_{con}, \beta_{mor}$

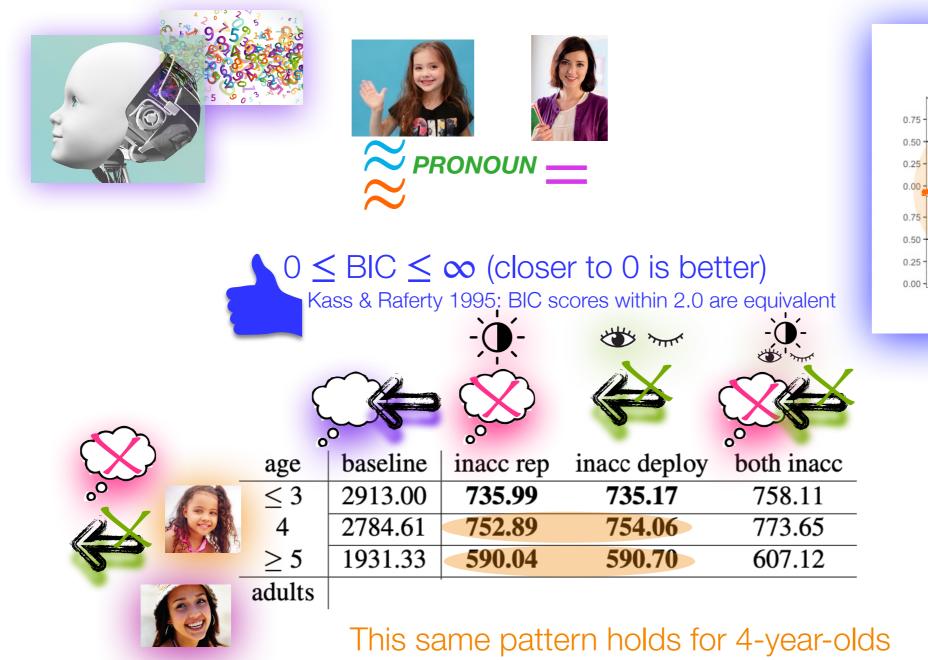




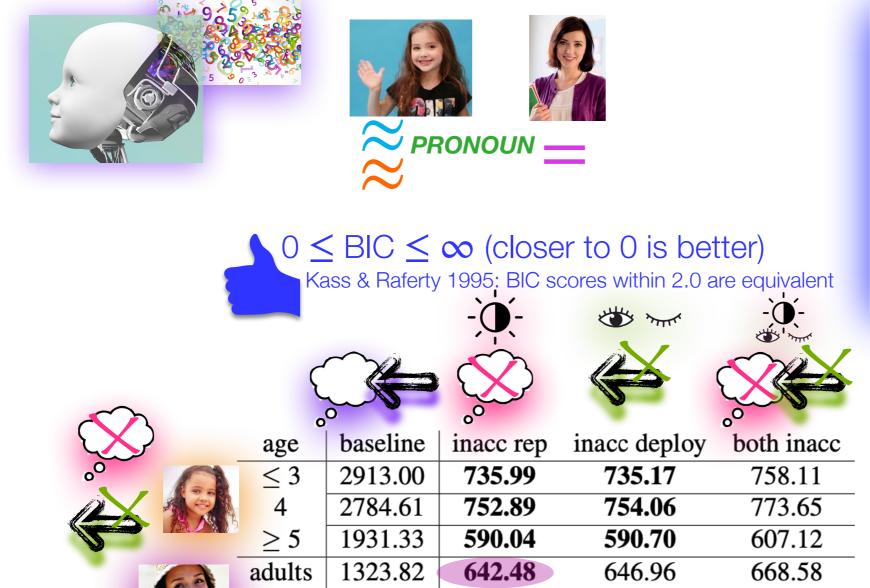
inaccurate deployment of those representations.







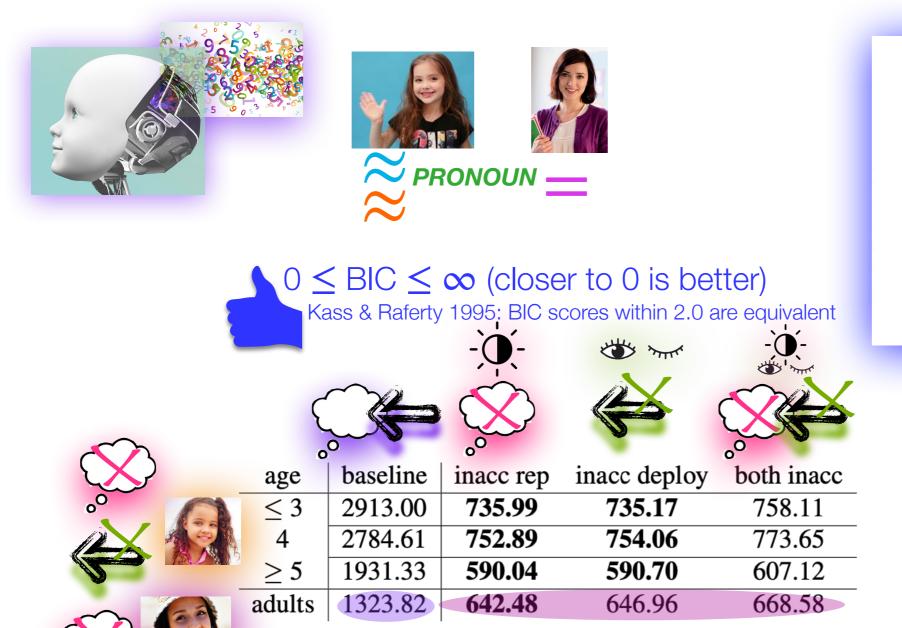
and 5-year-olds, too.

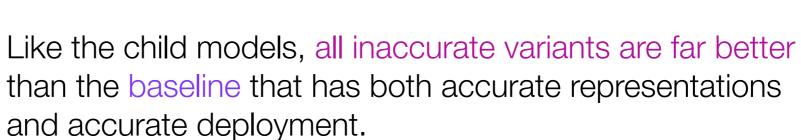


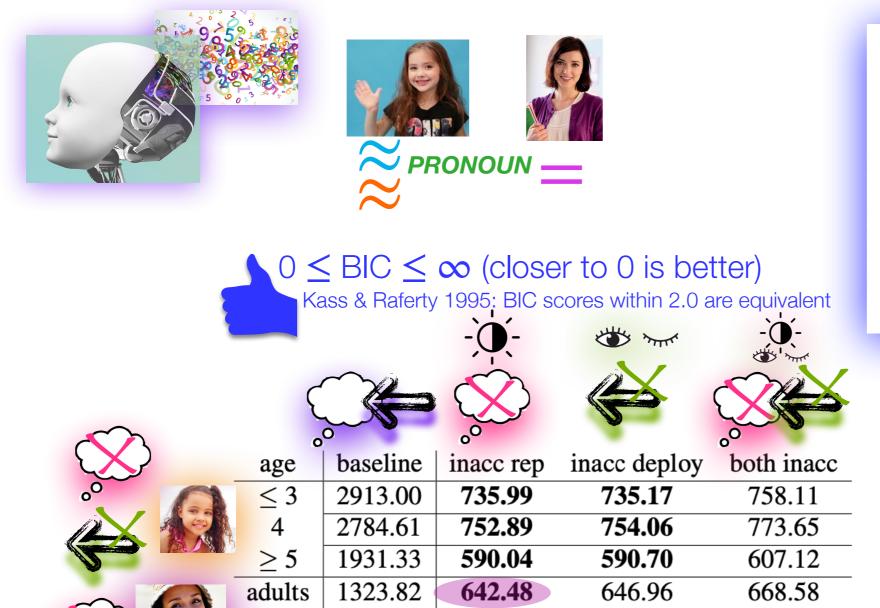


Adult pronoun interpretation behavior is best captured by the model that has inaccurate representations.











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



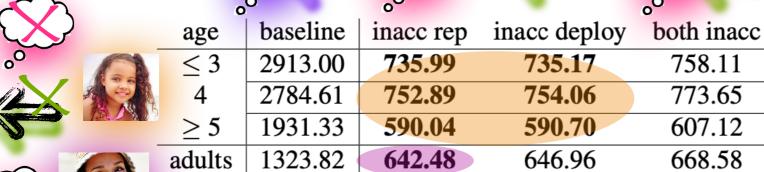


Kass & Raferty 1995: BIC scores within 2.0 are equivalent

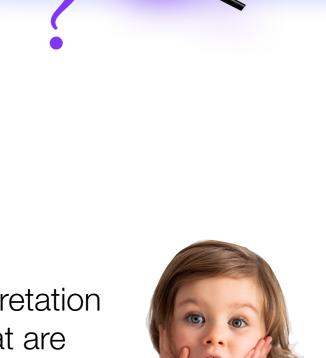








Takeaway: Both child and adult pronoun interpretation behavior are captured by modeled listeners that are inaccurate in some way.







Kass & Raferty 1995: BIC scores within 2.0 are equivalent





735.17

754.06

590.70

646.96

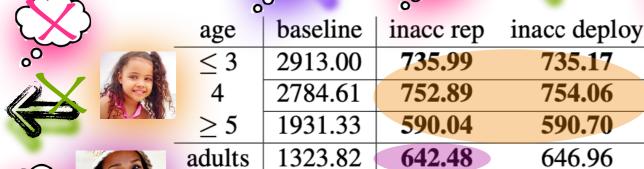


758.11

773.65

607.12

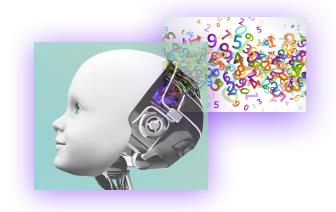
668.58





Becoming adult-like doesn't mean becoming accurate!





















O		O .		O
age	baseline	inacc rep	inacc deploy	both inacc
≤ 3	2913.00	735.99	735.17	758.11
4	2784.61	752.89	754.06	773.65
≥ 5	1931.33	590.04	590.70	607.12
adults	1323.82	642.48	646.96	668.58

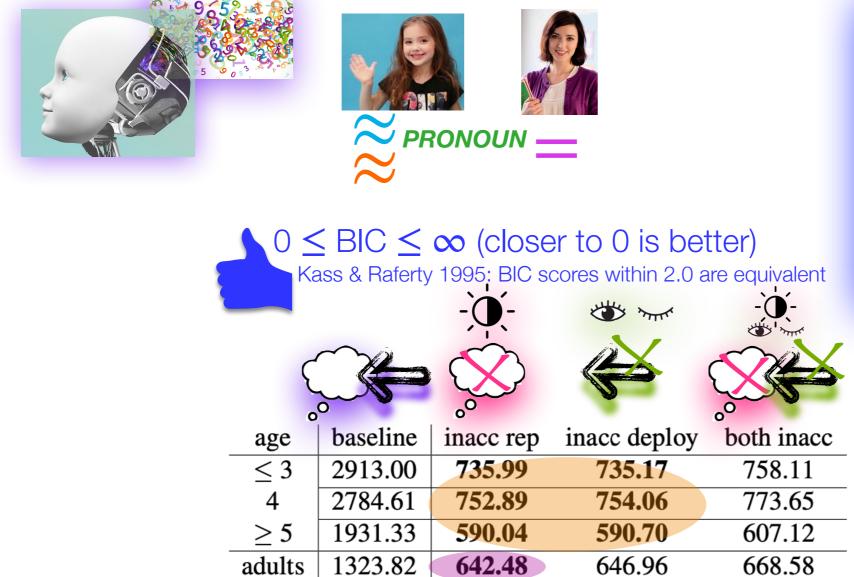


But what does it mean?

It means learning to become inaccurate in adult-like ways.









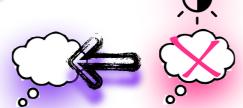
We know that adults inaccurately represent the available information.







Kass & Raferty 1995: BIC scores within 2.0 are equivalent







0		O		0
age	baseline	inacc rep	inacc deploy	both inacc
≤ 3	2913.00	735.99	735.17	758.11
4	2784.61	752.89	754.06	773.65
≥ 5	1931.33	590.04	590.70	607.12
adults	1323.82	642.48	646.96	668.58

So we can look at the σ values to see how the inaccurate representations are inaccurate — are they too smooth or too sharp?







age	baseline	inacc rep	inacc deploy	both inacc
≤ 3	2913.00	735.99	735.17	758.11
4	2784.61	752.89	754.06	773.65
≥ 5	1931.33	590.04	590.70	607.12
adults	1323.82	642.48	646.96	668.58

0.75 0.50 0.25 0.00 0.75 0.50 0.25 0.00

Development: How do these σ values compare to those that the best-fitting child model used that relied on inaccurate representations?







age	baseline	inacc rep	inacc deploy	both inacc
<u>≤</u> 3	2913.00	735.99	735.17	758.11
4	2784.61	752.89	754.06	773.65
≥ 5	1931.33	590.04	590.70	607.12
adults	1323.82	642.48	646.96	668.58





We know that adults accurately deploy their representations.







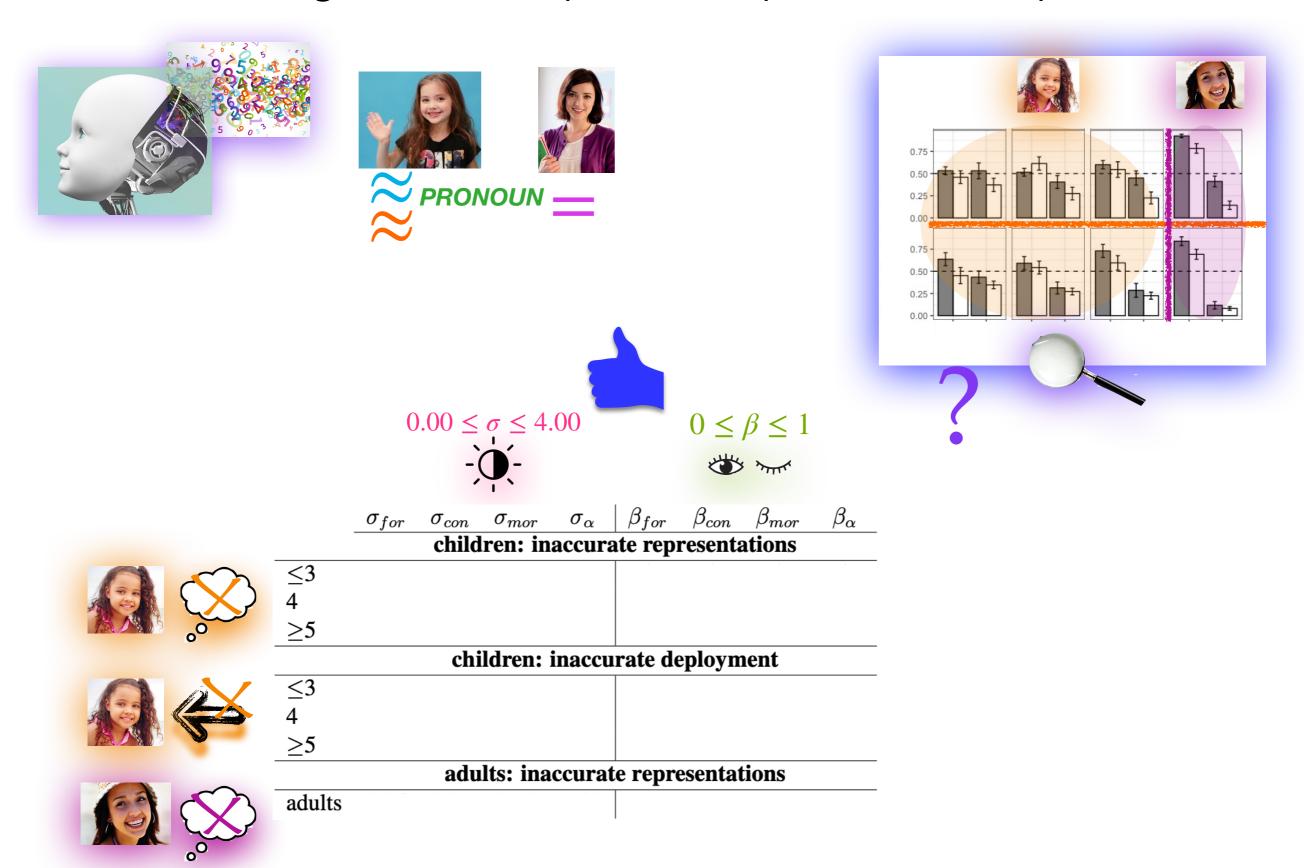
age	baseline	inacc rep	inacc deploy	both inacc
≤ 3	2913.00	735.99	735.17	758.11
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≥ 5	1931.33	590.04	590.70	607.12
adults	1323.82	642.48	646.96	668.58

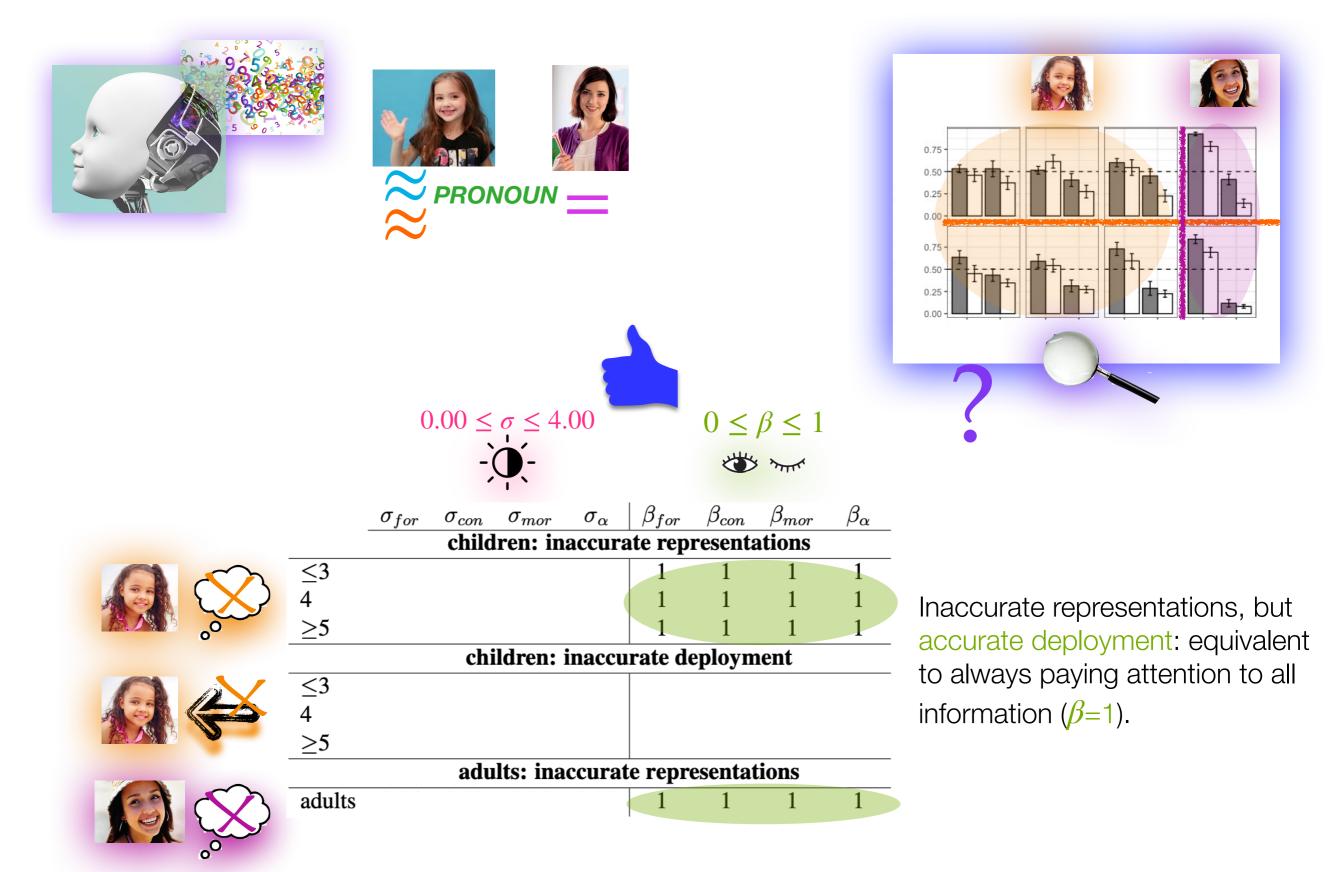


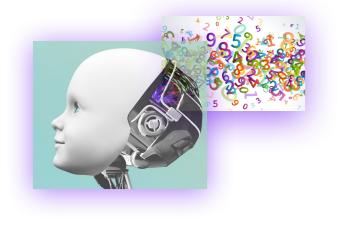


Development: How much <1 are the β values that the best-fitting child model used that relied on inaccurate deployment?

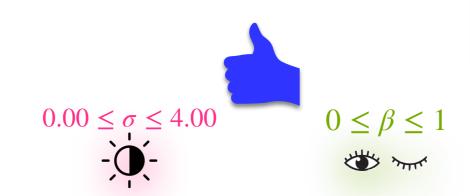












 σ_{lpha}

 eta_{for}

 β_{con} β_{mor}

		children: inaccurate representations								
		<u>≤3</u>	0.00	0.11	0.04	0.00	1	1	1	1
	(X)	4	0.00	0.01	0.09	0.00	1	1	1	1
		≥5	0.02	0.28	0.11	0.00	1	1	1	1
		children: inaccurate deployment								
		<u>≤3</u>								
		4								
		≥5								
100		adults: inaccurate representations								
198		adults	0.25	0.33	0.28	0.00	1	1	1	1
							'			

 σ_{con}

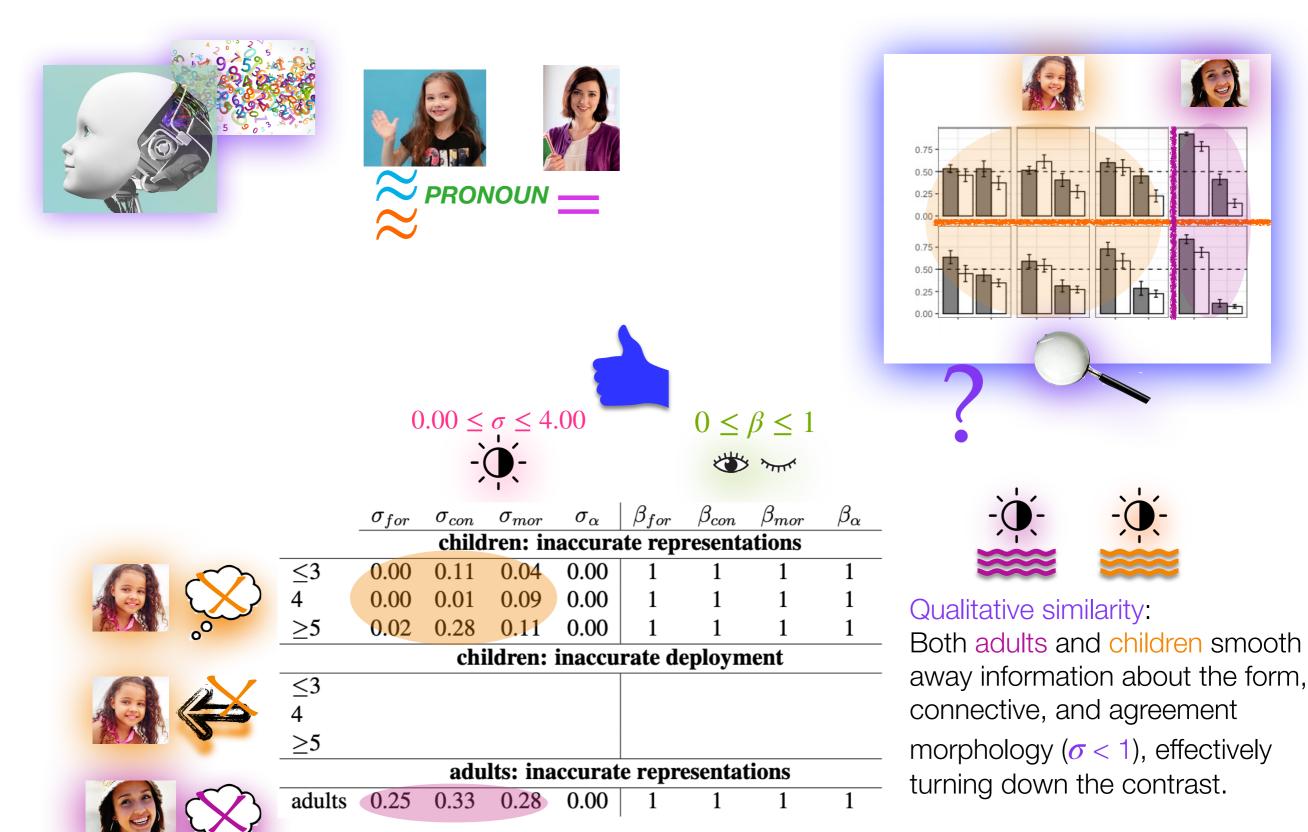
 σ_{mor}



Both adults and children completely smooth away information about the prior over possible antecedents (σ_{α}) .

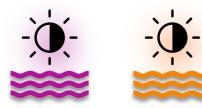
No change needed!

Note: This is functionally the same as never using this information (as if $\beta_{\alpha} = 0.00$).





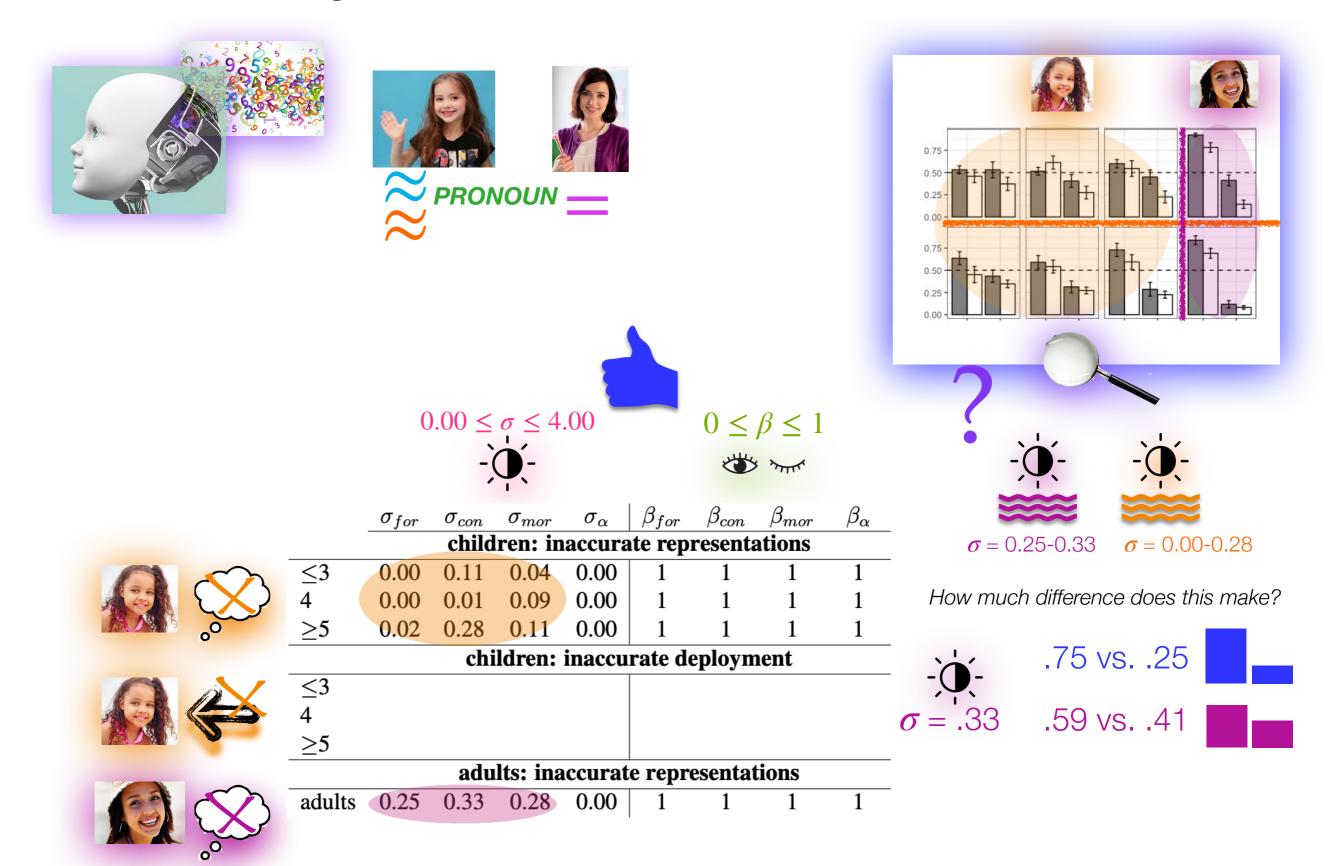


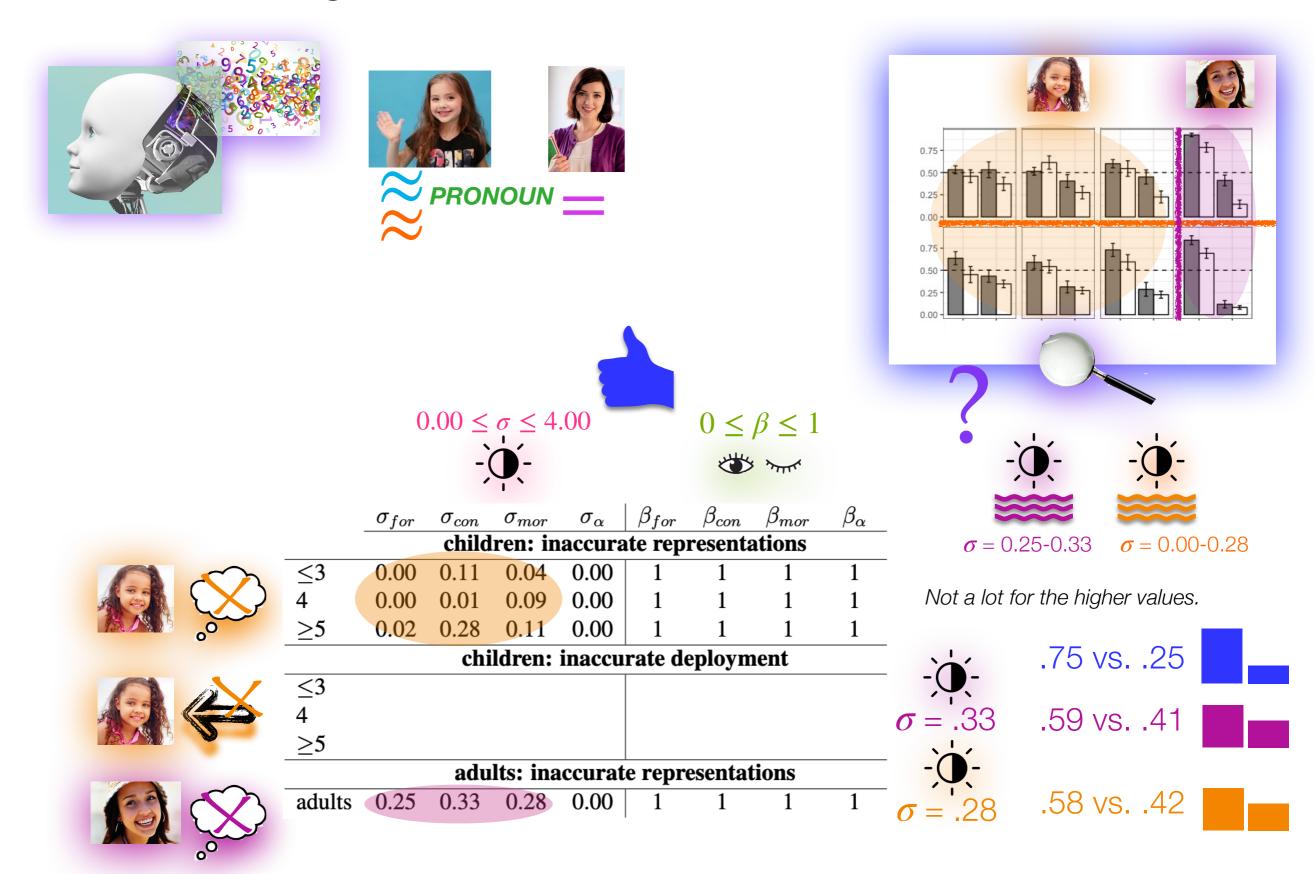


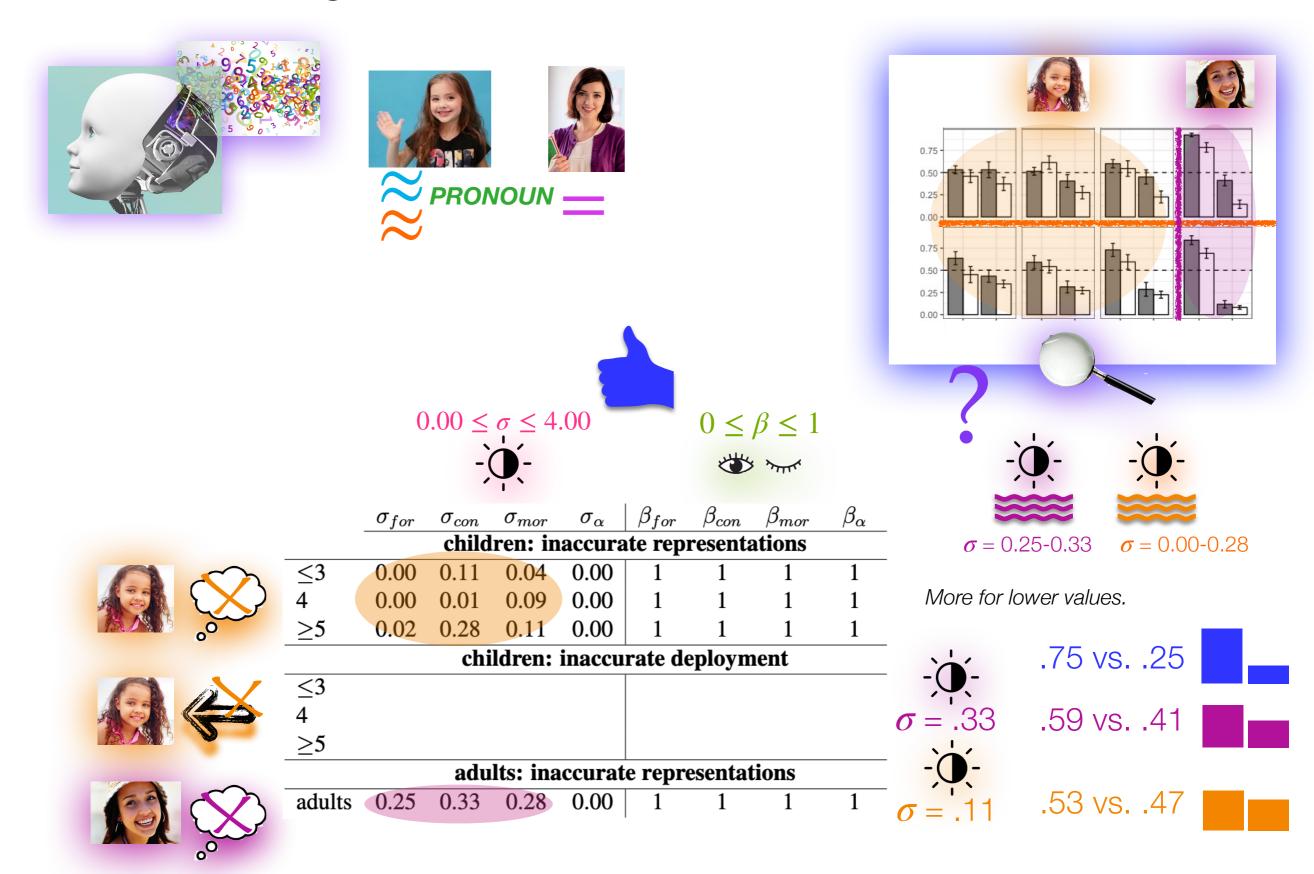
Quantitative differences:

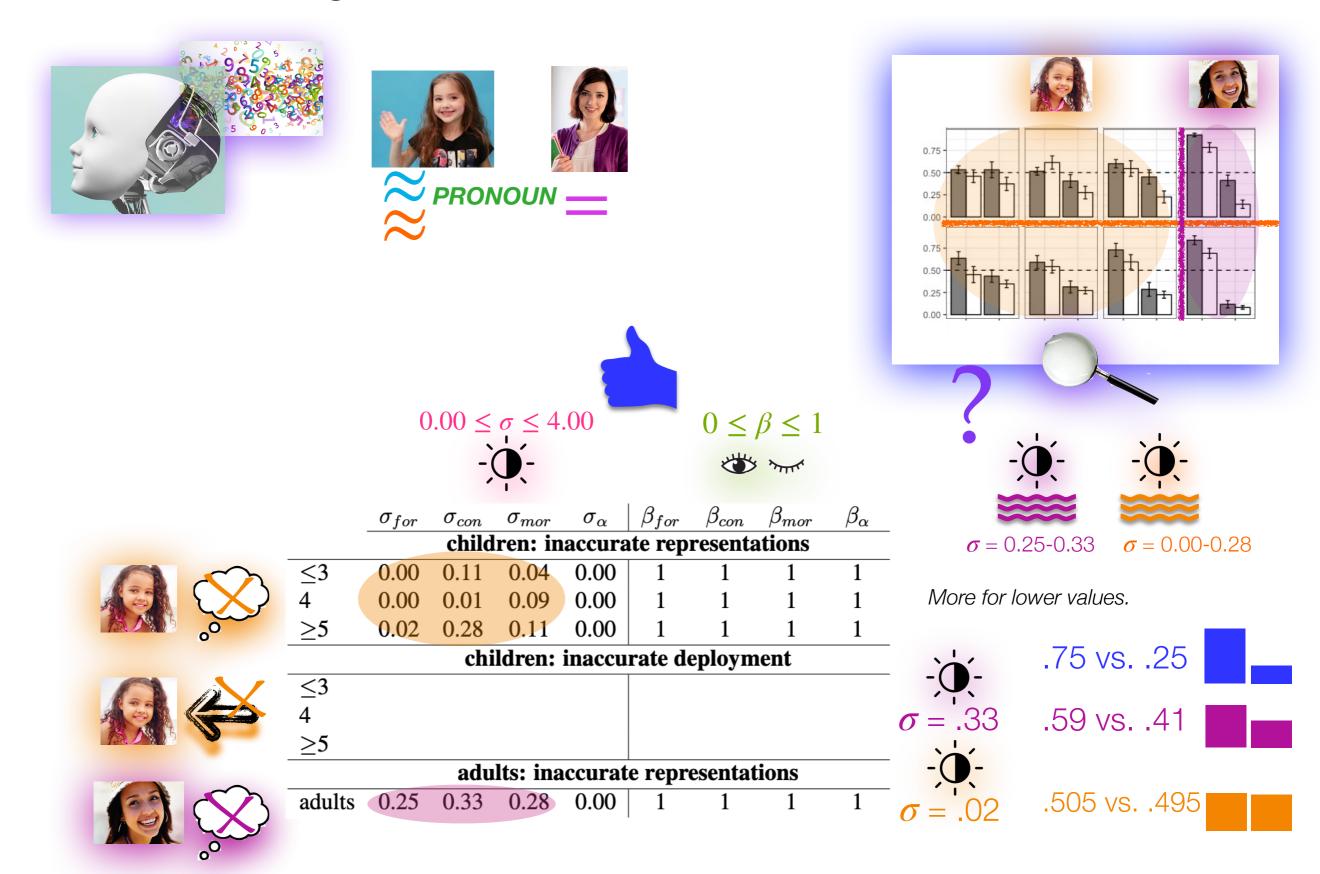
Children smooth more overall ($\sigma = 0.00$ -0.28) compared to adults ($\sigma = 0.25$ -0.33)

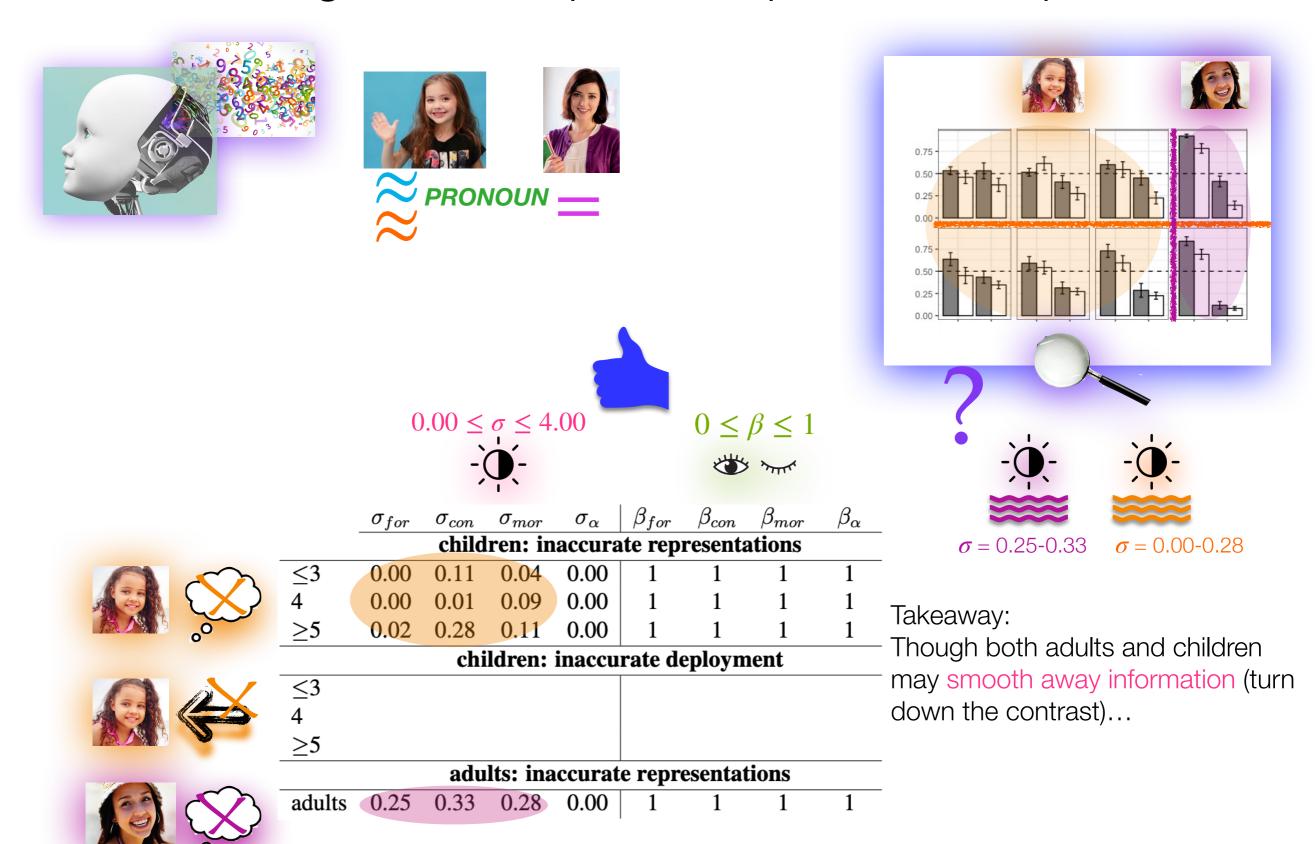
How much difference does this make?

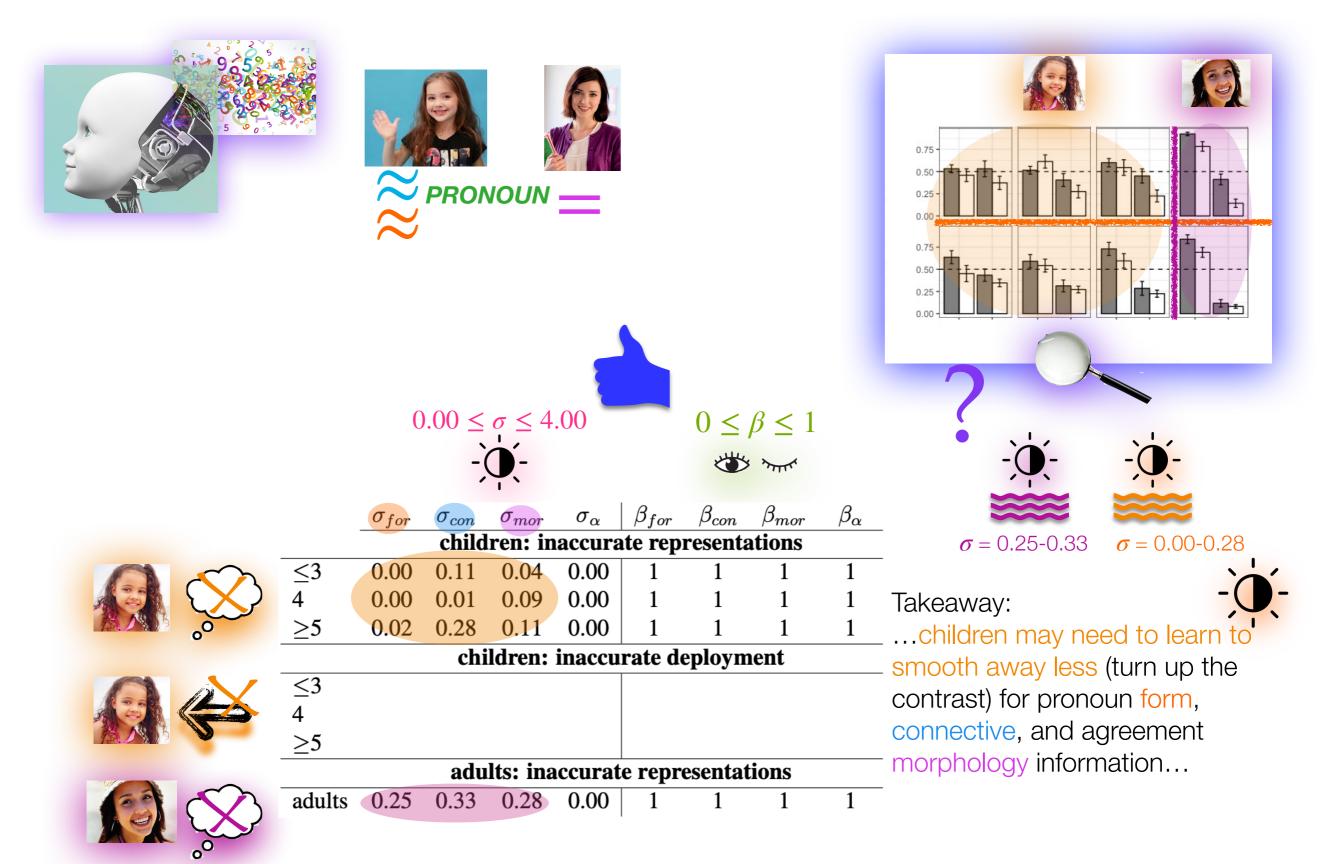


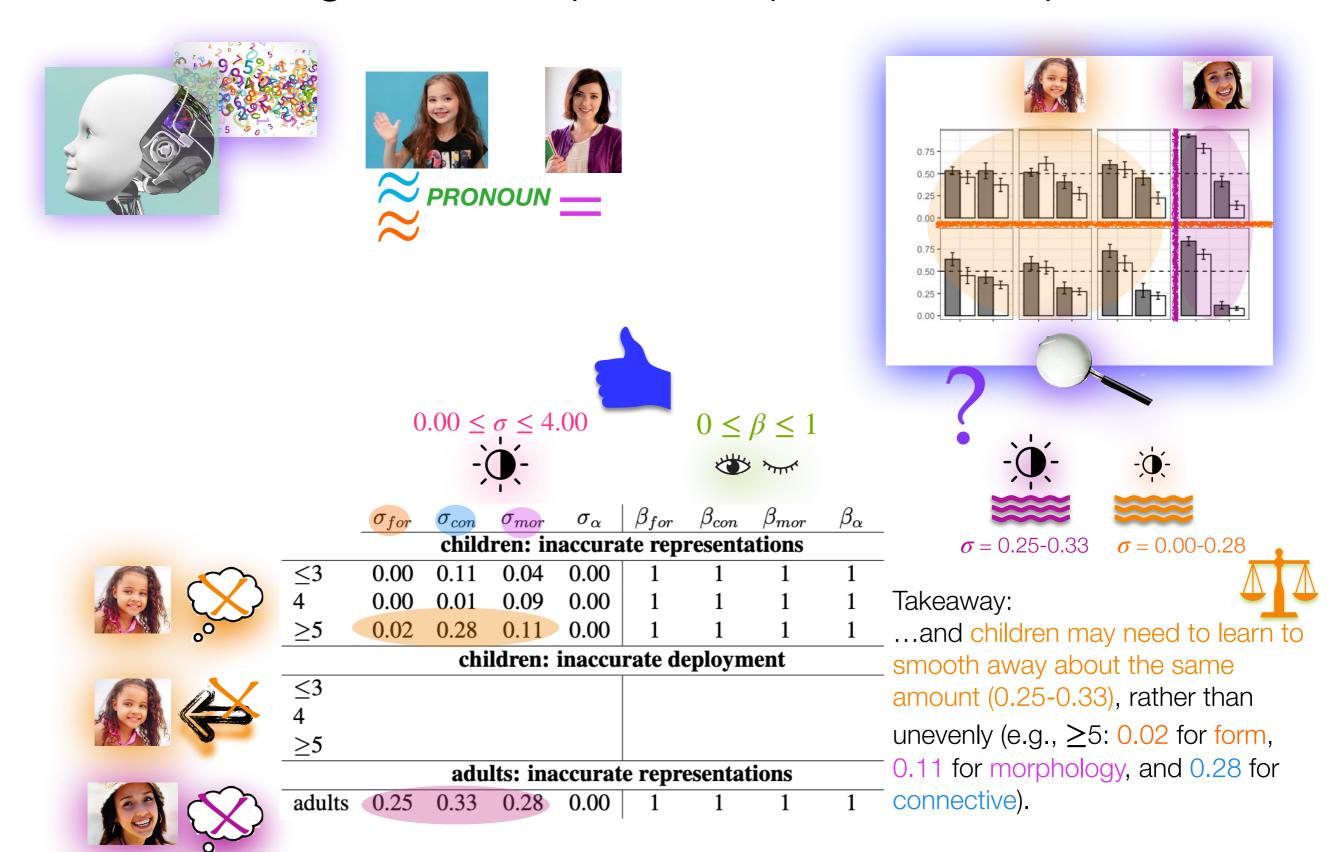








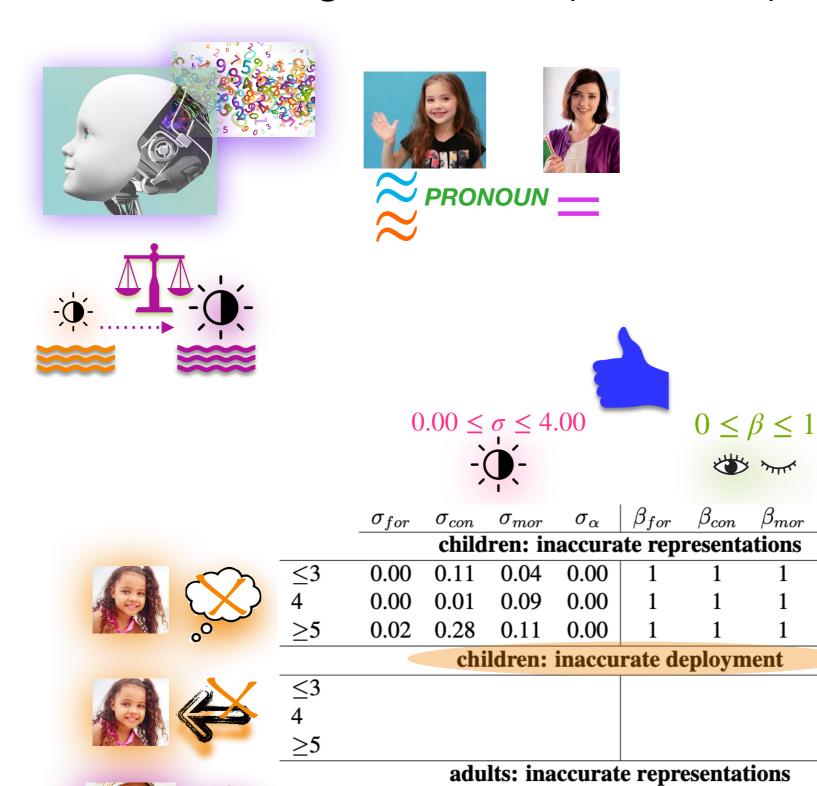




W mm

 β_{mor}

 eta_{lpha}



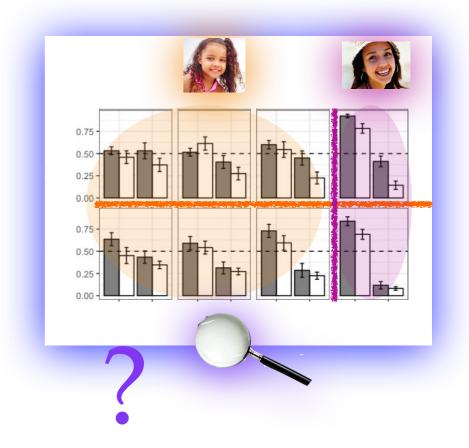
0.25

adults

0.33

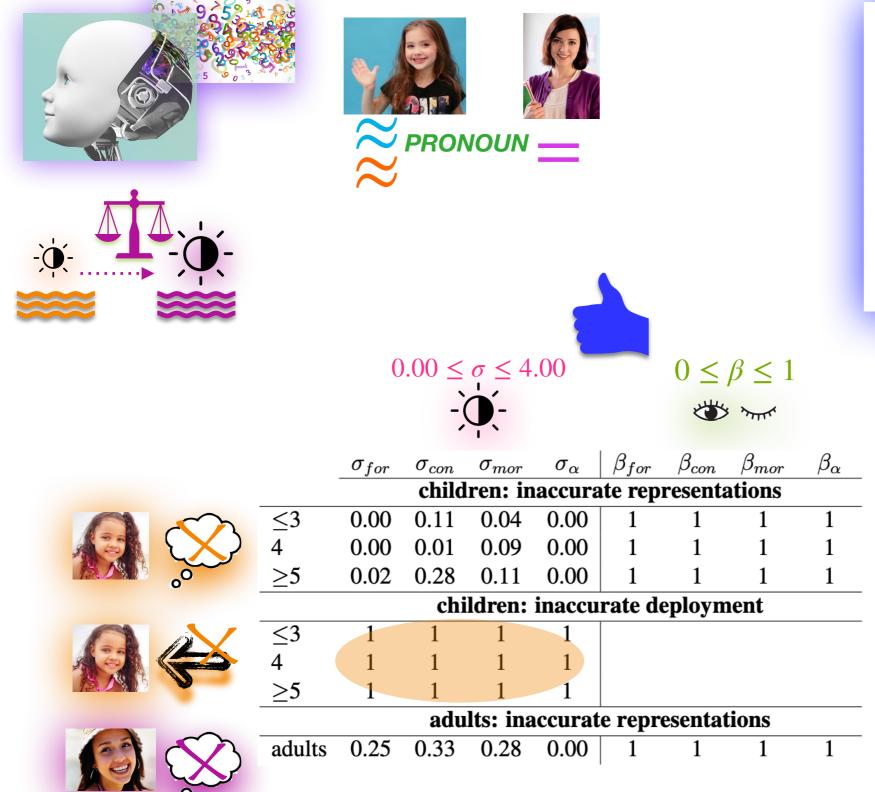
0.28

0.00



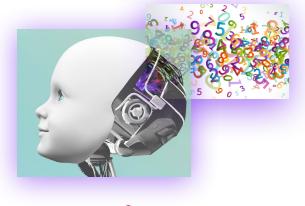
Or... it could be that children have accurate representations, but inaccurate deployment.

This would be a qualitative difference from adults.



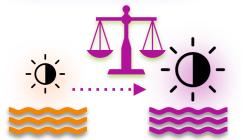


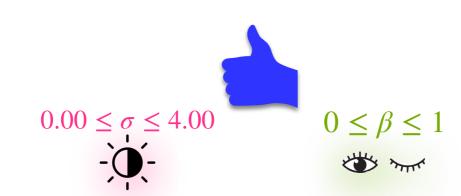
Accurate representations have σ =1.





 σ_{con} σ_{mor}





 β_{for}

 β_{con} β_{mor}

 eta_{lpha}





	children: inaccurate representations										
<u>≤3</u>	0.00	0.11	0.04	0.00	1	1	1	1			
4	0.00	0.01	0.09	0.00	1	1	1	1			
≥5	0.02	0.28	0.11	0.00	1	1	1	1			

 σ_{lpha}

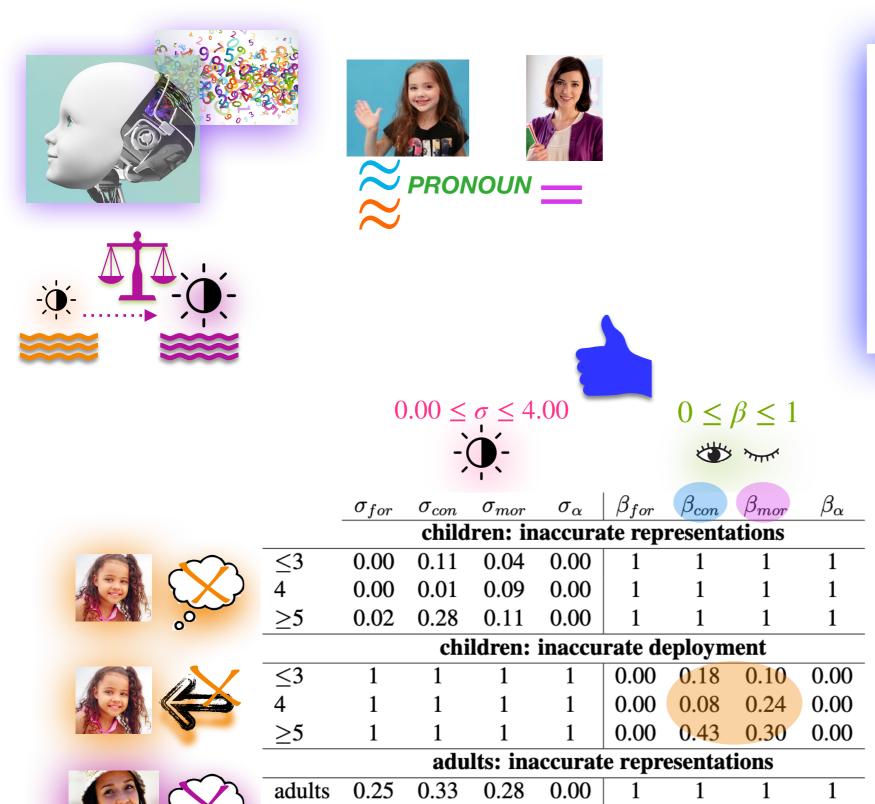


		chi	ldren:	inaccu	ırate de	eploym	ent	
<u>≤3</u>	1	1	1	1	0.00	0.18	0.10	0.00
4	1	1	1	1	0.00	0.08	0.24	0.00
≥5	1	1	1	1	0.00	0.43	0.30	0.00
			T				•	



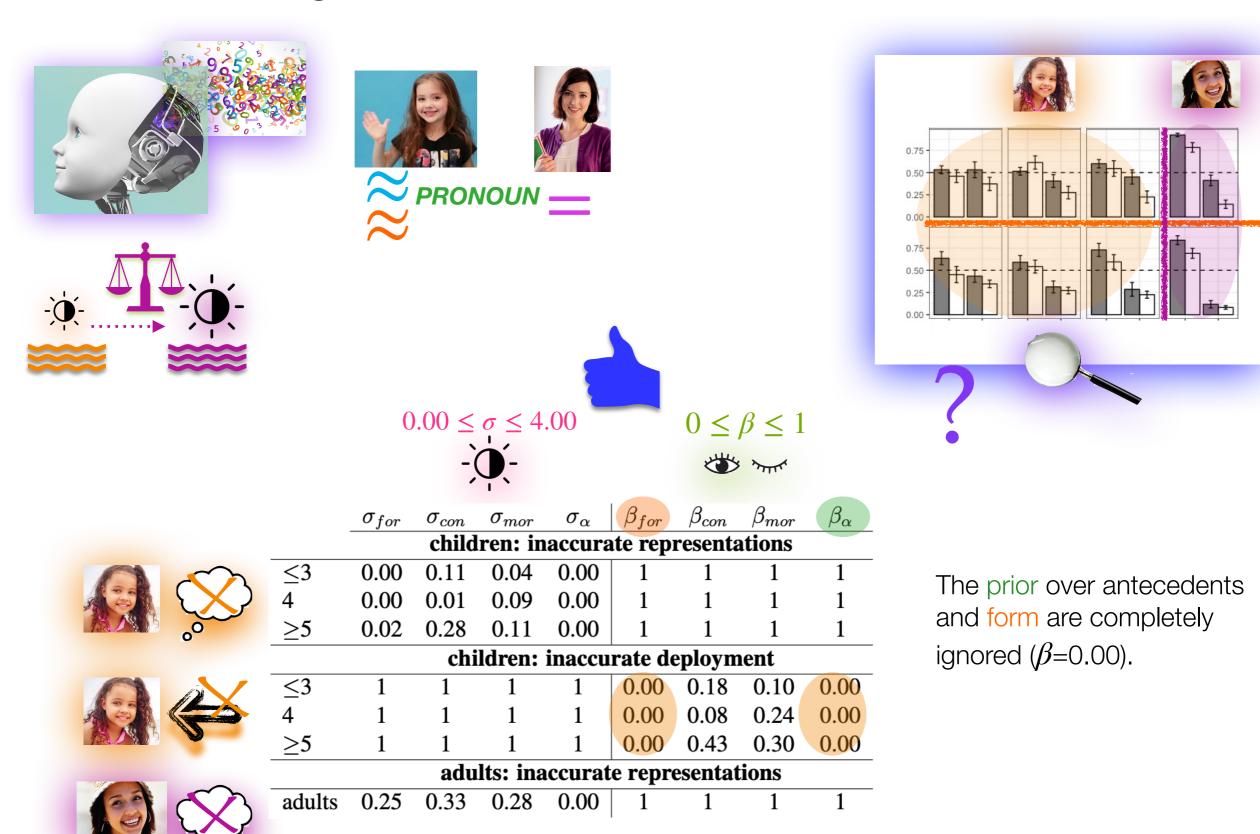
		adu	lts: ina	ccurat	e repr	esentat	ions	
adults	0.25	0.33	0.28	0.00	1	1	1	1

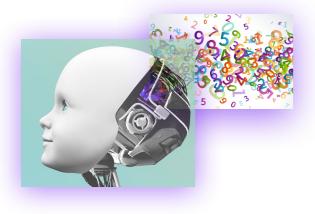
Inaccurate deployment varies across information types, though all are used <50%.





The connective and agreement morphology are heeded varying amounts.



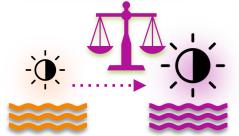


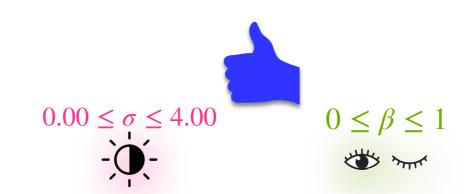


 σ_{con} σ_{mor}

0.25

adults











			child	ren: in	accura	ite rep	resenta	tions			
ĺ	≤ 3	0.00	0.11	0.04	0.00	1	1	1	1		
	4	0.00	0.01	0.09	0.00	1	1	1	1		
	≥5	0.02	0.28	0.11	0.00	1	1	1	1		
	children: inaccurate deployment										

 σ_{lpha}

 β_{for} β_{con} β_{mor}





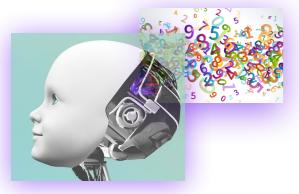
		CIII	iui cii.	maccu	i acc u	cproj m	CIIC	
≤3	1	1	1	1	0.00	0.18	0.10	0.00
4	1			1				
≥5	1	1	1	1	0.00	0.43	0.30	0.00
							_	

adults: inaccurate representations 0.28 0.33

Note: This is equivalent to smoothing away all information in the representation.

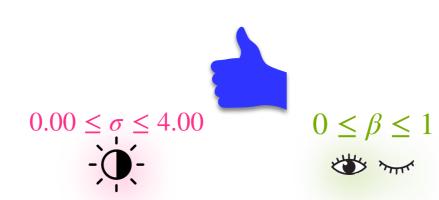


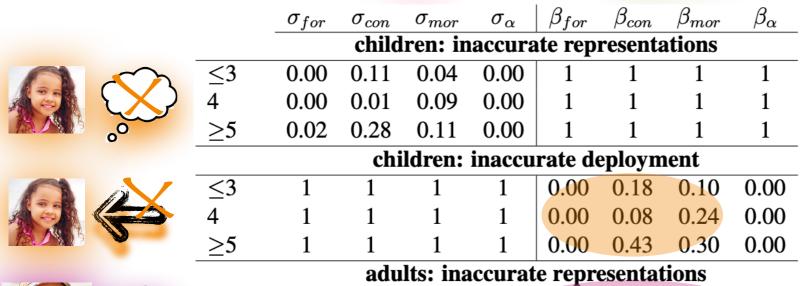
So, a β_{α} =0.00 accomplishes the same thing as a $\sigma_{\alpha}=0.00$, -0. which is adult-like. Similarity!











0.28

0.00

0.33

0.25

adults

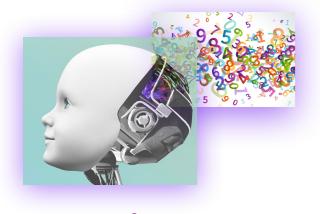


Takeaway:

1

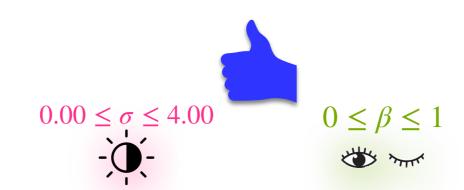
To become adult-like, children would need to switch to using most of their representations accurately (all the time)...

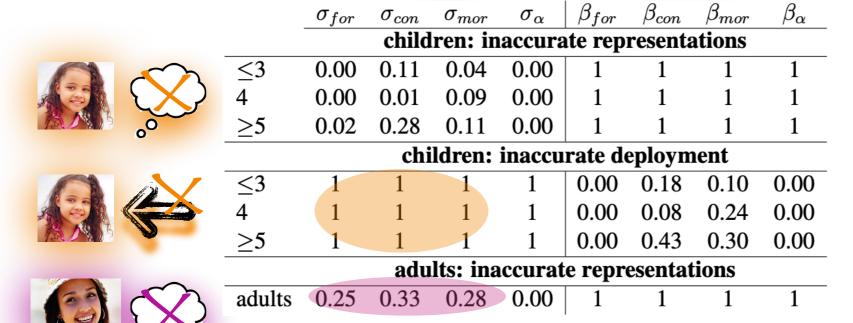








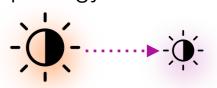






Takeaway:

...and smooth away information (turn down the contrast) for pronoun form, connective, and agreement morphology.





Takeaway:

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



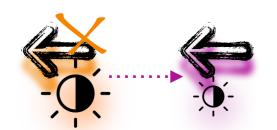




Takeaway:

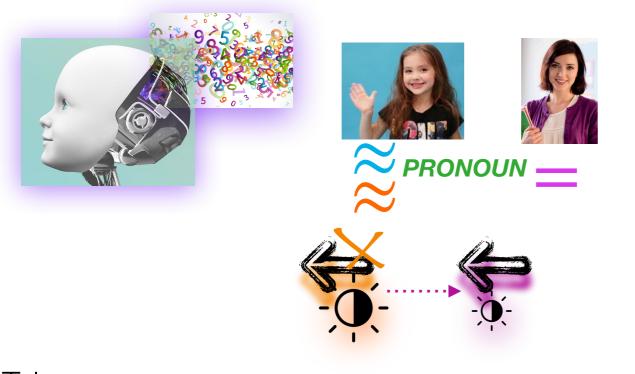
In one case, children are qualitatively different from adults in two basic ways:

- (1) They use accurate representations
- (2) They deploy those representations inaccurately





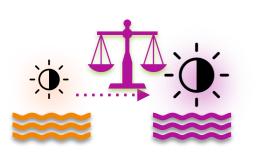
Development means changing both of these to be adult-like (turning down the contrast, always deploying information)



Takeaway:

In another case, children are qualitatively similar to adults:

- (1) They use inaccurate representations
- (2) They deploy those representations accurately





Development means learning to represent information inaccurately in an adult-like way (turning up the contrast equally)



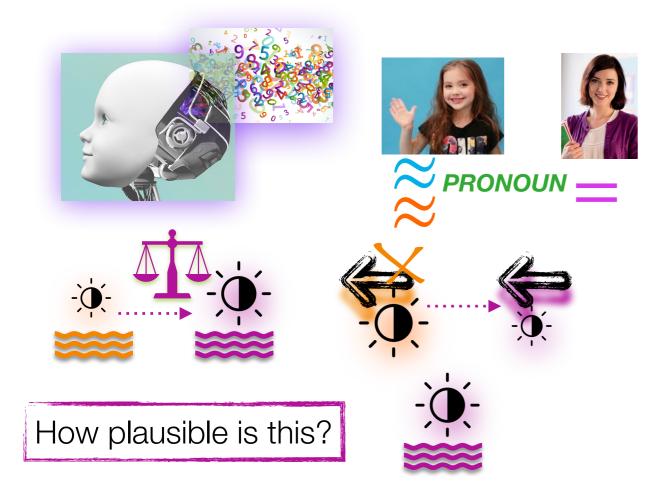
Takeaway:

Being adult-like doesn't mean being accurate! • Here, the best explanation for adult behavior is representations that are inaccurate because they smooth away information.



How plausible is this?

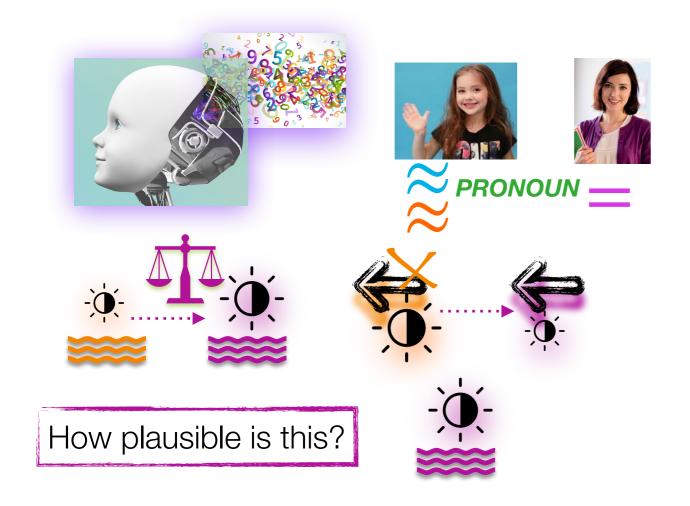




Adults smoothing away information:

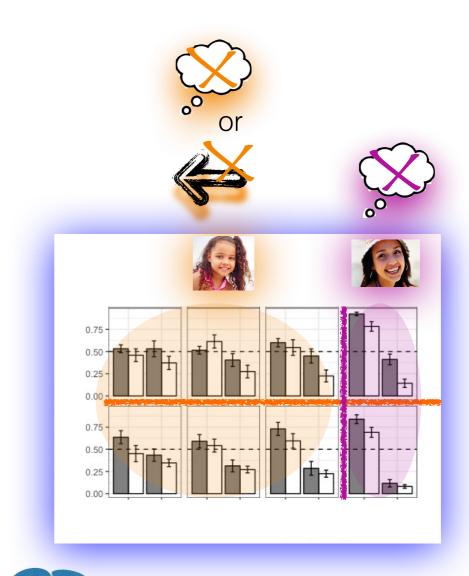
In adult decision-making studies (Kahneman & Tversky 1979), adults "interpret" representations, rather than using them accurately





Adults smoothing away information:

More recently, adult decision-making studies have found limits on the "dynamic range of the neural representation of probability" (Zhang & Maloney 2012, Zhang et al. 2020)



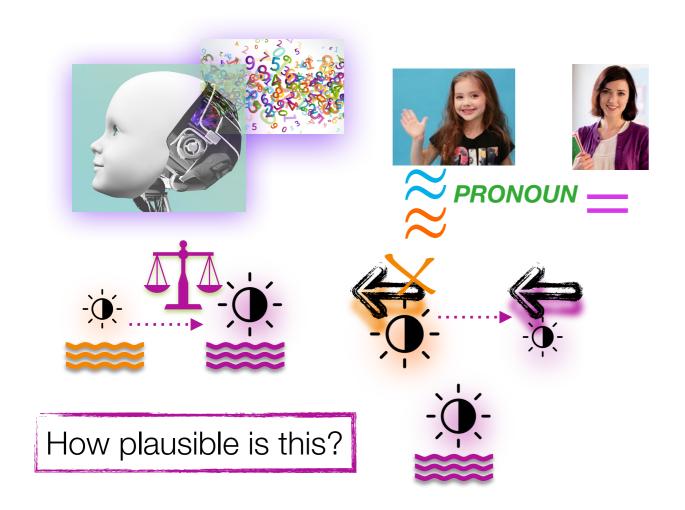
antecedent type

SUBJ

¬SUBJ

SG

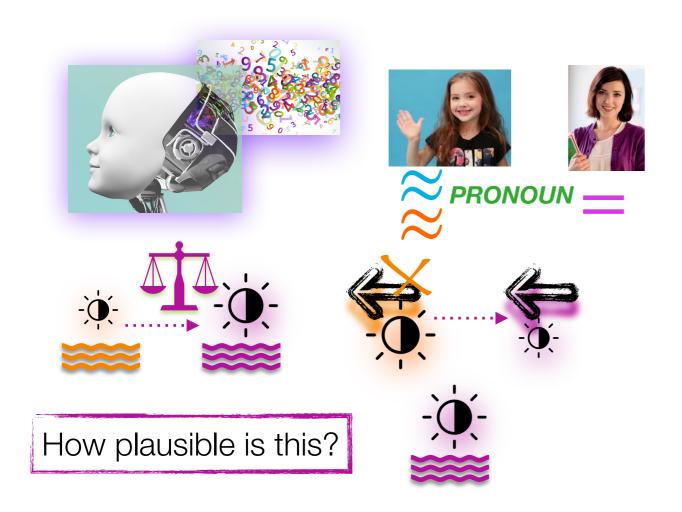
SG



Adults smoothing away information:

Zhang & Maloney 2012, Zhang et al. 2020: This neural limitation causes endpoints (near 0 and 1) to get smoothed away into something between 0.16 to 0.80

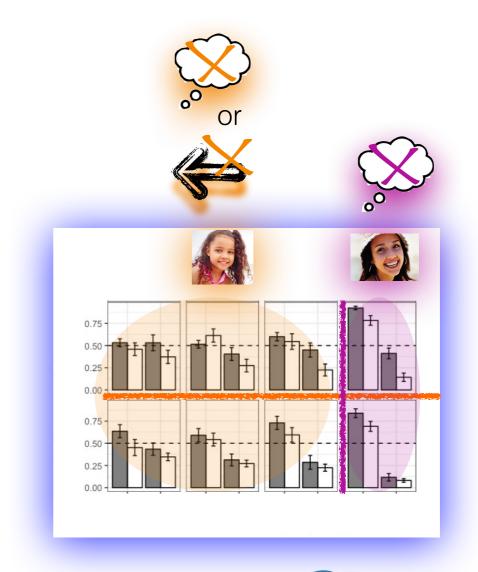
0.2	50	†	<u> </u>	± ±		
prior			m-	and the same		
$p(\alpha)$	p(FO)	$RM \alpha)$	()	Jan of	p(MC)	$OR \alpha)$
	Ø	overt	des	yme yme	SG	PL
0.362	0.938	0.062	0.324	0.676	0.998	0.002
0.071	00	016	0.750	0.250	0.005	0.995
0.438	0.8	0.183	0.132	0.868	0.998	0.002
0.129	0.959	0.041	0.394	0.606	0.005	0.995



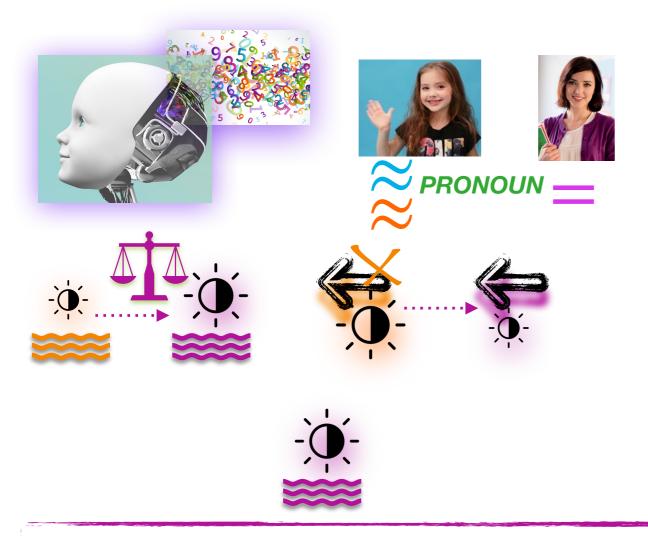


Zhang & Maloney 2012, Zhang et al. 2020: ...which is what these sigma values do.

	σ_{for}	σ_{con}	σ_{mor}	
adults	0.25	0.33	0.28	



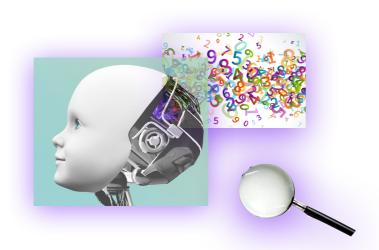
		prior	11 my					
		$p(\alpha)$	$p(FORM \alpha)$		and many		$p(\text{MOR} \alpha)$	
antecede	nt type	$P(\alpha)$	Ø	overt	des	yne yne	SG	PL
CIIDI	SG	0.362	0.938	0.062	0.324	0.676	0.998	0.002
SUBJ	PL	0.071	00	016	0.750	0.250	0.005	0.995
-CIIDI	SG	0.438	0.8	0.183	0.132	0.868	0.998	0.002
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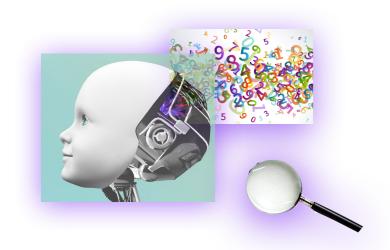
So actually, adults having inaccurate representations where they smooth away information this amount is pretty plausible!

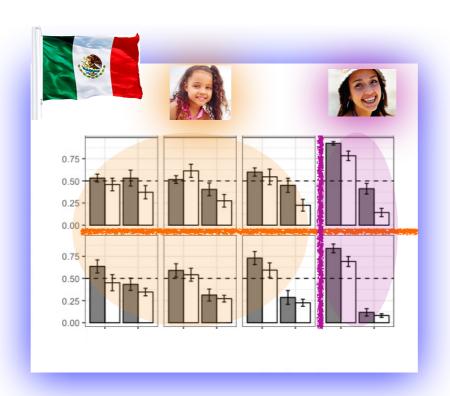




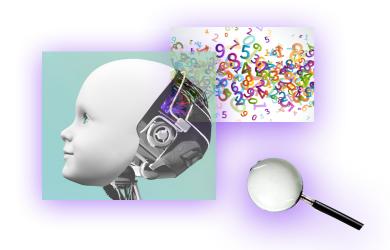


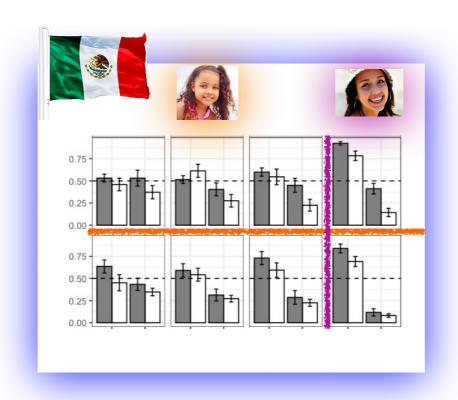
More generally, this case study demonstrates how we can use computational cognitive modeling...



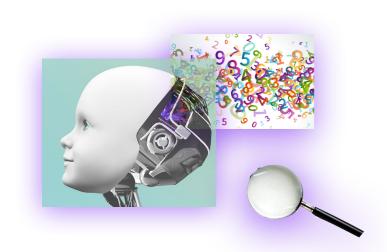


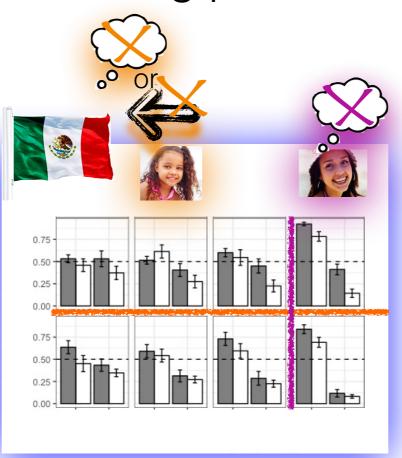
More generally, this case study demonstrates how we can use computational cognitive modeling, grounded in empirical data...





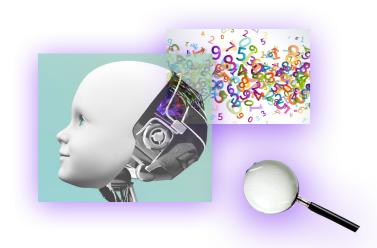
More generally, this case study demonstrates how we can use computational cognitive modeling, grounded in empirical data, to better understand how children and adults...





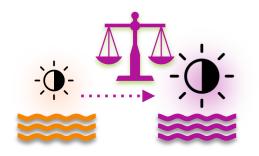
More generally, this case study demonstrates how we can use computational cognitive modeling, grounded in empirical data, to better understand how children and adults can solve complex linguistic tasks (like interpreting pronouns in a context with multiple, potentially conflicting, cues).

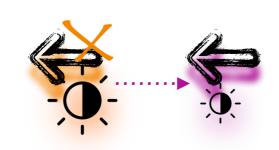
PRONOUN —

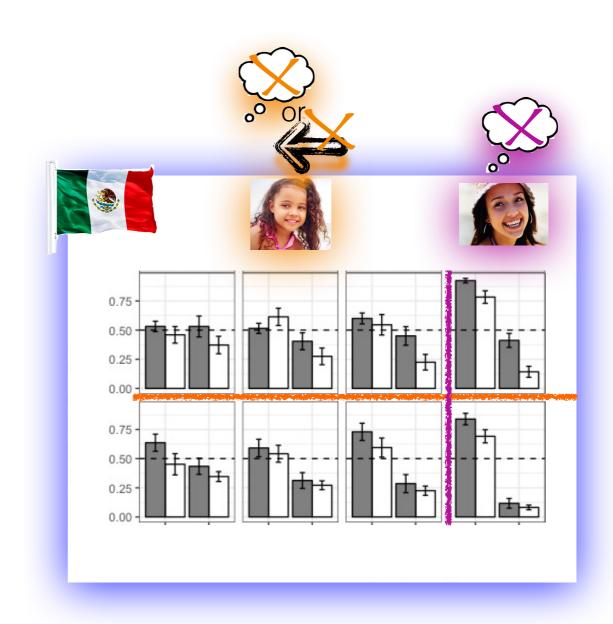




This helps us better understand what children may need to do to become adult-like (and it seems to be about learning to be inaccurate the adult way).





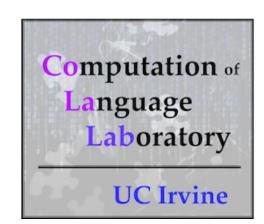


Thank you!

Hannah Forsythe



UMaryland CLIP Colloquium 2021 UCI QuantLang Collective







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