

# Simplicity, data and inter-related systems: A computational account of kinship term acquisition

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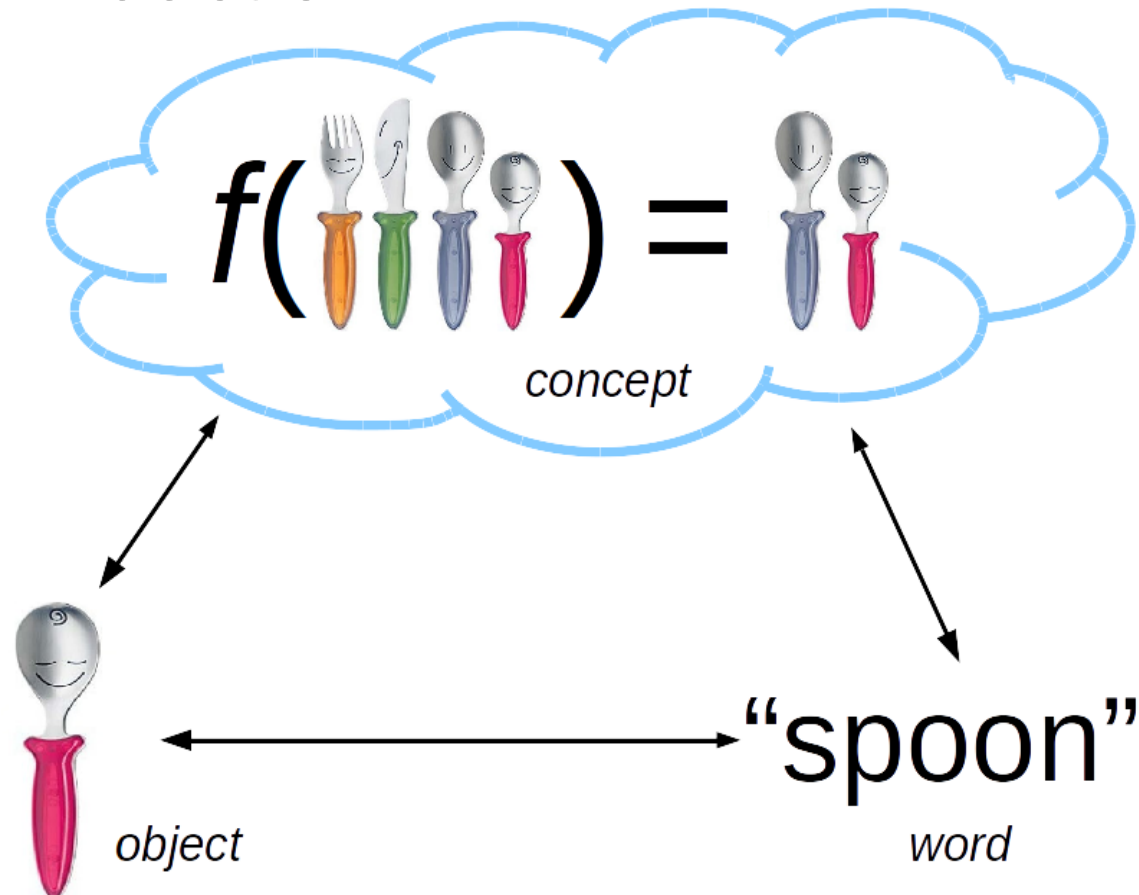


# Children's early verbal knowledge is “off”

- What's an uncle?
  - “He's a man.” (5;4 yo)
- What's a mother?
  - “A big girl” (4;6 yo)
- What's an aunt?
  - “It's a lady that helps you a lot of times . . .” (6;2 yo)
- What's an uncle?
  - “Your mother or father's brother.” (22;2 yo)

# Our Approach

- Formalize conceptual development as logical program induction.



# Ideal Learner Model

- Computational Level Analysis (Marr, 1982)
  - Specify a **Hypothesis Space** of concepts
  - Specify a **Prior** over hypotheses
  - Specify a **Likelihood** function
  - Specify the environment
- In a Bayesian learning model, learning corresponds to the movement of probability mass over a hypothesis space.

# Hypothesis Space & Prior

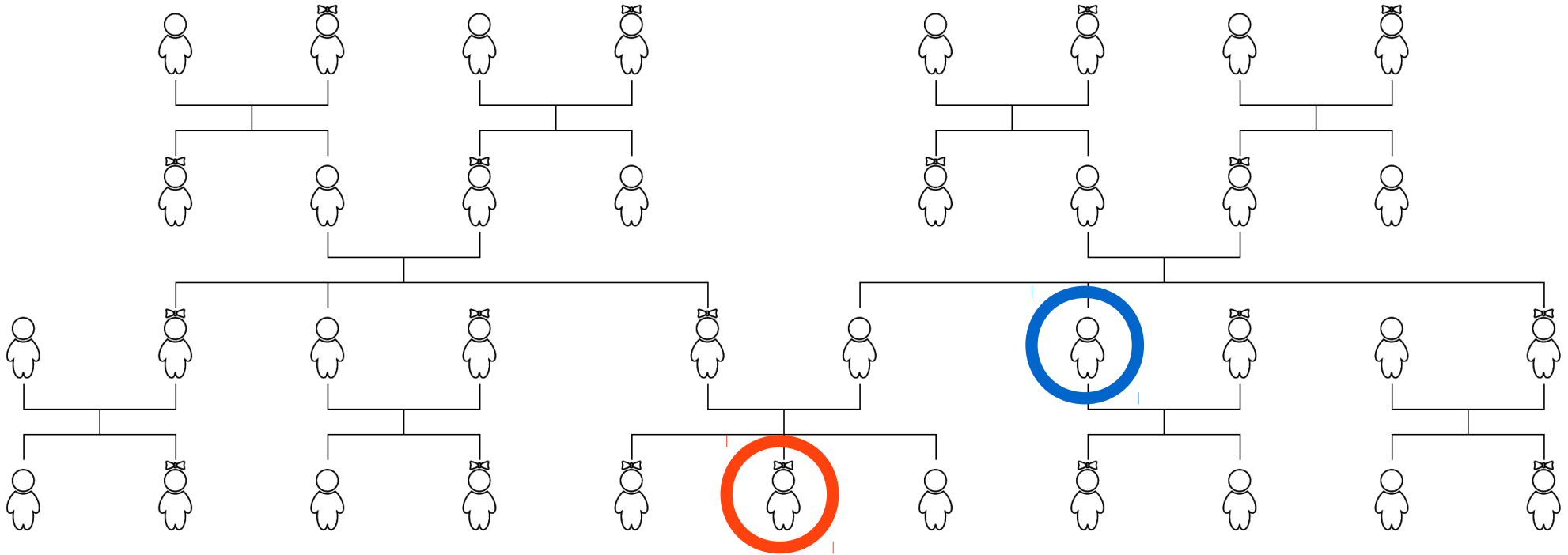
Tree Moving	Set Operations	Gender	Age	Inputs
Child Parent Spouse	Union Intersection Difference Complement	Female Male SameGender	SameGeneration ParentGeneration GparentGeneration	All Speaker Individual

For example:

1. All
2. Tito
3. SameGeneration(Tito)
4. Male(Parent(Speaker))

# Where does data come from?

- Context:



- Data Point:

- Context

- Word *uncle*

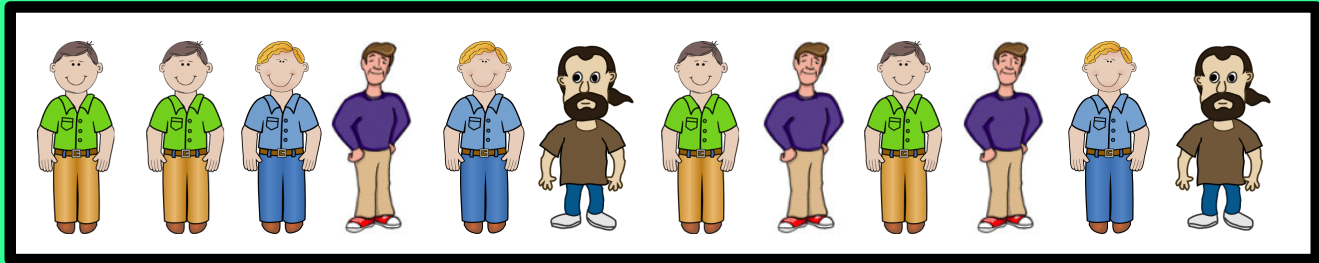
- Speaker ○

- Referent ○

# How do we fit to the data?

- **Size Principle Likelihood** (e.g., Tenenbaum & Griffiths, 2001; Xu & Tenenbaum, 2007)

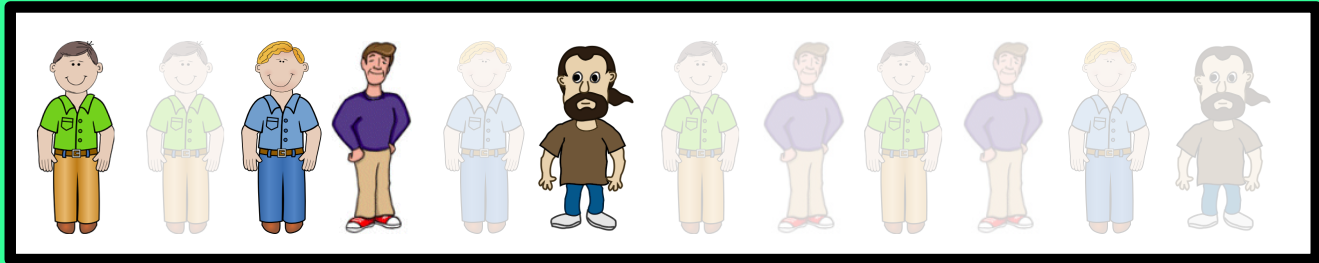
Data  
Distribution:



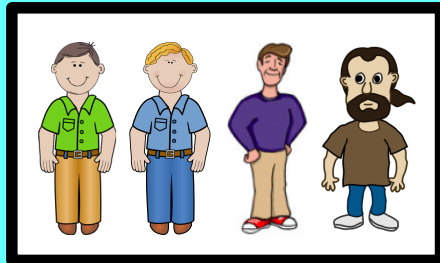
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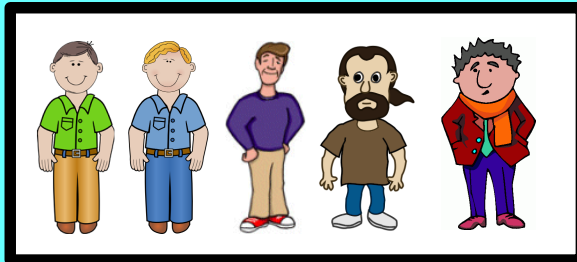
Data  
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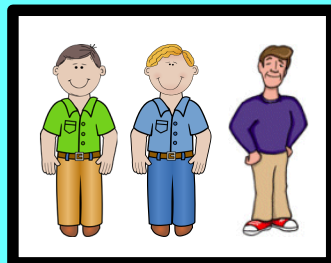
Hypothesis A:



Hypothesis B:



Hypothesis C:

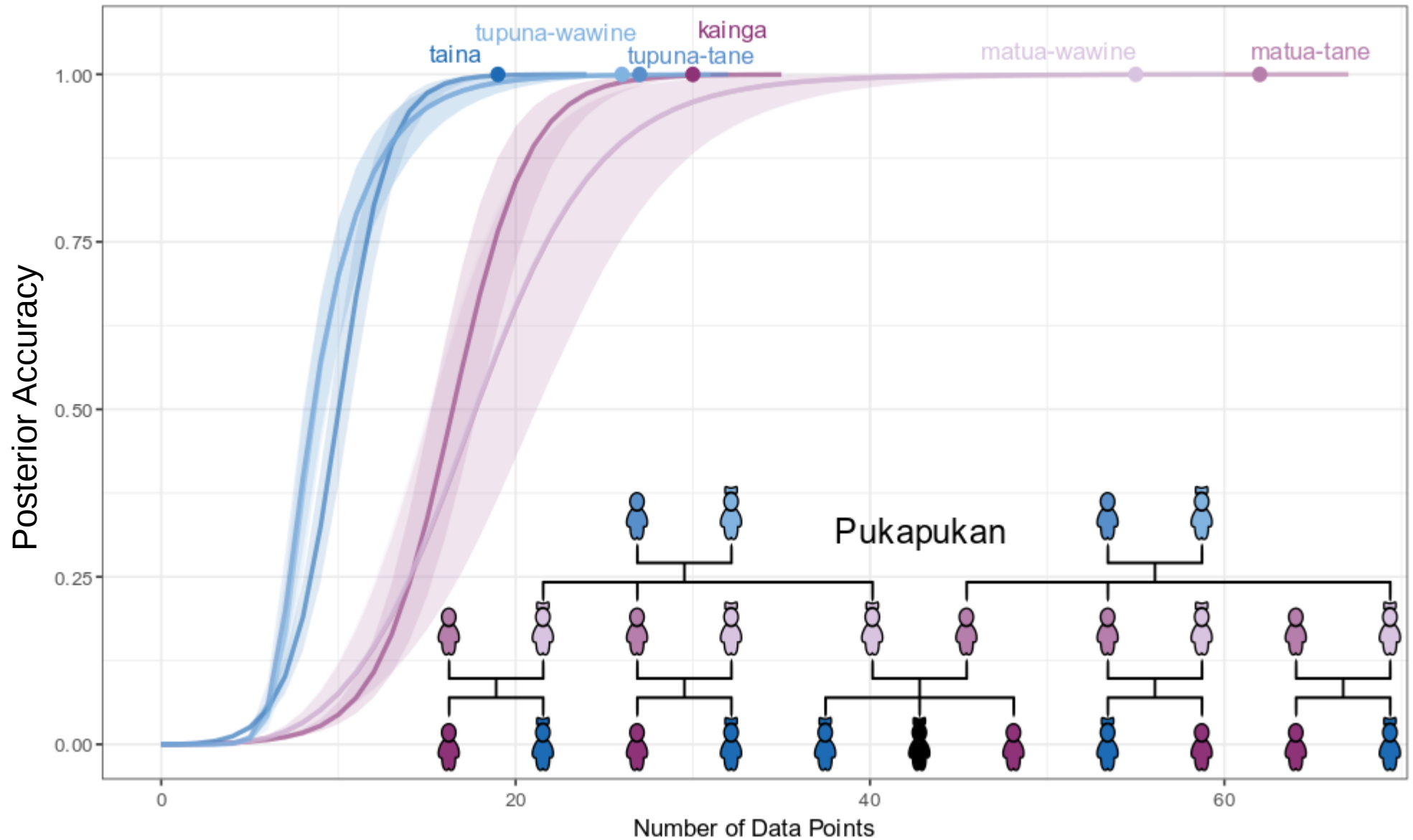




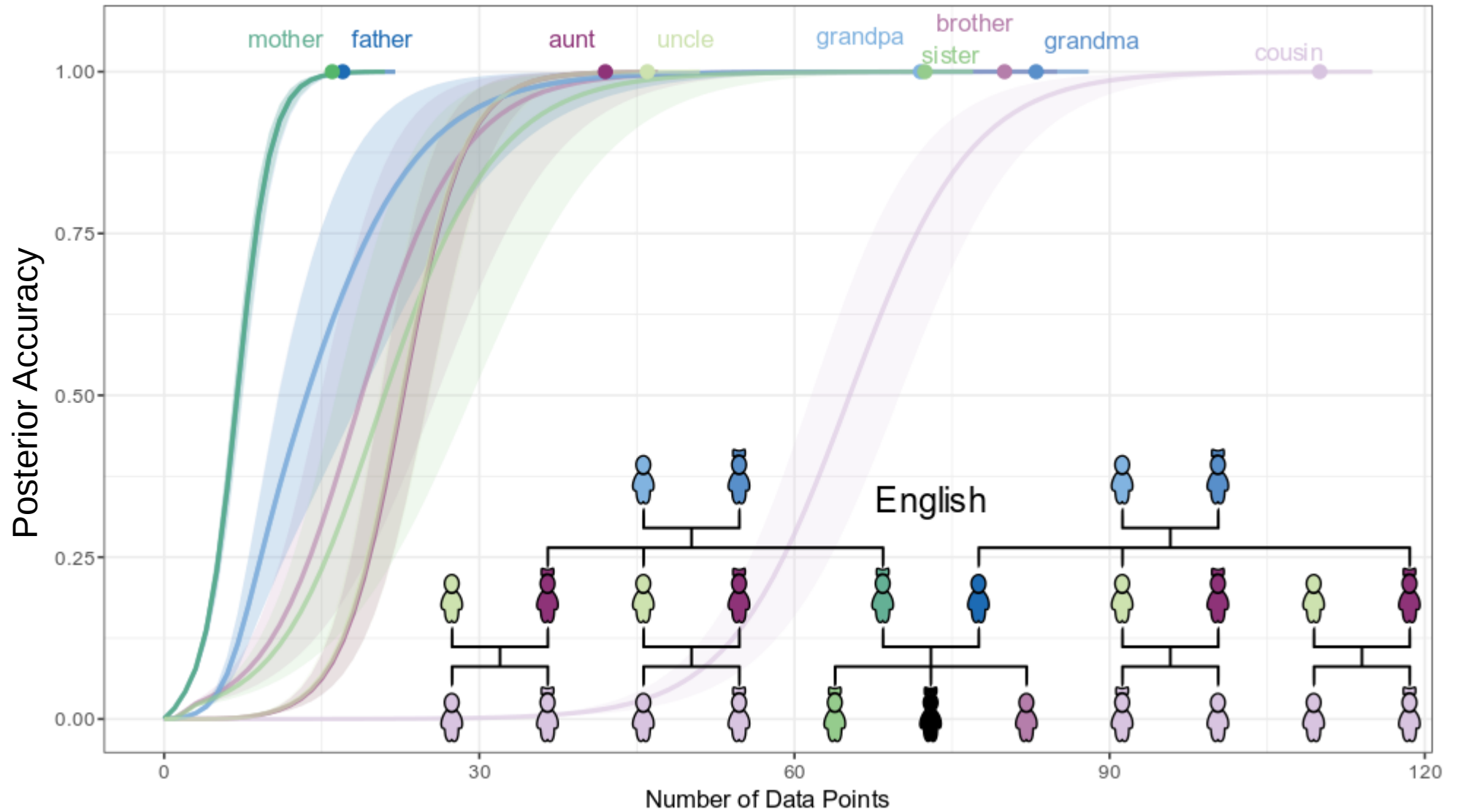
# Kinship Acquisition Phenomena

- O: Kids learn **their** kinship system
- O: Young kids prefer concrete referents
- O: Older kids over-generalize
- O: Generalization shifts from characteristic features to defining features
- O: English kinship terms are acquired in a roughly consistent order

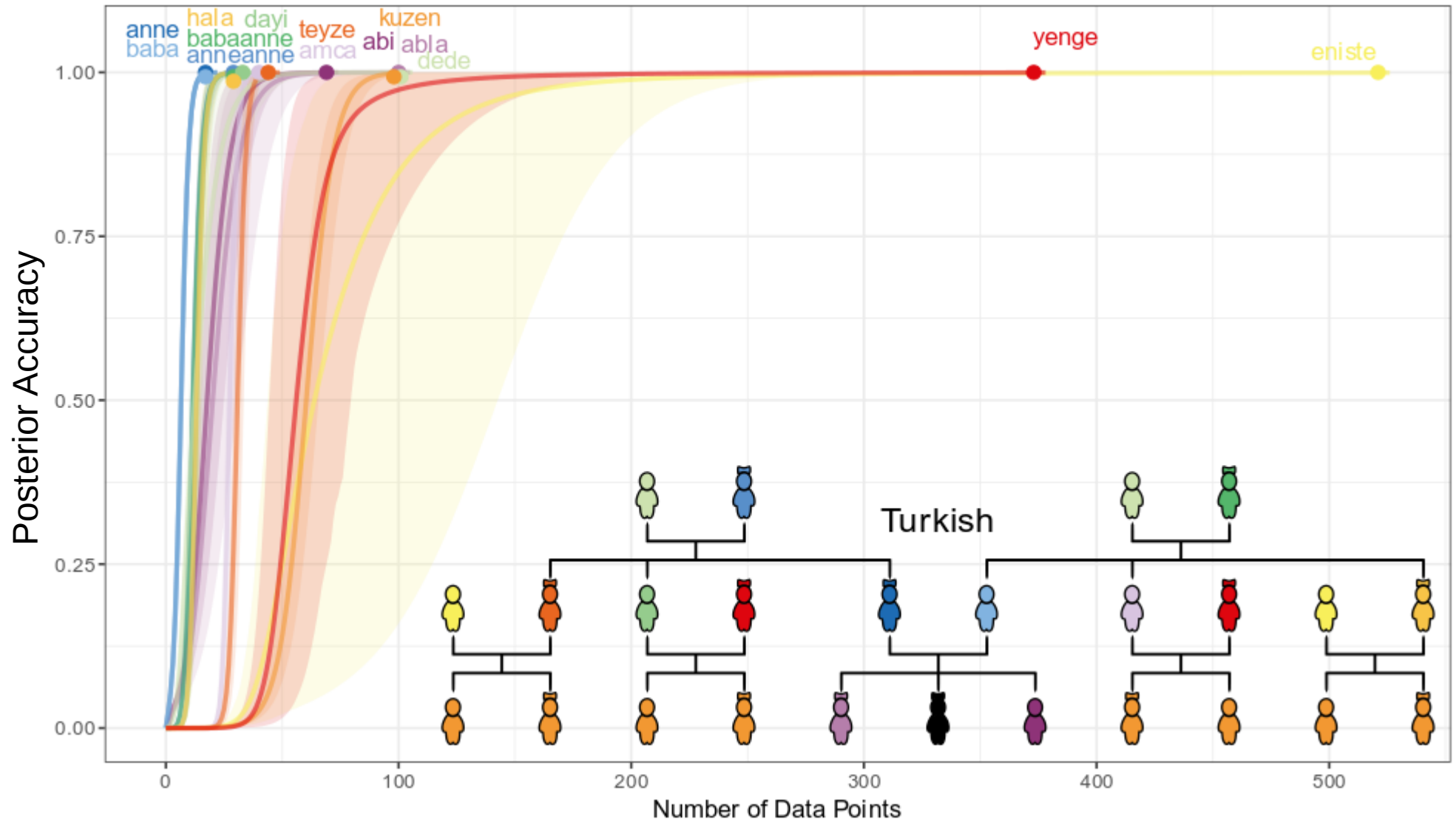
# O: Kids learn **their** kinship system



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# O: Young kids prefer concrete referents

I: What is an uncle?

S: Uncle Anthony

I: Tell me everything you know about an uncle.

S: Uncle Henry

I: Anything else?

S: You know ... Yogi's an uncle ... Yea ... Booboo's an uncle.

I: Is Booboo an uncle?

S: No, he's a Booboo boy ...

I: What kind of a thing is an uncle?

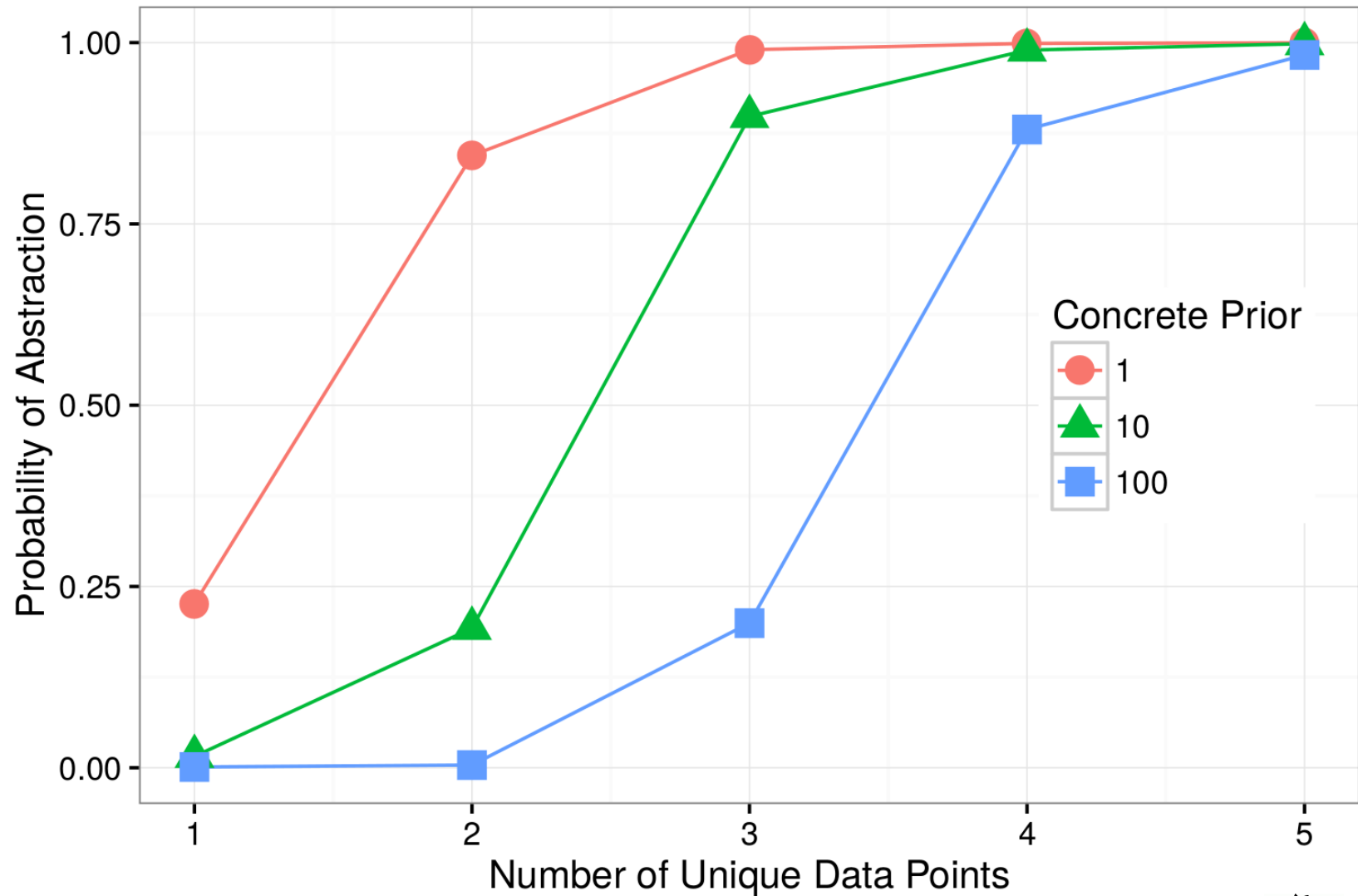
S: Um ... um ... Yogi.



**3;0 YO**

(Benson & Anglin, 1987)

# O: Young kids prefer concrete referents



# O: Older kids over-generalize

I: What is an uncle?

S: Uncle. I don't know ...

I: Tell me everything you know about an uncle.

S: He's a man.

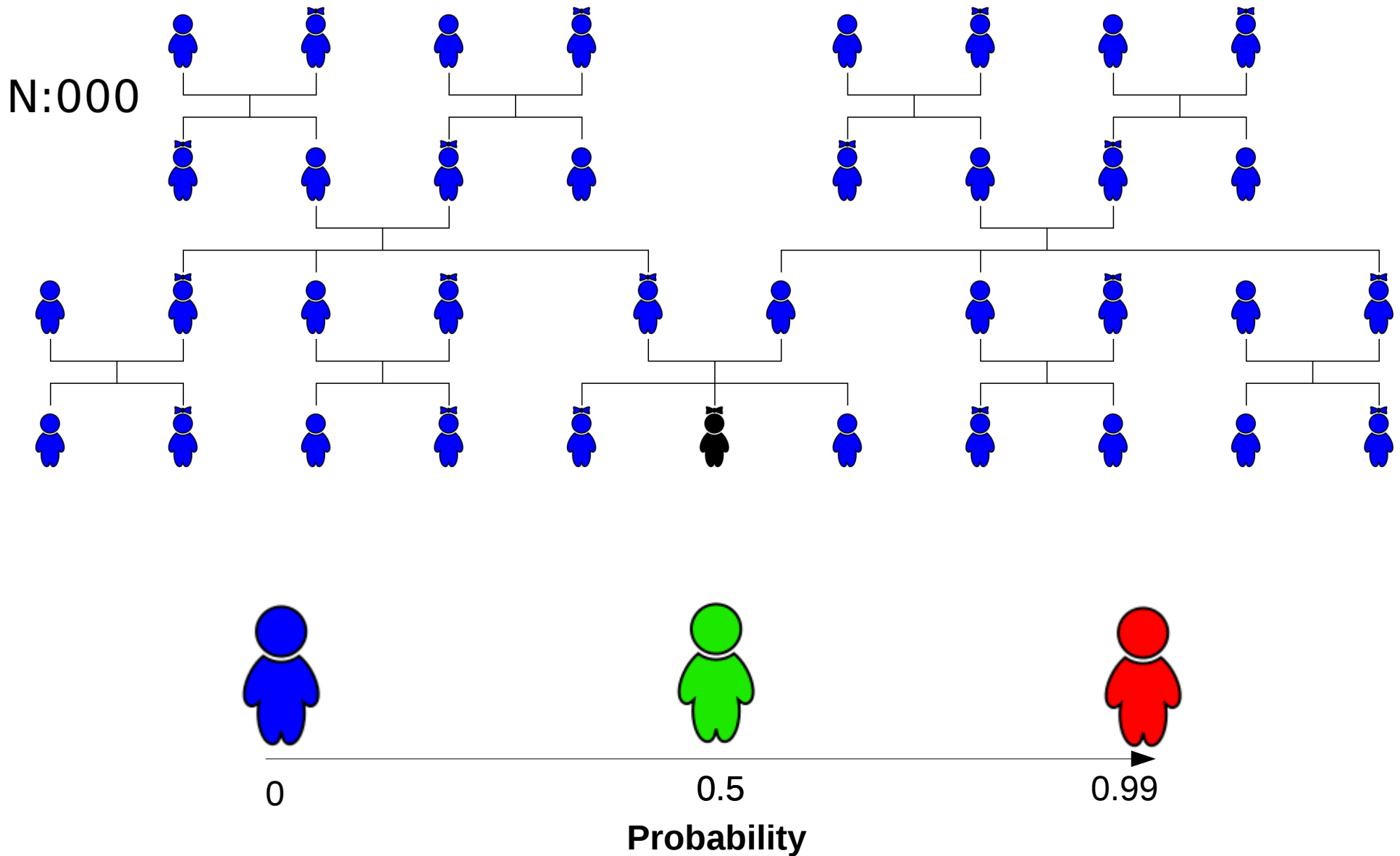
I: What kind of a thing is an uncle?

S: He's a man.

**5;4 YO**

(Benson & Anglin, 1987)

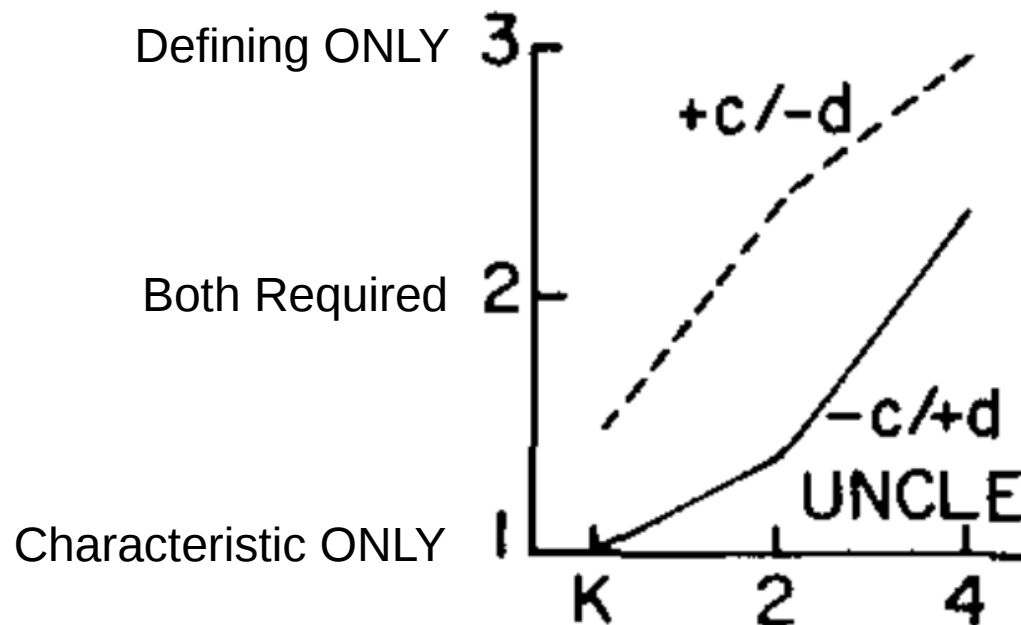
# O: Older kids over-generalize



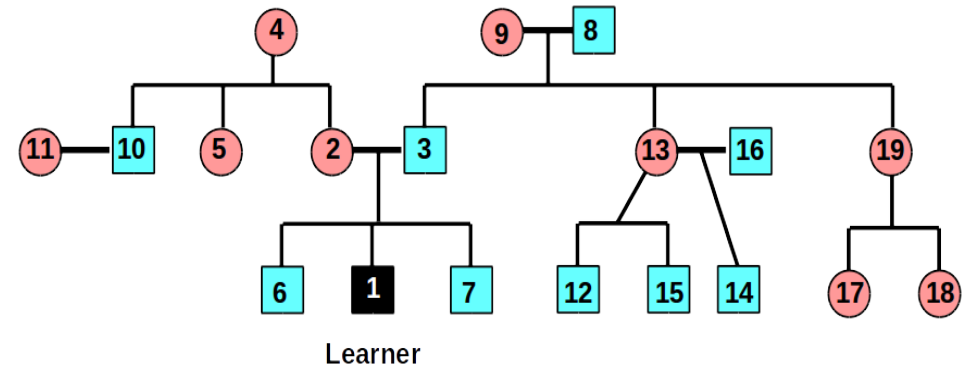
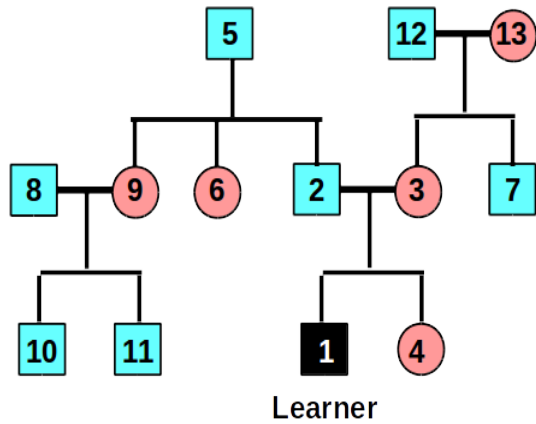
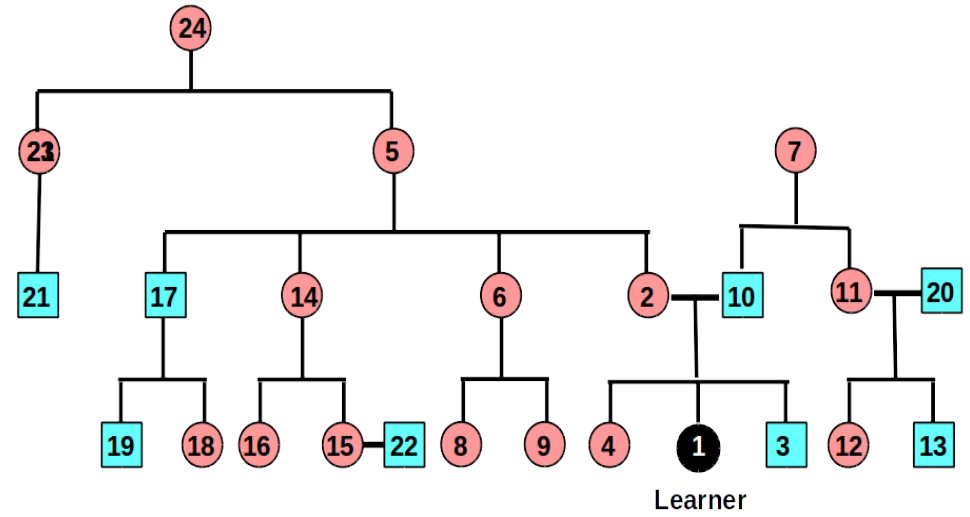
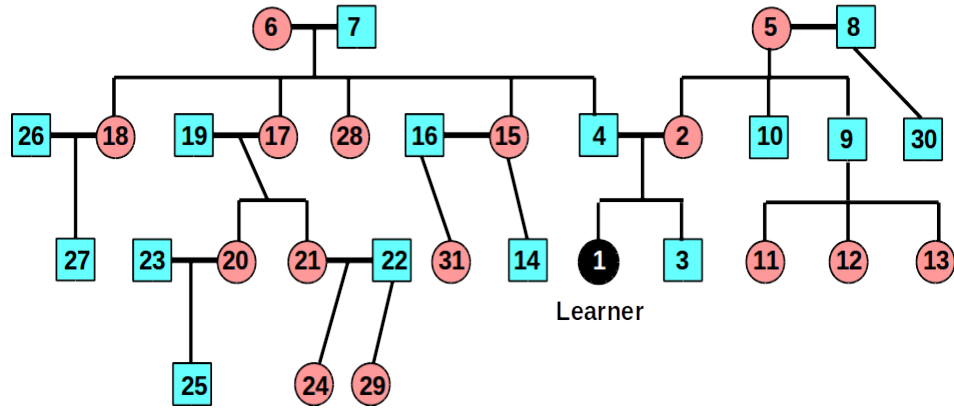


# O: Generalization shifts from characteristic to defining features

- This man your daddy's age loves you and your parents and loves to visit and bring presents, but he's not related to your parents at all. He's not your mommy or daddy's brother or sister or anything like that. Could that be an uncle?
- Suppose your mommy has all sorts of brothers, some very old and some very, very young. One of your mommy's brothers is so young he's only 2 years old. Could that be an uncle?



# Family Tree Data Collection (N=4)



# Feature Matrix Data Collection

## (N=4)

Family Members

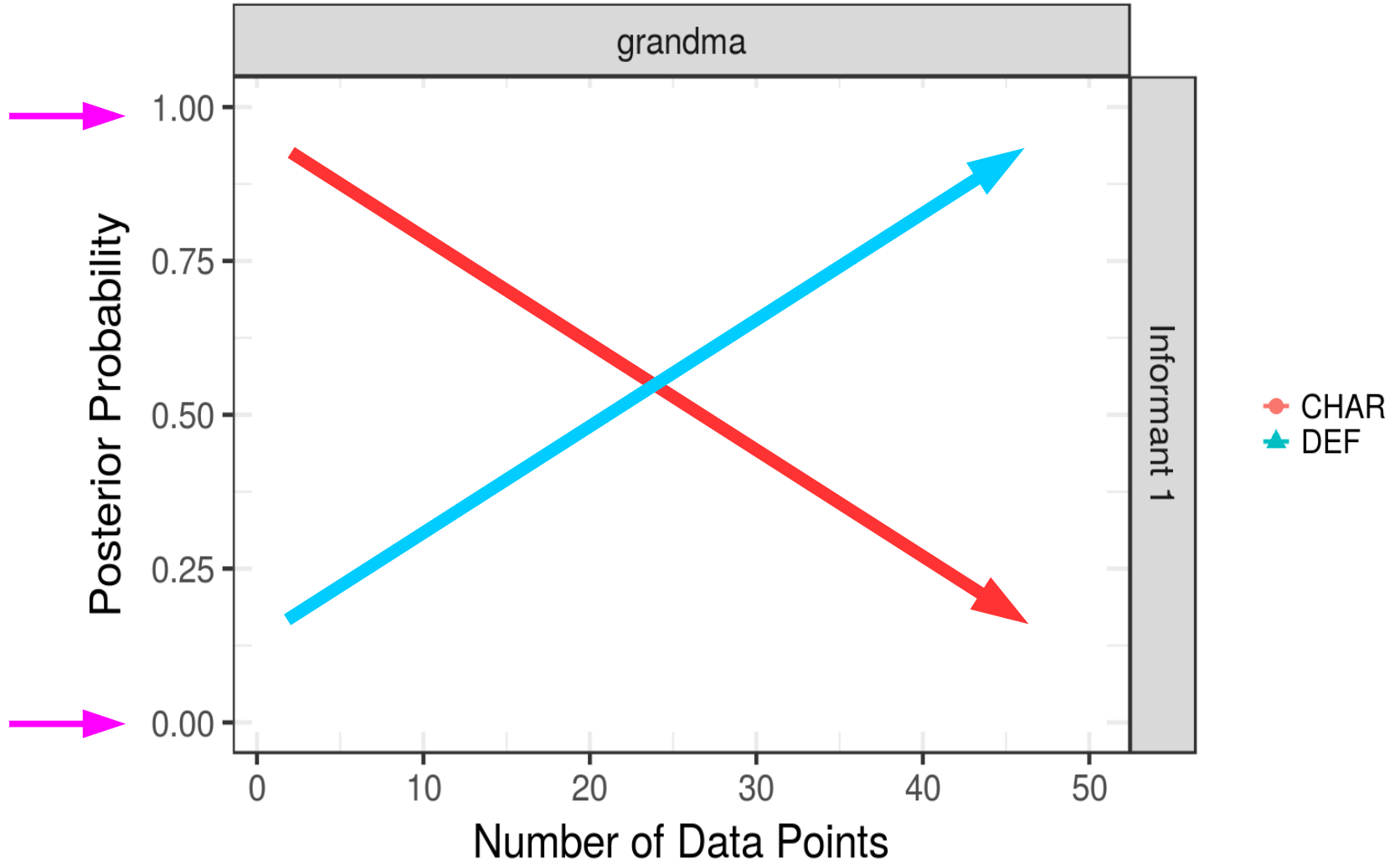


Elicited Features

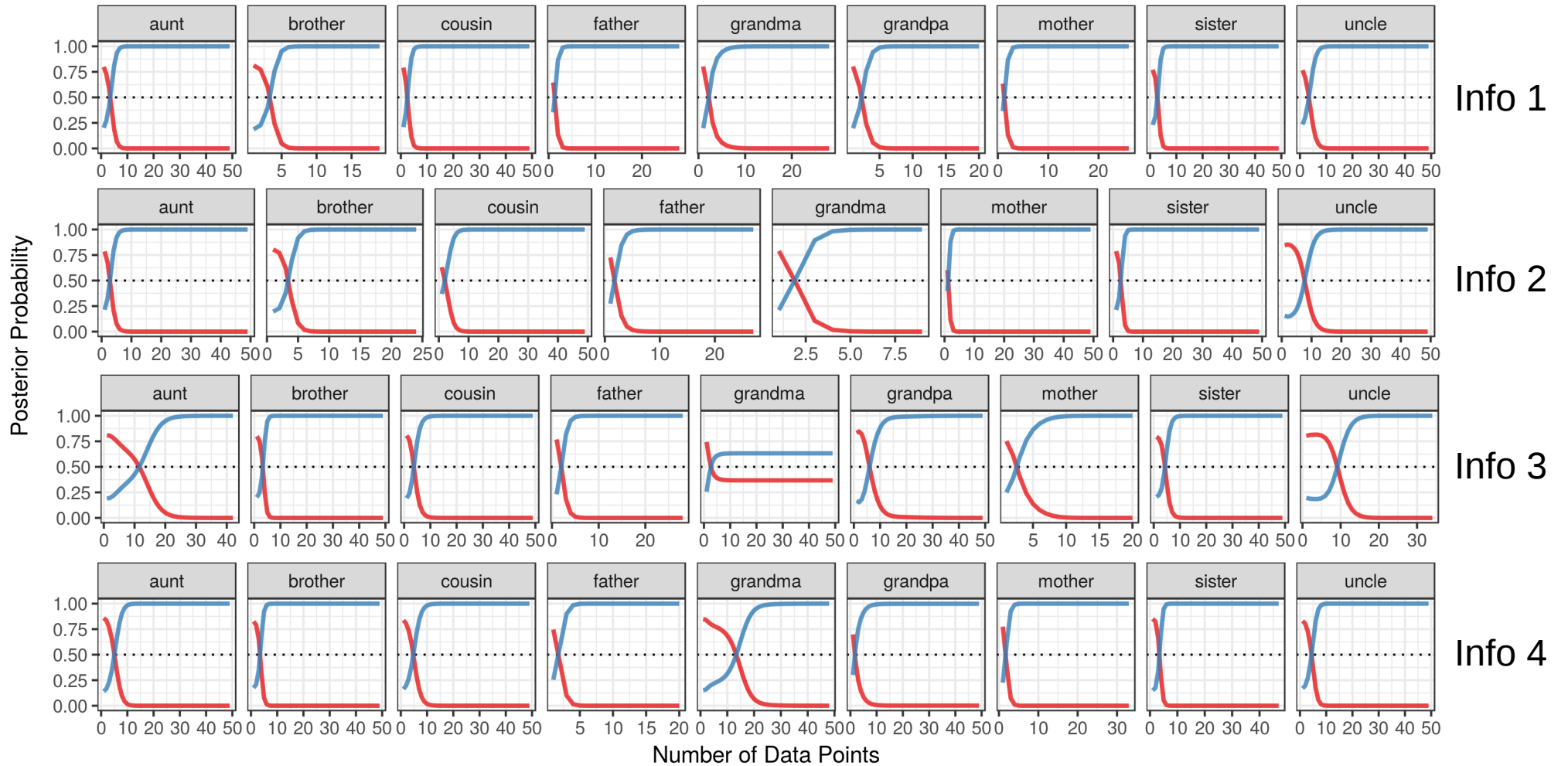
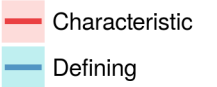
# Characteristic-to-Defining Shift

Likely to use

Unlikely to use



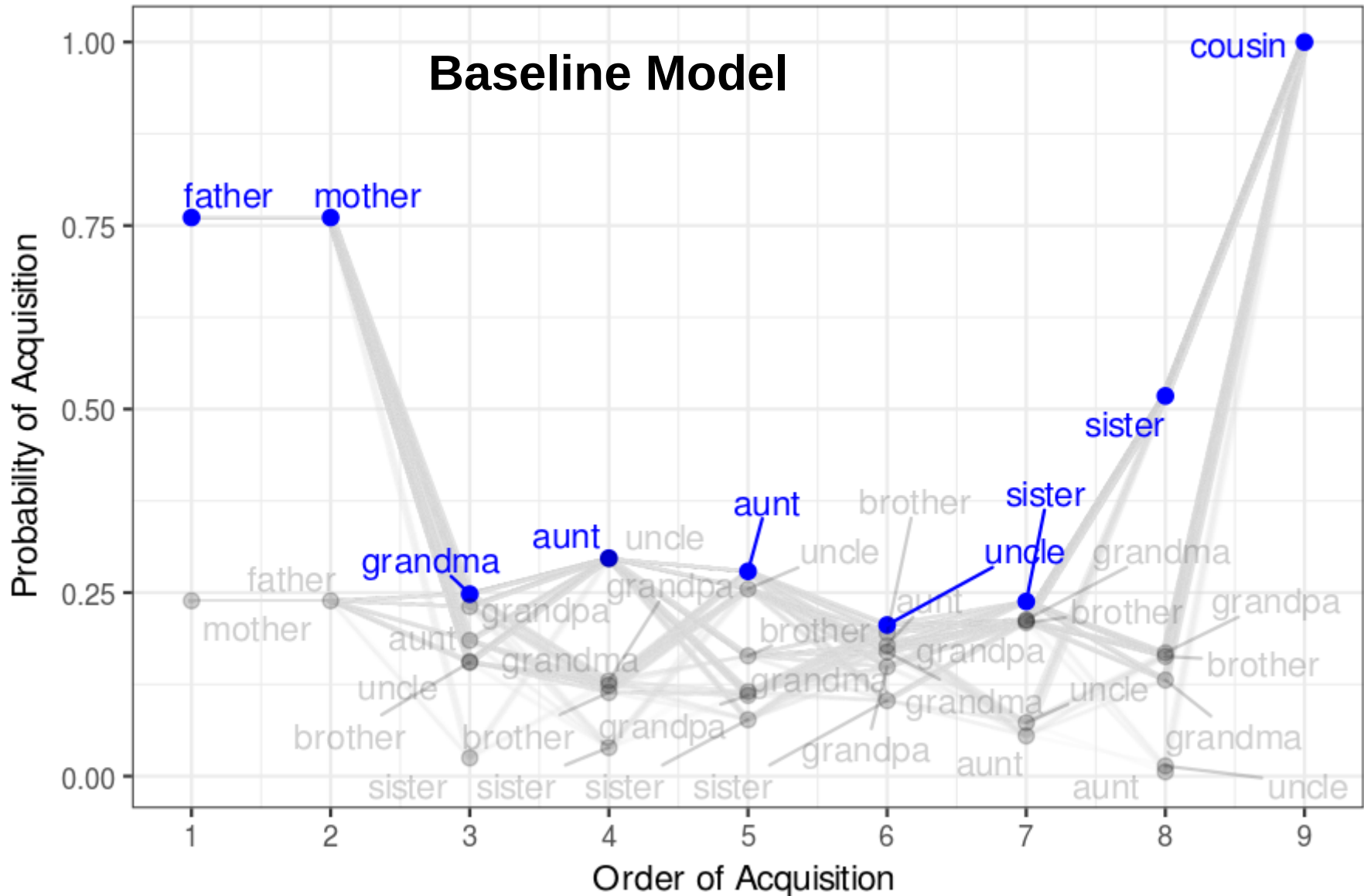
# Characteristic-to-Defining Shift



# O: “Consistent” Order of Acquisition

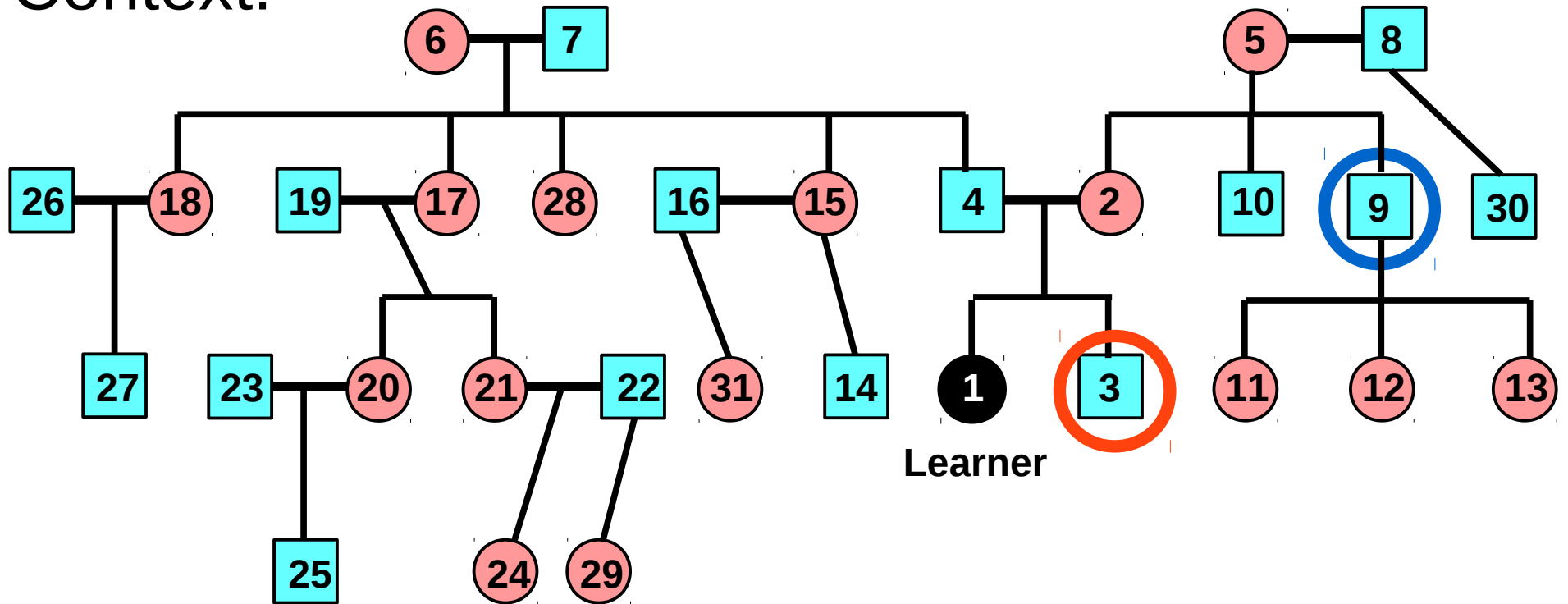
Empirical Order	Word	Original H&C Order & Formalization	Log Prior	CHILDES Freq.
1	<i>mother</i>	Level I: [X PARENT Y][FEMALE]	-9.457	6812
1	<i>father</i>	Level I: [X PARENT Y][MALE]	-9.457	3605
2	<i>brother</i>	Level III: [X CHILD A][A PARENT Y][MALE]	-13.146	41
2	<i>sister</i>	Level III: [X CHILD A][A PARENT Y][FEMALE]	-13.146	89
3	<i>grandma</i>	Level II: [X PARENT A][A PARENT Y][FEMALE]	-13.146	526
3	<i>grandpa</i>	Level II: [X PARENT A][A PARENT Y][MALE]	-13.146	199
4	<i>aunt</i>	Level IV: [X SIB A][A PARENT Y][FEMALE]	-19.320	97
4	<i>uncle</i>	Level IV: [X SIB A][A PARENT Y][MALE]	-19.320	68
4	<i>cousin</i>	Level IV: [X CHILD A][A SIB B][B PARENT Y]	-18.627	14

# O: "Consistent" Order of Acquisition



# Where does data come from?

- Context:

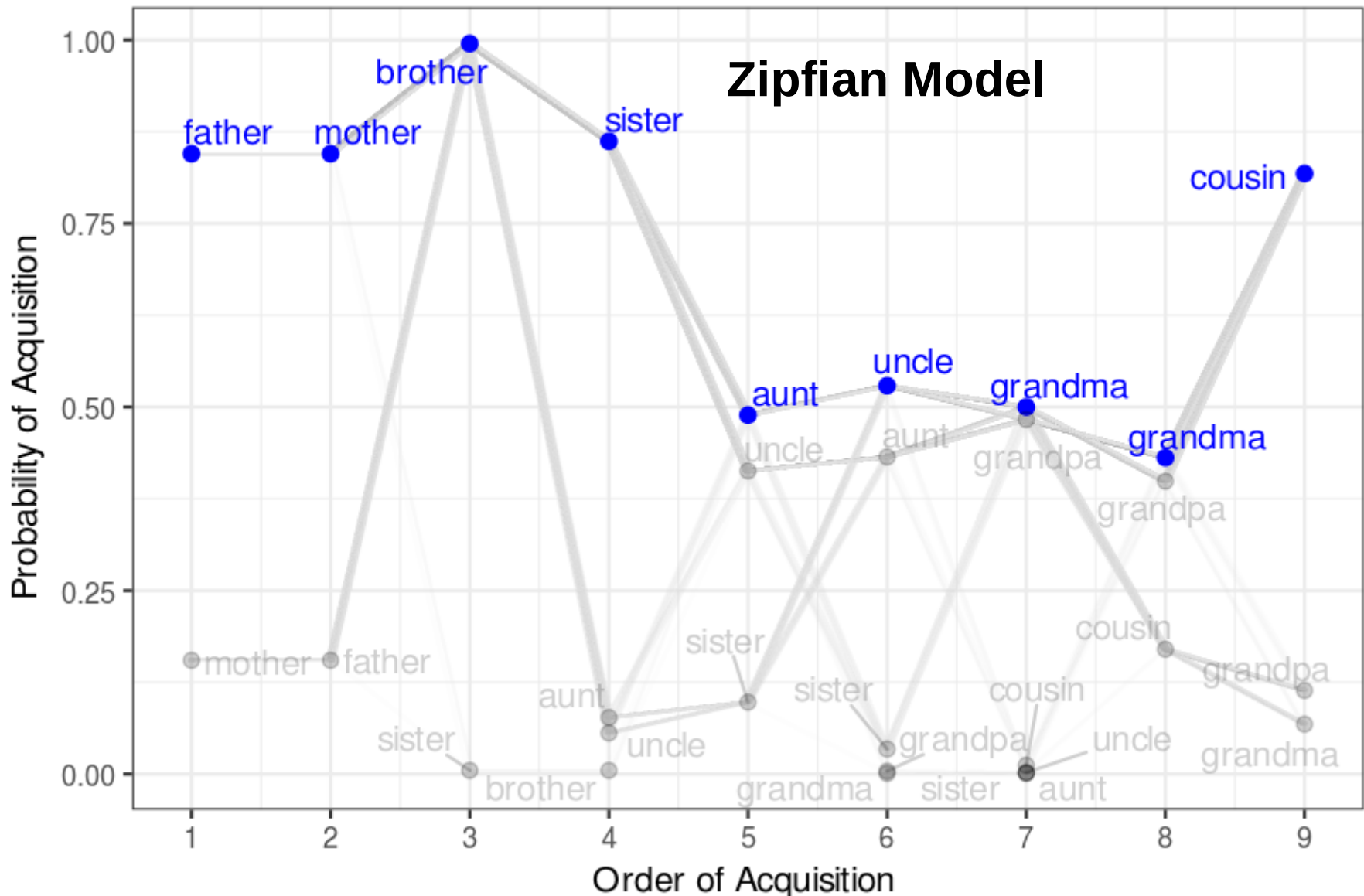


- Data Point:

- Context
- *Word uncle*
- Speaker ○
- Referent ○

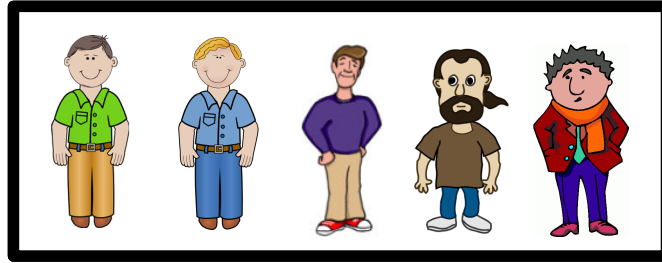


# O: “Consistent” Order of Acquisition

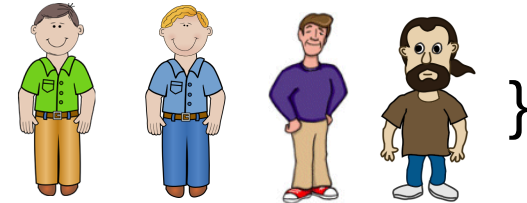


# Hypothesis vs. Lexicon

Context =



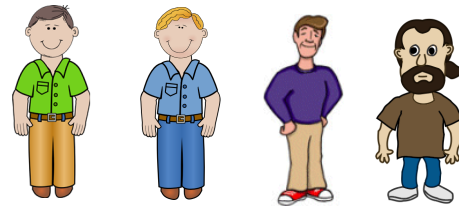
$H(\text{Context}) = L(\textit{uncle}, \text{Context}) = \{$



$H(\text{Context}) = L(\textit{father}, \text{Context}) = \{$



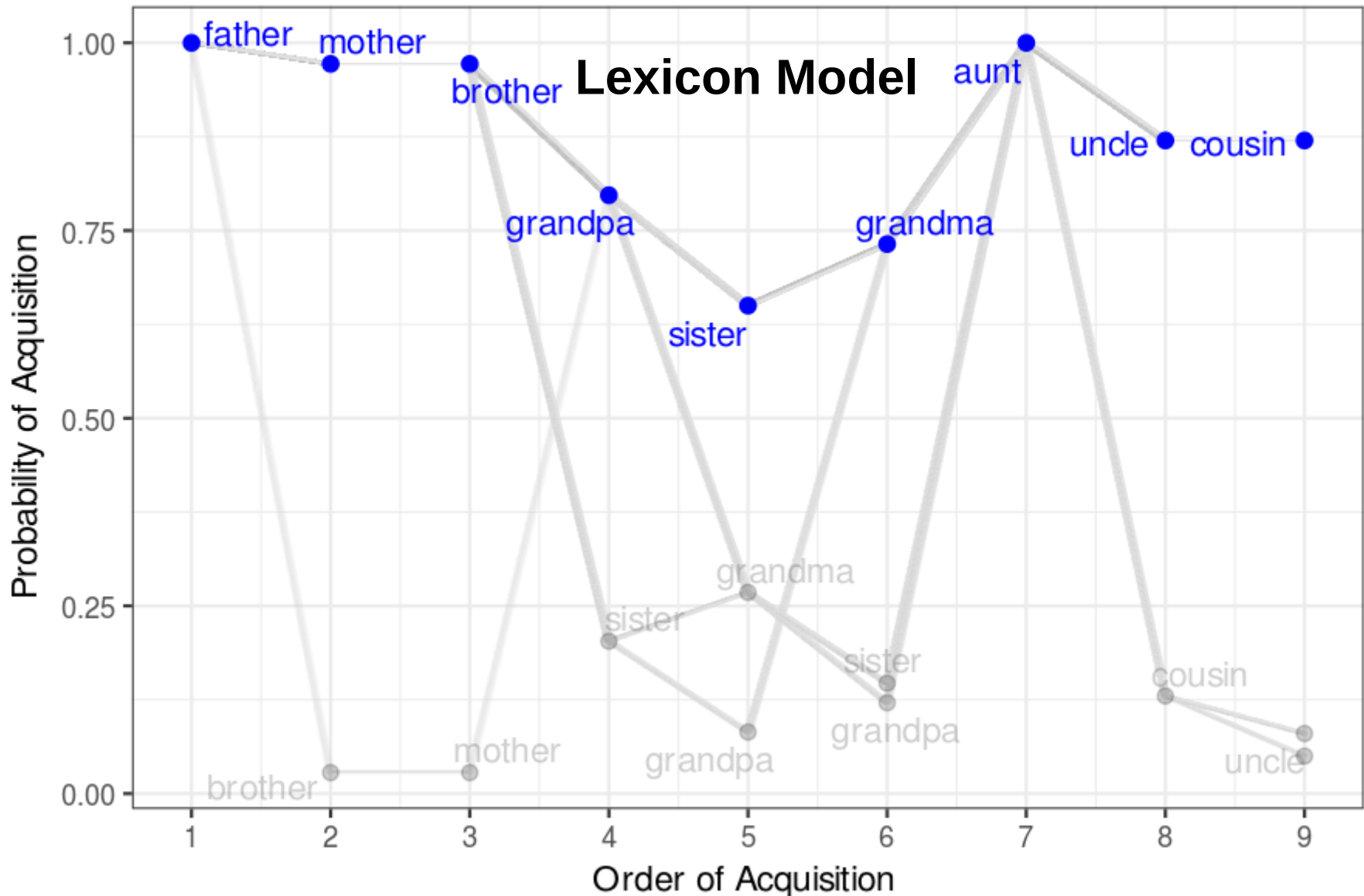
$L(\text{Context}) = \{ \textit{uncle}:$



$\textit{father}:$



# O: “Consistent” Order of Acquisition



# Take-Aways

- Model provides convergent evidence:

	<b>Simplicity</b>	<b>Data / Environment</b>	<b>Inter-related Systems</b>
<b>Learnability</b>		X	
<b>Concrete Preference</b>	X		
<b>Over-generalization</b>	X	X	
<b>Characteristic-to-Defining Shift</b>		X	
<b>Order of Acquisition</b>		X	X

- Model generates non-verbal predictions for word generalization.
- Model is a potential data analysis tool.

# Acknowledgements



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# References

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# Further Details

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