

Psych 215L: Language Acquisition

Lecture 14 Poverty of the Stimulus: Syntactic Islands

An induction problem by any other name...

One of the most controversial claims in linguistics is that children face an **induction problem**:

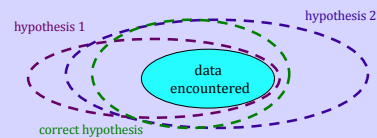
“**Poverty of the Stimulus**” (Chomsky 1980, Crain 1991, Lightfoot 1989, Valian 2009)

“**Logical Problem of Language Acquisition**” (Baker 1981, Hornstein & Lightfoot 1981)

“**Plato’s Problem**” (Chomsky 1988, Dresher 2003)

Basic claim:

The data encountered are **compatible with multiple hypotheses**.

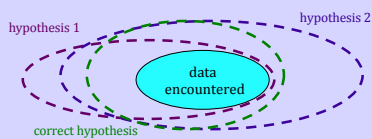


The induction problem

Extended claim:

Given this, the data are insufficient for identifying the correct hypothesis as quickly as children do (Legate & Yang 2002) – or at all.

Big question: **How do children do it, then?**



One answer: Children come prepared

- Children are not unbiased learners.



- But if children come equipped with helpful learning biases, then what is the nature of these necessary biases?

- Are they **innate** or **derived** from the input somehow?
- Are they **domain-specific** or domain-general?
- Are they about **the hypothesis space** or about the learning mechanism?

The Universal Grammar (**UG**) hypothesis (Chomsky 1965, Chomsky 1975):

These biases are **innate** and **domain-specific**.

The Plan

- (1) Look at **syntactic islands**: phenomena central to UG-based syntactic theories.
- (2) Explicitly define the **target knowledge state**, based on adult acceptability judgments.
- (3) Identify the kind of **data** children and adults have in their input, using realistic samples of child-directed and adult-directed input.
- (4) Implement a **computational learner** that is able to reach the target knowledge state, given realistic data distributions, and see what kind of learning biases it requires. It turns out that none of these are necessarily **innate** and **domain-specific**, and so learning syntactic islands does not require UG-like biases.

Syntactic Islands

Dependencies can exist between two non-adjacent items, and these do not appear to be constrained by length (Chomsky 1965, Ross 1967).



- What does Jack think __?
- What does Jack think that Lily said __?
- What does Jack think that Lily said that Sarah heard __?
- What does Jack think that Lily said that Sarah heard that Jareth stole __?

Syntactic Islands

However, if the gap position appears inside certain structures (called "syntactic islands" by Ross (1967)), the dependency seems to be **ungrammatical**.



- *What did you make [the claim that Jack bought __]?
- *What do you think [the joke about __] offended Jack?
- *What do you wonder [whether Jack bought __]?
- *What do you worry [if Jack buys __]?
- *What did you meet [the scientist who invented __]?
- *What did [that Jack wrote __] offend the editor?
- *What did Jack buy [a book and __]?
- *Which did Jack borrow [__ book]?

Syntactic Islands

Predominant learning theory in generative syntactic theory: syntactic islands require **innate, domain-specific** learning biases.

Example: Subjacency

A dependency cannot cross two or more bounding nodes (Chomsky 1973, Huang 1982, Lasnik & Saito 1984).

Bounding nodes: language-specific (CP, IP, and/or NP)

Learning biases:

- (1) **Innate, domain-specific** knowledge of **hypothesis space**: Exclude hypotheses that allow dependencies crossing 2+ bounding nodes.
- (2) **Innate, domain-specific** knowledge of **hypothesis space**: Hypothesis space consists of bounding nodes for all languages, and the child must identify the ones applicable to her language.

The target state: Adult knowledge of syntactic islands

Sprouse et al. (2012) collected magnitude estimation judgments for four different islands:

Complex NP islands

*What did you make [the claim that Jack bought ___]?

Subject islands

*What do you think [the joke about ___] offended Jack?

Whether islands

*What do you wonder [whether Jack bought ___]?

Adjunct islands

*What do you worry [if Jack buys ___]?

The target state: Adult knowledge of syntactic islands

Sprouse et al. (2012)'s factorial definition controls for two salient properties of island-crossing dependencies:

- length of dependency (short vs. long)
- presence of an island structure (non-island vs. island)

Island = **superadditive** interaction of the two factors (additional unacceptability that arises when the two factors are combined, above and beyond the independent contribution of each factor)



The target state: Adult knowledge of syntactic islands

Sprouse et al. (2012)'s factorial definition controls for two salient properties of island-crossing dependencies:

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- presence of an island structure (non-island vs. island)

Complex NP islands

Who ___ claimed that Lily forgot the necklace? short | non-island
What did the teacher claim that Lily forgot ___? long | non-island
Who ___ made the claim that Lily forgot the necklace? short | island
*What did the teacher make the claim that Lily forgot ___? long | island

The target state: Adult knowledge of syntactic islands

Sprouse et al. (2012)'s factorial definition controls for two salient properties of island-crossing dependencies:

- length of dependency (short vs. long)
- presence of an island structure (non-island vs. island)

Subject islands

Who ___ thinks the necklace is expensive? short | non-island
What does Jack think ___ is expensive? long | non-island
Who ___ thinks the necklace for Lily is expensive? short | island
*Who does Jack think the necklace for ___ is expensive? long | island

The target state: Adult knowledge of syntactic islands

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- presence of an island structure (non-island vs. island)

Whether islands

Who __ thinks that Jack stole the necklace?	short	non-island
What does the teacher think that Jack stole __?	long	non-island
Who __ wonders whether Jack stole the necklace?	short	island
*What does the teacher wonder whether Jack stole __?	long	island

The target state: Adult knowledge of syntactic islands

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Adjunct islands

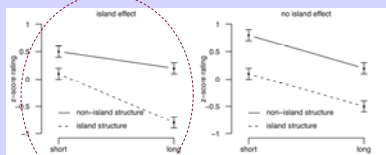
Who __ thinks that Lily forgot the necklace?	short	non-island
What does the teacher think that Lily forgot __?	long	non-island
Who __ worries if Lily forgot the necklace?	short	island
*What does the teacher worry if Lily forgot __?	long	island

The target state: Adult knowledge of syntactic islands

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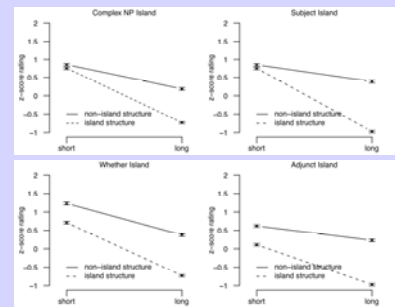
Superadditivity is visually salient



The target state: Adult knowledge of syntactic islands

Sprouse et al. (2012)'s data on the four island types (173 subjects)

Superadditivity present for all islands tested
= Knowledge that dependencies cannot cross these island structures is part of the adult knowledge state



The input: Induction problems

Data from three corpora of child-directed speech (Brown-Adam, Brown-Eve, Valian) from CHILDES (MacWhinney 2000): speech to 23 children between the ages of one and four years old.

Total words: 340,913

Utterances containing a *wh*-word and a verb: 14,260

Sprouse et al. (2012) stimuli types:

	SHORT	NON-ISLAND	LONG	NON-ISLAND	SHORT	ISLAND	LONG	ISLAND
Complex NP	4		177		0		0	
Subject	4		13		0		0	
Whether	4		177		0		0	
Adjunct	4		177		3		0	

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These kinds of utterances are fairly rare in general - the most frequent appears less than 0.01% of the time (177 of 14,260.)

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Being grammatical doesn't necessarily mean an utterance will appear in the input at all.

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Unless the child is sensitive to very small frequencies, it's difficult to tell the difference between grammatical and ungrammatical dependencies sometimes...

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...and impossible to tell no matter what the rest of the time. This looks like an induction problem for the language learner.

Building a computational learner: Proposed learning biases

Learning Bias: Children track the occurrence of structures that can be derived from phrase structure trees - **container nodes**.

[_{CP} Who did [_{IP} she [_{VP} like _]]]?
IP VP

Container node sequence: IP-VP

[_{CP} Who did [_{IP} she [_{VP} think [_{CP} [_{IP} the gift] [_{VP} was [_{PP} from _]]]]]]]?
IP VP CP IP VP PP

Container node sequence: IP-VP-CP-IP-VP-PP

Building a computational learner: Proposed learning biases

Learning Bias: Children track the occurrence of structures that can be derived from phrase structure trees - **container nodes**.

How to do this:

Identifying container nodes

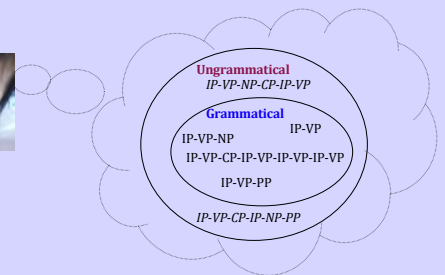
- applies to language data: **domain-specific**
- requires child to represent the **hypothesis space** a certain way
- **derived** from ability to parse utterances

Parsing utterances

- requires chunking data into cohesive units: likely to be **innate** and **domain-general**
- units being chunked are phrasal units: **derived** from distributional data and **domain-specific**

Building a computational learner: Proposed learning biases

Learning Bias: Children's hypotheses are about what container node sequences are grammatical for dependencies in the language.



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Building a computational learner: Proposed learning biases

Learning Bias: Implicitly assign a probability to a container node sequence by tracking **trigrams of container nodes**. A sequence's probability is the smoothed product of its trigrams.

[_{CP} Who did [_{IP} she [_{VP} like _]]]?
 IP VP
 start-IP-VP-end =
 start-IP-VP
 IP-VP-end

$$\begin{aligned} \text{Probability(IP-VP)} &= p(\text{start-IP-VP-end}) \\ &= p(\text{start-IP-VP}) * p(\text{IP-VP-end}) \end{aligned}$$

Building a computational learner: Proposed learning biases

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[_{CP} Who did [_{IP} she [_{VP} think [_{CP} [_{IP} the gift] [_{VP} was [_{PP} from _]]]]]]]?
 IP VP CP IP VP PP
 start-IP-VP-CP-IP-VP-PP-end =
 start-IP-VP
 IP-VP-CP
 VP-CP-IP
 CP-IP-VP
 IP-VP-PP
 VP-PP-end

$$\begin{aligned} \text{Probability(IP-VP-CP-IP-VP-PP)} &= p(\text{start-IP-VP-CP-IP-VP-PP-end}) \\ &= p(\text{start-IP-VP}) * p(\text{IP-VP-CP}) * p(\text{VP-CP-IP}) * p(\text{CP-IP-VP}) \\ &\quad * p(\text{IP-VP-PP}) * p(\text{VP-PP-end}) \end{aligned}$$

Building a computational learner: Proposed learning biases

Learning Bias: Implicitly assign a probability to a container node sequence by tracking **trigrams of container nodes**. A sequence's probability is the smoothed product of its trigrams.

What this does:

- longer dependencies are less probable than shorter dependencies, all other things being equal
- individual trigram frequency matters: short dependencies made of infrequent trigrams will be less probably than longer dependencies made of frequent trigrams

Effect: the frequencies observed in the input temper the detrimental effect of dependency length.

Building a computational learner: Proposed learning biases

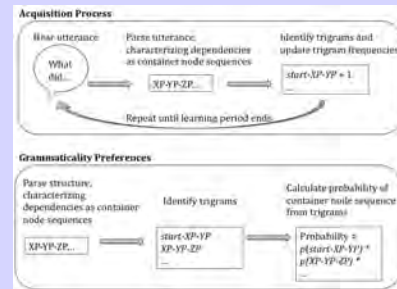
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How to do this:

- have enough memory to hold the utterance and its dependency in mind: **innate** and **domain-general**
- have enough memory to hold three units in mind (Mintz 2006, Wang & Mintz 2008, Saffran et al. 1996, Aslin et al. 1996, Saffran et al. 1999, Graf Estes et al. 2007, Saffran et al. 2008, Pelucchi et al. 2009a, 2009b): **innate** and **domain-general**
- track trigrams of units: **innate**, **domain-general**, **learning mechanism**

Building a computational learner: Proposed learning biases

Learning biases operating together to generate grammaticality preferences



Building a computational learner: Proposed learning biases

None of the proposed learning biases are **innate** and **domain-specific**.

Description of process	Domain-specific	Domain-general	Innate	Derived
Parse utterance & identify dependencies	*			*
Identify container nodes	*			*
Extract trigram sequences		*	*	
Update probability of each trigram		*	*	
Calculate probability of utterance's dependency		*	*	

Building a computational learner: Empirical grounding

Child-directed speech (Brown-Adam, Brown-Eve, Valian) from CHILDES: If we want to model child learners.

Adult-directed speech (Treebank-3-Switchboard corpus: Marcus et al. 1999) and text (Treebank-3-Brown corpus: Marcus et al. 1999): If we want to model adult learners, since we have adult data.

	Child-directed: speech	Adult-directed: speech	Adult-directed: text
total utterances	65932	74576	24243
total wh-dependencies	11308	8508	4230

Note: Child-directed speech and adult-directed speech are **qualitatively similar** in being mostly IP-VP and IP dependencies, with many more IP-VP dependencies (child: 80%_{IP-VP}/11%_{IP}; adult: 73%_{IP-VP}/17%_{IP}). Adult-directed text is still mostly IP-VP and IP dependencies, but there are more IP dependencies compared to the speech samples (63%_{IP-VP}/33%_{IP}).

Building a computational learner: Empirical grounding

Hart & Risley 1995: Children hear approximately 1 million utterances in their first three years.

Assumption: learning period for modeled learners is 3 years (ex: between 2 and 5 years old for modeling children's acquisition)



Estimating proportion of *wh*-dependencies in the input, based on child-directed speech sample: total learning period is 175,000 *wh*-dependency data points, drawn from distribution observed in speech and/or text samples.

Success metrics

Compare learned grammaticality preferences to Sprouse et al. (2012) judgment data.

To do this, we need to identify the container node sequences for each stimuli for each island type.

Complex NP islands

IP	short	non-island
IP-VP-CP/CP _{that} -IP-VP	long	non-island
IP-VP-CP/CP _{that} -IP-VP	short	island
*IP-VP-NP-CP/CP _{that} -IP-VP	long	island

Success metrics

Compare learned grammaticality preferences to Sprouse et al. (2012) judgment data.

To do this, we need to identify the container node sequences for each stimuli for each island type.

Subject islands

IP	short	non-island
IP-VP-CP/CP _{null} -IP	long	non-island
IP	short	island
*IP-VP-CP/CP _{null} -IP-NP-PP	long	island

Success metrics

Compare learned grammaticality preferences to Sprouse et al. (2012) judgment data.

To do this, we need to identify the container node sequences for each stimuli for each island type.

Whether islands

IP	short	non-island
IP-VP-CP/CP _{that} -IP-VP	long	non-island
IP	short	island
*IP-VP-CP/CP _{whether} -IP-VP	long	island

Success metrics

Compare learned grammaticality preferences to Sprouse et al. (2012) judgment data.

To do this, we need to identify the container node sequences for each stimuli for each island type.

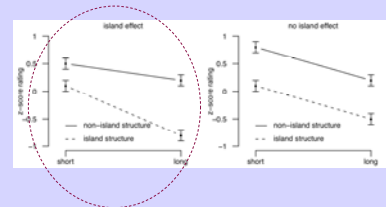
Adjunct islands

IP	short	non-island
IP-VP-CP/CP _{that} -IP-VP	long	non-island
IP	short	island
*IP-VP-CP/CP _{if} -IP-VP	long	island

Success metrics

Compare learned grammaticality preferences to Sprouse et al. (2012) judgment data.

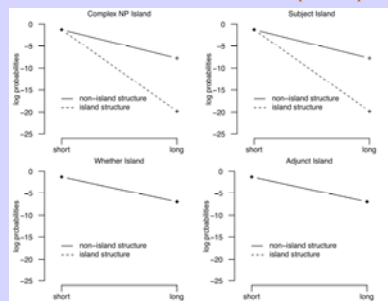
Then, for each island, we plot the predicted grammaticality preferences from the modeled learner on an interaction plot, using log probability of the dependency on the y-axis. Non-parallel lines indicate the presence of islands.



The non-UG learner

Using basic-level container nodes
- ex: only CP rather than CP_{null}, CP_{that}, etc.

Child-directed speech input

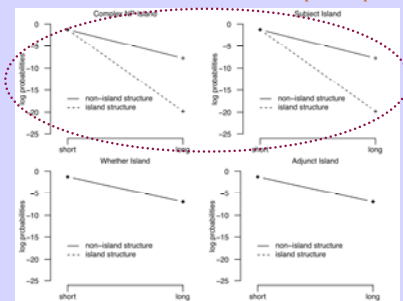


The non-UG learner

Using basic-level container nodes
- ex: only CP rather than CP_{null}, CP_{that}, etc.

Child-directed speech input

Complex NP and Subject islands have the correct superadditive behavior...



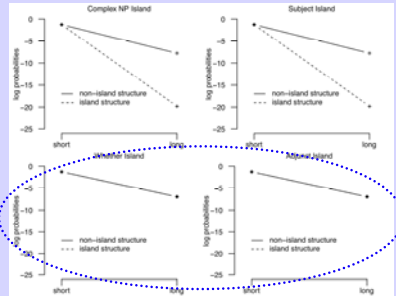
The non-UG learner

Using basic-level container nodes

- ex: only CP rather than CP_{null} , CP_{that} , etc.

Child-directed speech input

But Whether and Adjunct islands **don't**. In fact, the lines are overlapping - the learner thinks the **grammatical long** | **non-island stimuli** and **ungrammatical long** | **island stimuli** are **equally good**.

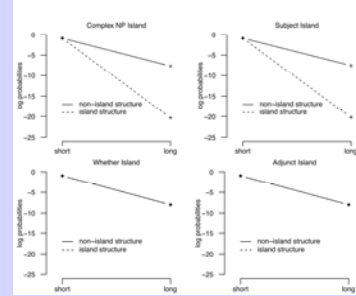


The non-UG learner

Using basic-level container nodes

- ex: only CP rather than CP_{null} , CP_{that} , etc.

Adult-directed speech & text input



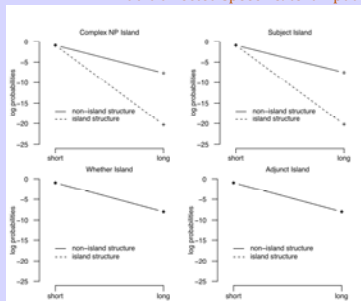
The non-UG learner

Using basic-level container nodes

- ex: only CP rather than CP_{null} , CP_{that} , etc.

Adult-directed speech & text input

The same is true for adult-directed input: the learner has the **correct** preferences for Complex NP islands and Subject islands, but has the **incorrect** preferences for Whether and Adjunct islands.



The non-UG learner

Using basic-level container nodes

- Why do we see this behavior?

The learner does not distinguish between grammatical structures with the sequence IP-VP- $CP_{null/that}$ -IP-VP

What did he think (that) she saw?

and structures with the ungrammatical sequence IP-VP- $CP_{whether/if}$ -IP-VP

* What did he wonder whether/if she saw?

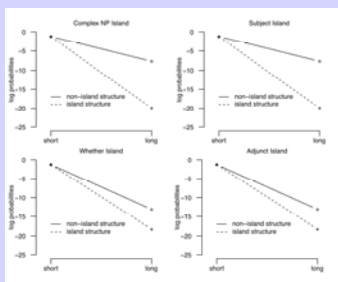


This means that Whether and Adjunct island violations, which contain specific types of CPs ($CP_{whether}$ and CP_{if}), are treated identically to grammatical utterances containing CP_{null} or CP_{that} .

The non-UG learner

Using finer-grained container nodes: include CP specification
 - ex: use CP_{null} , CP_{that} , etc.

Child-directed speech input

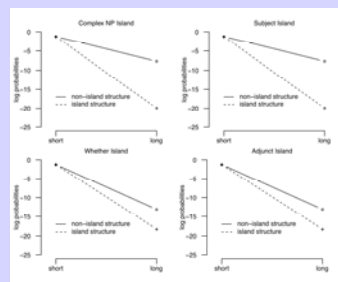


The non-UG learner

Using finer-grained container nodes: include CP specification
 - ex: use CP_{null} , CP_{that} , etc.

Child-directed speech input

Problem solved!
Superadditivity
 observed for all four
 island types.

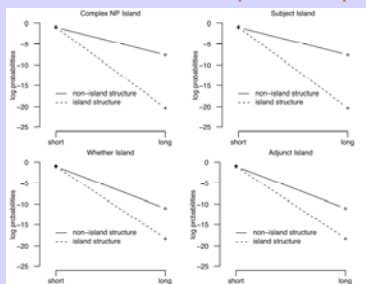


The non-UG learner

Using finer-grained container nodes: include CP specification
 - ex: use CP_{null} , CP_{that} , etc.

Adult-directed speech & text input

The same is true for
 the learner using adult-
 directed input: all four
 island plots show
superadditivity for the
 ungrammatical island
 dependency.



Implications of this learner

Basic: A learner using no biases that would traditionally be considered part of UG (i.e., innate and domain-specific biases) was able to learn the correct grammaticality preferences for dependencies over four different island types. This suggests that adult knowledge of these syntactic islands does not implicate UG.

Though there appears to be an induction problem, it does not require UG to solve it.



Implications of this learner

Something useful for children to have: Complex learning biases that are made up of simpler biases. (So, perhaps a bias to combine existing biases.)

Ex: Tracking trigrams of container nodes

- basic unit is container node (**derived**, **domain-specific**, **hypothesis space**)
- tracking 3 unit sequences (**innate**, **domain-general**, **learning mechanism**)



Implications of this learner

What about the CP specification requirement? Is that UG?

Not necessarily:

- uncontroversial to assume that children learn to distinguish different types of CPs since the lexical content of CPs has substantial consequences for the semantics of a sentence (e.g., declaratives versus interrogatives)
- adult speakers are sensitive to the distribution of *that* versus null complementizers (Jaeger 2010)



Likely a **derived**, **domain-specific** learning bias about the representation of the **hypothesis space**.

A remaining issue

This learner can't handle **parasitic gaps**, which are dependencies that span an island (and so should be ungrammatical) but which are somehow rescued by another dependency in the utterance.

*Which book did you laugh [before reading _]?

Which book did you judge ---_{true} [before reading $\text{---}_{\text{parasitic}}$]?

Adjunct island

*What did [the attempt to repair _] ultimately damage the car?

What did [the attempt to repair $\text{---}_{\text{parasitic}}$] ultimately damage ---_{true} ?

Complex NP island

A remaining issue

Why not? The current learner would judge the parasitic gap as **ungrammatical** since it is inside an island, irrespective of what other dependencies are in the utterance.

*Which book did you laugh [before reading _]?

*Which book did you judge ---_{true} [before reading $\text{---}_{\text{parasitic}}$]?

Adjunct island

*What did [the attempt to repair _] ultimately damage the car?

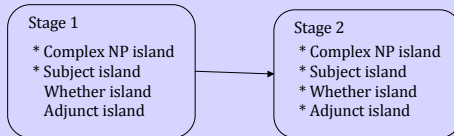
*What did [the attempt to repair $\text{---}_{\text{parasitic}}$] ultimately damage ---_{true} ?

Complex NP island

This may be able to be addressed in a learner that is able to combine information from multiple dependencies in an utterance (perhaps because the learner has observed multiple dependencies resolved in utterances in the input).

A developmental prediction

If children begin with only a basic specific of container nodes (CP instead of CP_{thad}), we may expect a period of time when they recognize Complex NP and Subject islands but view dependencies spanning Whether and Adjunct islands as grammatical. Once they allow CP specification, they will recognize Whether and Adjunct islands as well.



de Villiers & Roeper (1995) suggest that children as young as 3 years old may view dependencies spanning *wh*-islands (such as *whether* islands) as ungrammatical. If they recognize *whether* islands as well, this suggests Stage 2 would be complete by this age.