Language Evolution in the Lab: from computational models to human experiments in Evolutionary Linguistics

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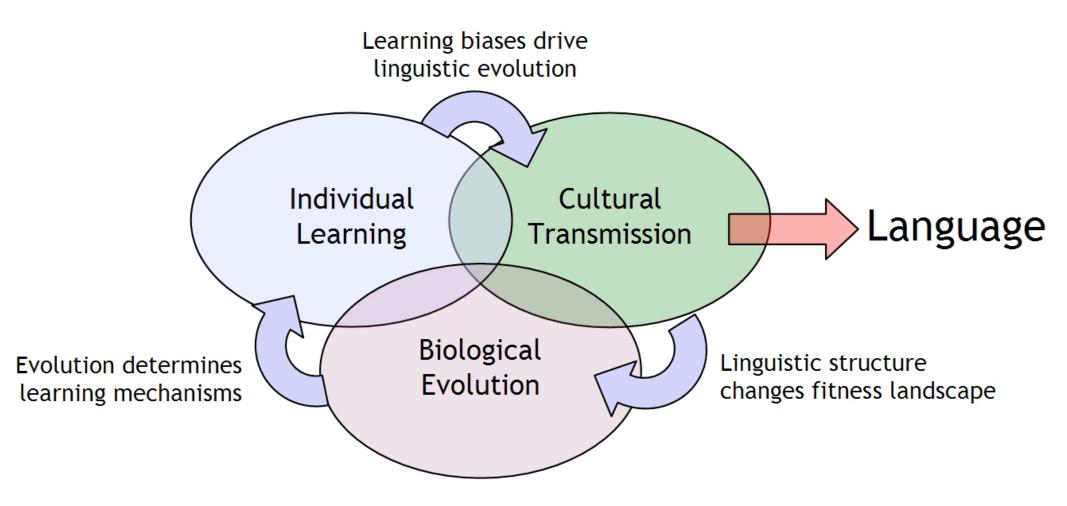
The evolutionary approach to language

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The evolutionary approach to language

- Language is (almost) unique in nature
- It involves three complex adaptive systems:
 - Biological evolution
 - Individual learning
 - Cultural transmission
- Language arises from the interaction of these three



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 - **Poverty of the stimulus** too hard to learn language without specific knowledge
 - **Complex design** natural selection is the only possible mechanism

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- Domain-specific strongly-constraining innate knowledge
- Two motivations:
 - **Poverty of the stimulus** too hard to learn language without specific knowledge
 - **Complex design** natural selection is the only possible mechanism
- These two are cast into doubt if we take cultural transmission seriously

From individuals to populations

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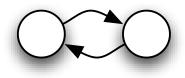
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 - Model learning as acquiring target from a fixed set of data

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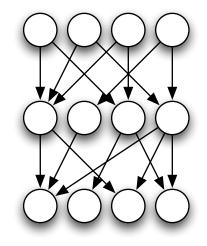
- Nativist perspective associated with idealisation of single speaker/hearer in homogenous community
 - Model learning as acquiring target from a fixed set of data
- An alternative: *iterated learning*
 - Behaviour is shaped by observation/interaction with others whose behaviour is shaped in the same way

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- One of the mechanisms of cultural transmission
- Implicated in language at every level:
 - Two individuals who align through mutual priming
 - Whole populations in which language is transmitted over generations



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- Some key names:
 - Batali, Steels, Brighton, Smith, Oudeyer, Zuidema, Baronchelli, Griffiths (many more)...
- Consistent results: adaptive structure emerges
 - For example: compositional structure emerges from "holistic" protolanguages purely through repeated transmission















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- How do we explain apparent design in nature?
 - Biological evolution by natural selection
 - Intelligent design by humans
- Modelling results suggest a third source:
 - Cultural transmission leads to adaptive structure in language
- This is an "invisible hand" process (Keller 1994)

- An experiment to demonstrate this kind of adaptation
 - Based on previous computational models of emergence of compositionality



Hannah Cornish



Kenny Smith

Kirby, Cornish & Smith (2008) PNAS

- An experiment to demonstrate this kind of adaptation
 - Based on previous computational models of emergence of compositionality
- Try to rule out other sources of design:
 - Our participants do not evolve (!)
 - Set up to rule out intentional participant design (cf. experiments by Galantucci, Garrod, Healey, Theisen, Scott-Phillips...)



Hannah Cornish



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Kirby, Cornish & Smith (2008) PNAS

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 - I. Start off with a random artificial language
 - 2. Ask an experimental subject to try and learn this language and test them
 - 3. Use their output on test as the language to teach the next subject in the experiment (and repeat)

Hypothesis

• There will be cumulative cultural adaptation of the language without intentional design by participants

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- Two ways of verifying this:
 - The language should become easier to learn
 - The language should become structured

The Language

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- A set of 27 possible "meanings"
 - Pictures with coloured objects in motion:
 - Three shapes $\Box \circ \Delta$
 - Three colours 🖉 🧲 🗲
 - Three motions ---- > /``./` (()

The Language

- A set of 27 possible "meanings"
 - Pictures with coloured objects in motion:
 - Three shapes $\Box \circ \Delta$

 - Three motions ----► , ``, ``
- A large set of possible "signals"
 - Random sequences of between two and four syllables chosen from a set of nine
 - No spaces

Procedure

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 - SEEN set: 14 string-picture pairs
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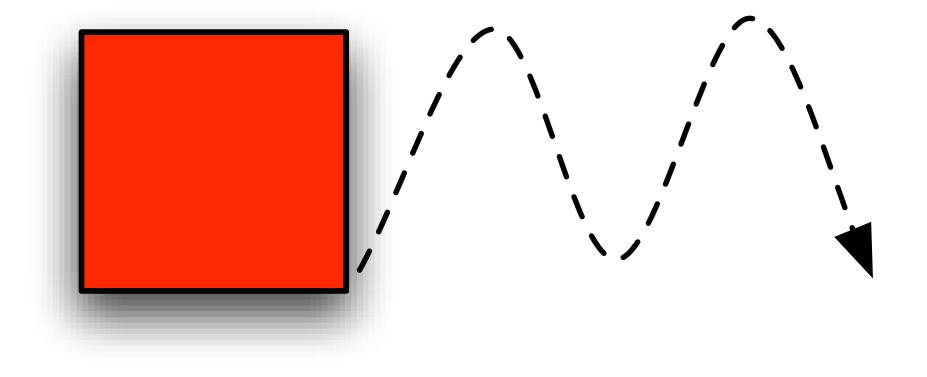
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- String displayed for I second, then string and picture for a further 5 seconds

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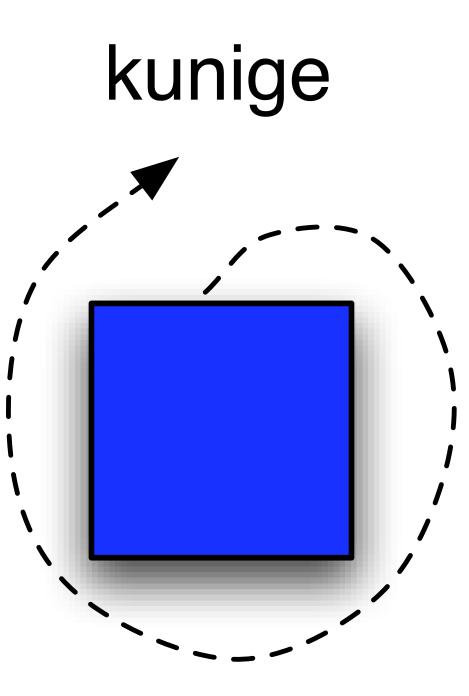
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 - SEEN set: 14 string-picture pairs
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- Subjects trained on SEEN set
- String displayed for I second, then string and picture for a further 5 seconds
- Tested on complete set,
 - randomly redivided into new SEEN and UNSEEN sets for next generation

kihemiwi

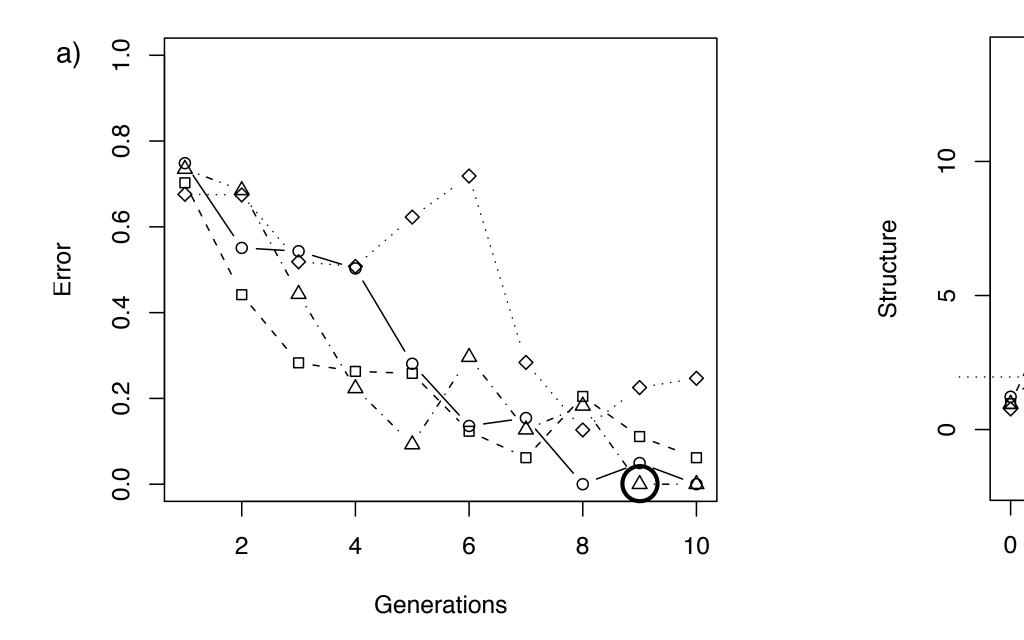
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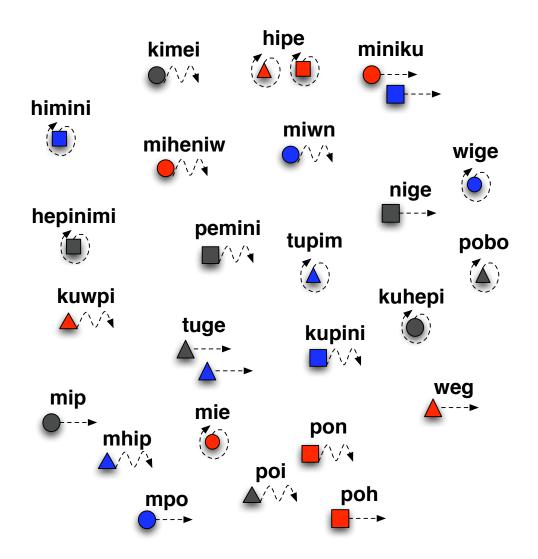
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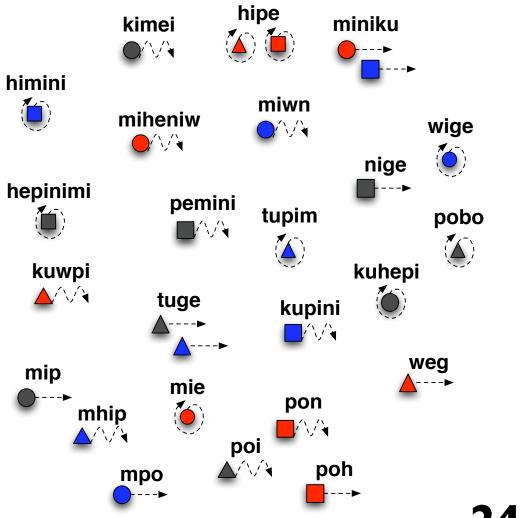
Language becomes easier to learn



After Generation I:

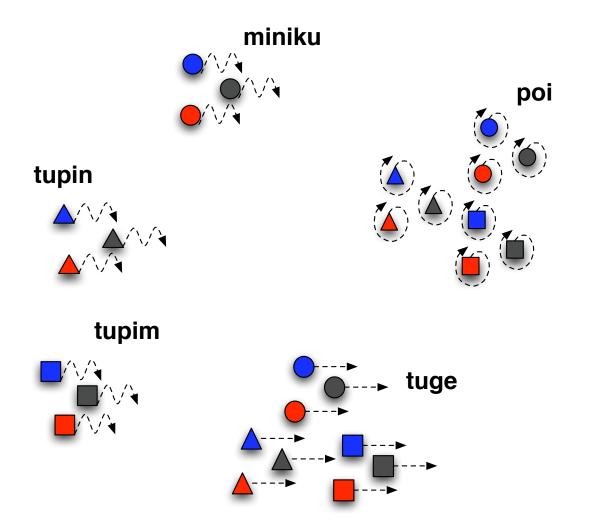


After Generation I:

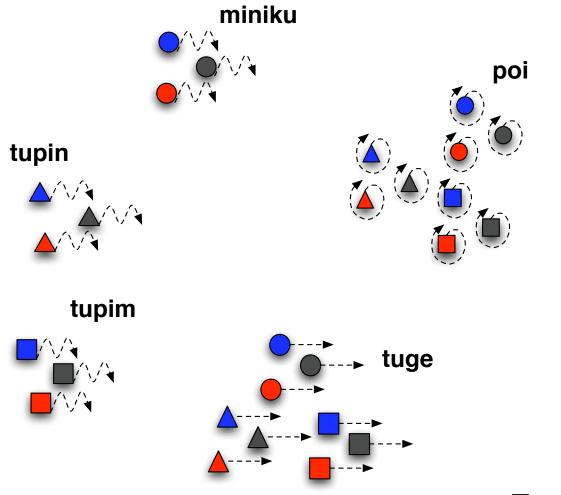


24 words

After Generation 10:



After Generation 10:



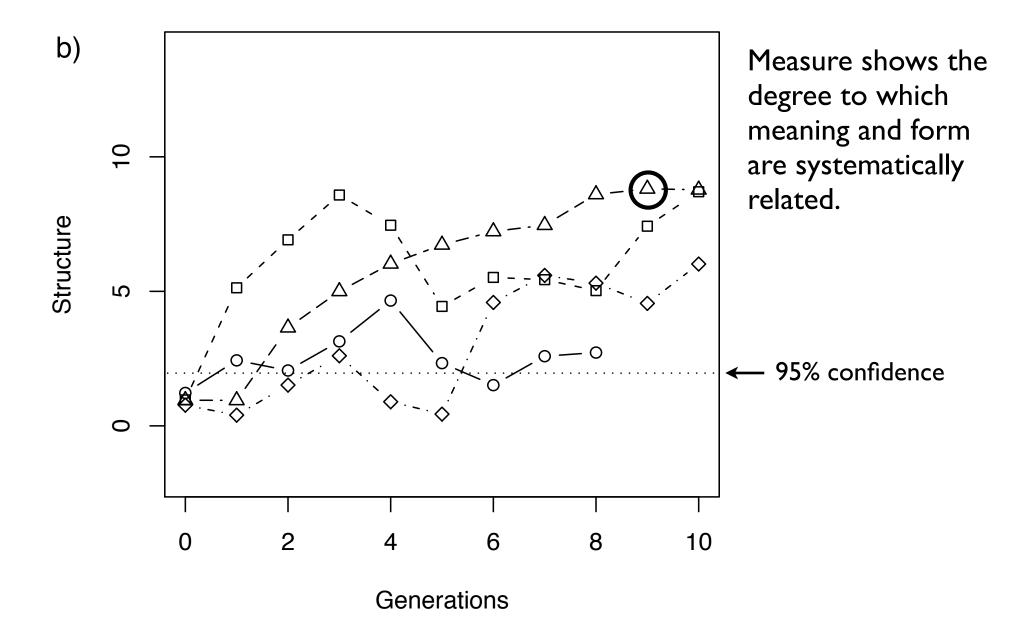
5 words

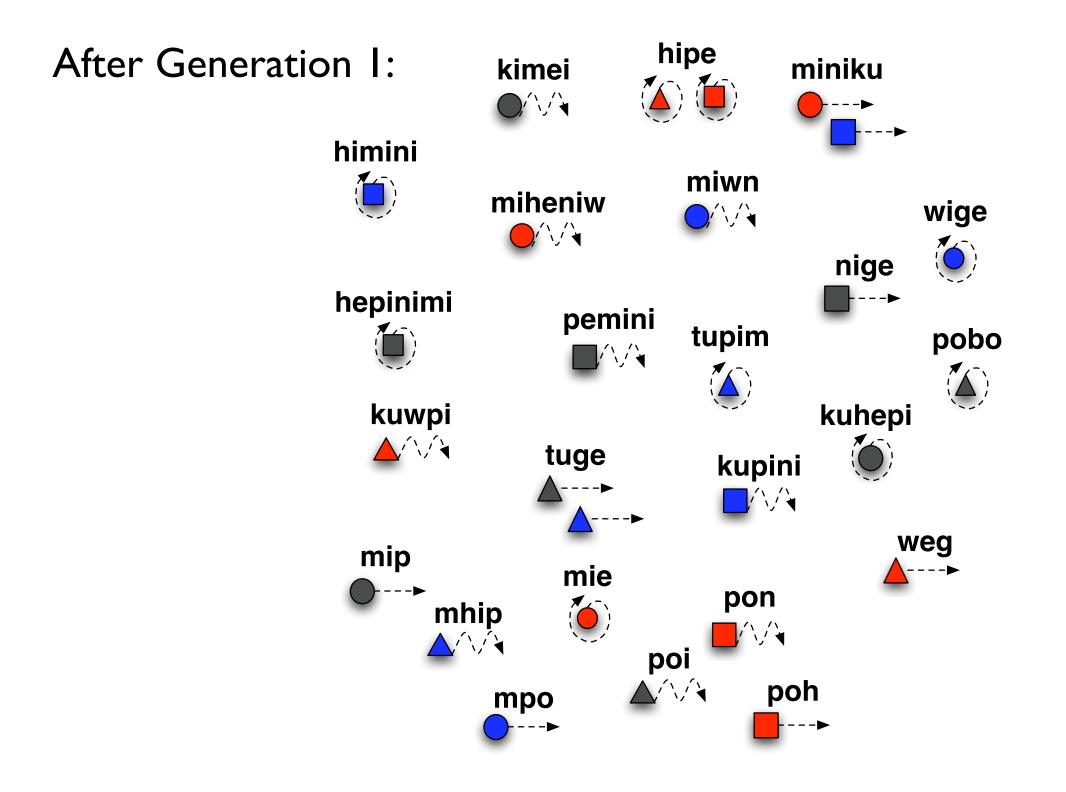
• Looks like it might be just that there are fewer words.

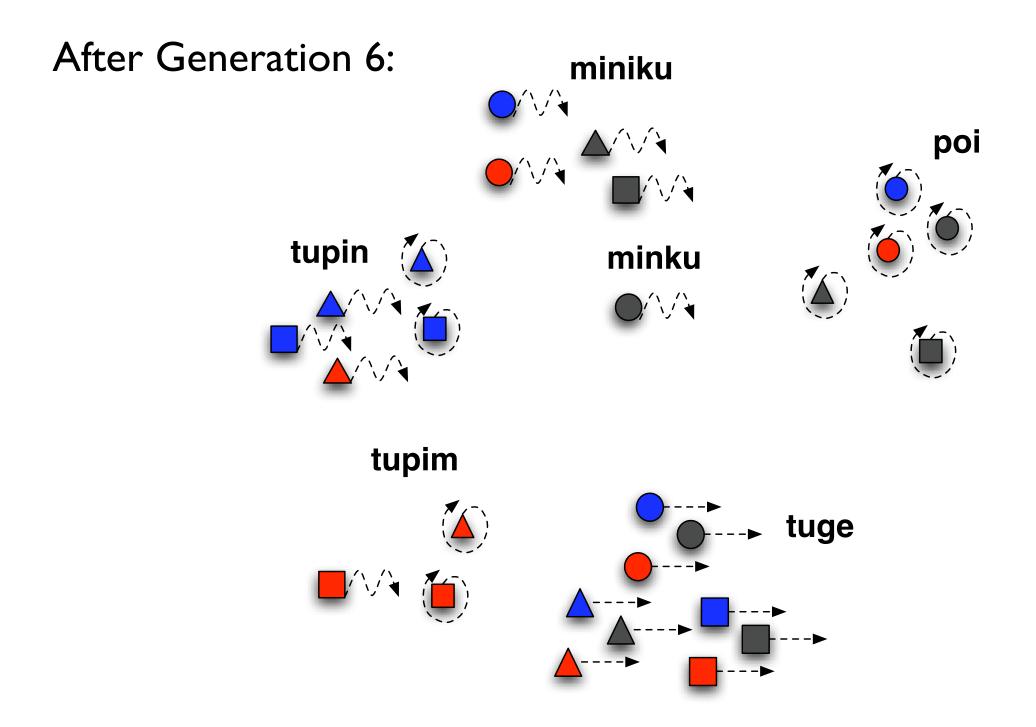
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- This doesn't appear to be the case...

Language becomes systematic



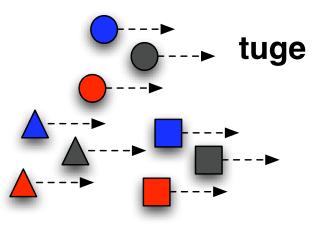


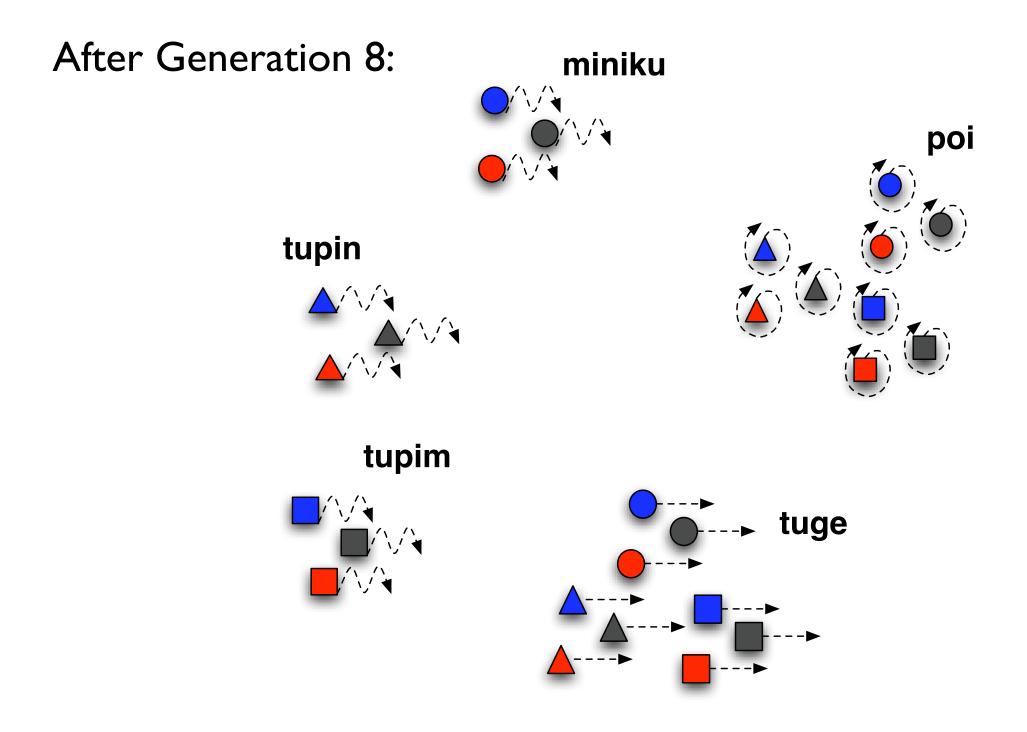


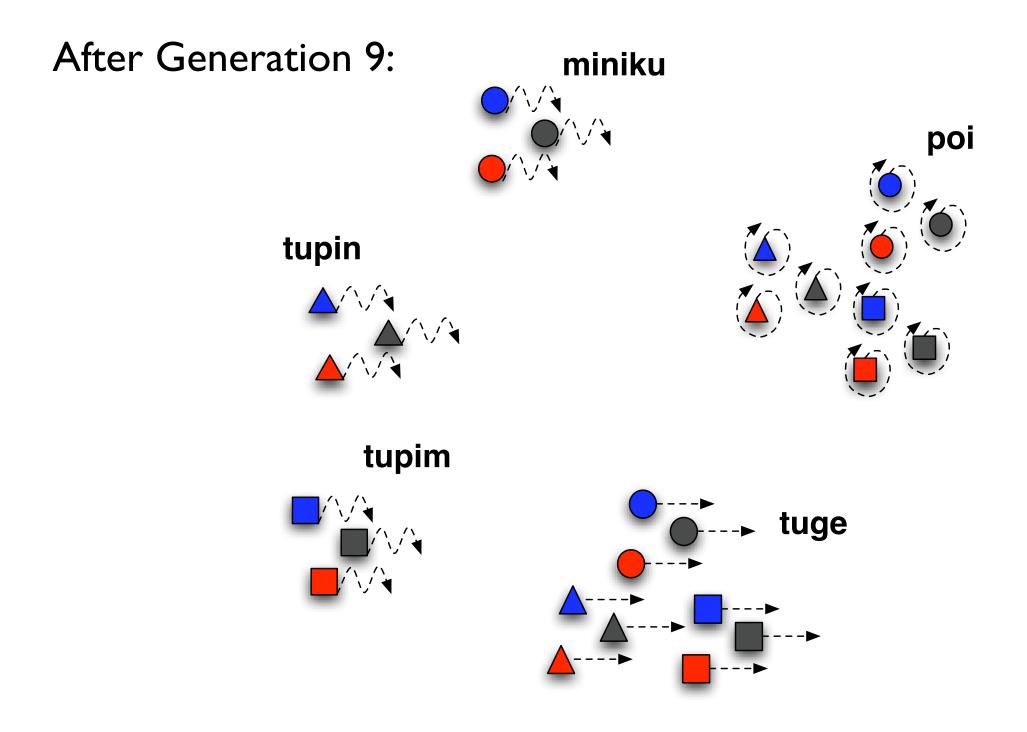
After Generation 7: miniku

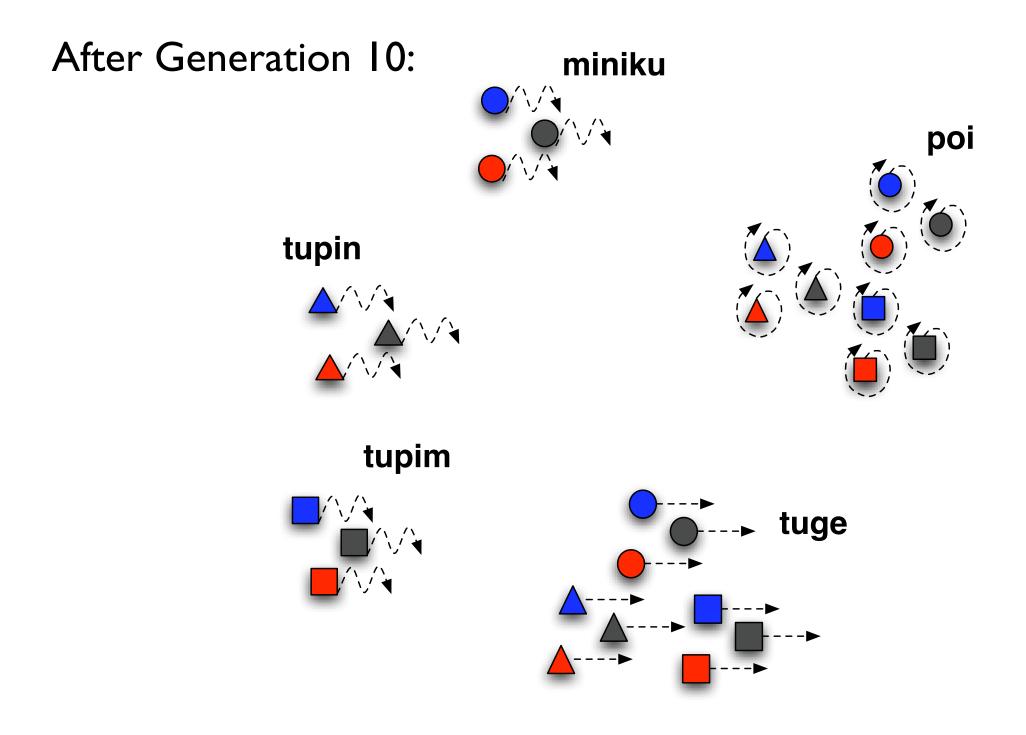
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Language adapts to be structured

- Language adapts
 - Subjects are not aware of this (they aren't even aware they are being shown unseen items!)
 - Systematic underspecification is an adaptation by language to aid its own survival
- Cumulative cultural adaptation without intention

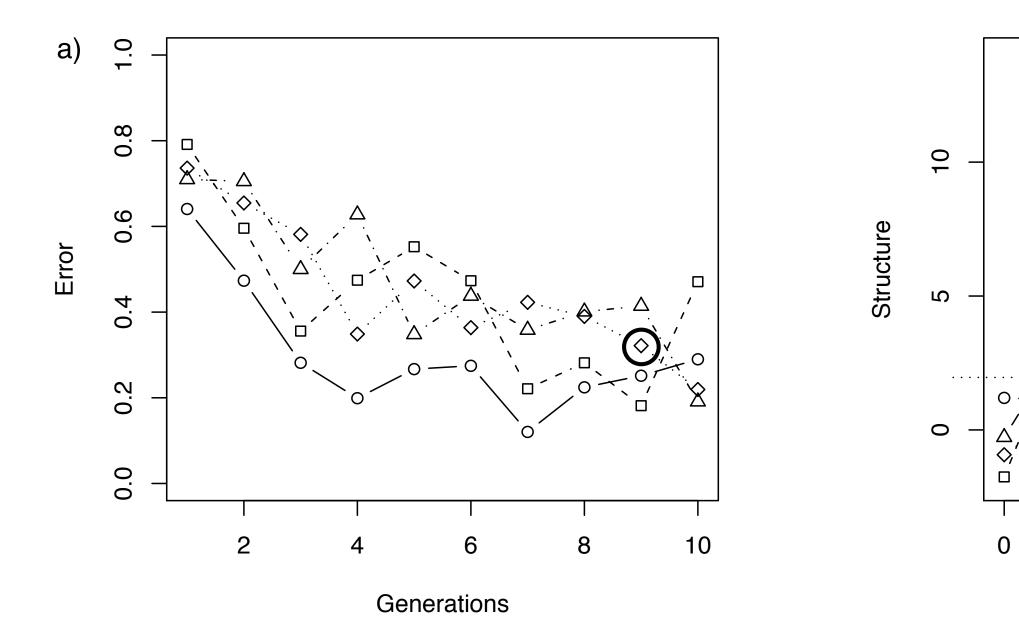
More interesting structure?

- In reality language exhibits compositional structure (e.g. morphology, syntax) that makes it learnable and expressive
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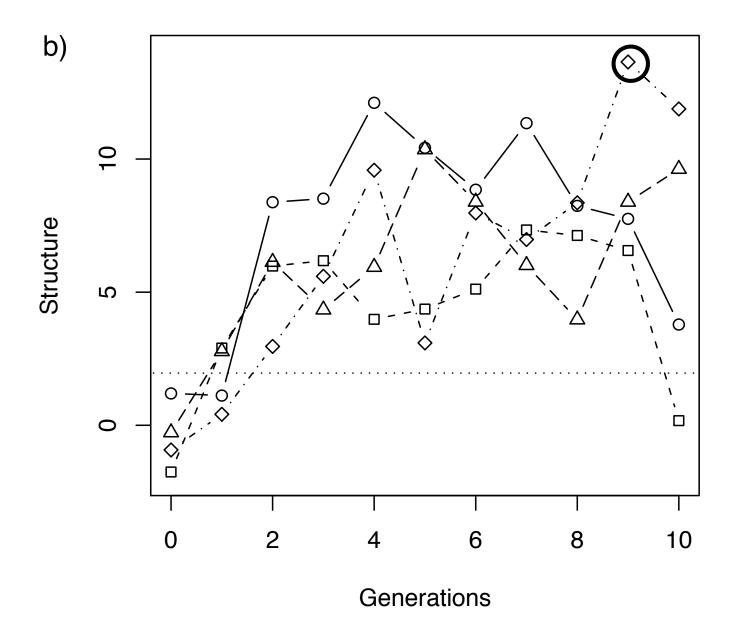
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- Simple modification: filter out all ambiguous items from SEEN set before subjects see them

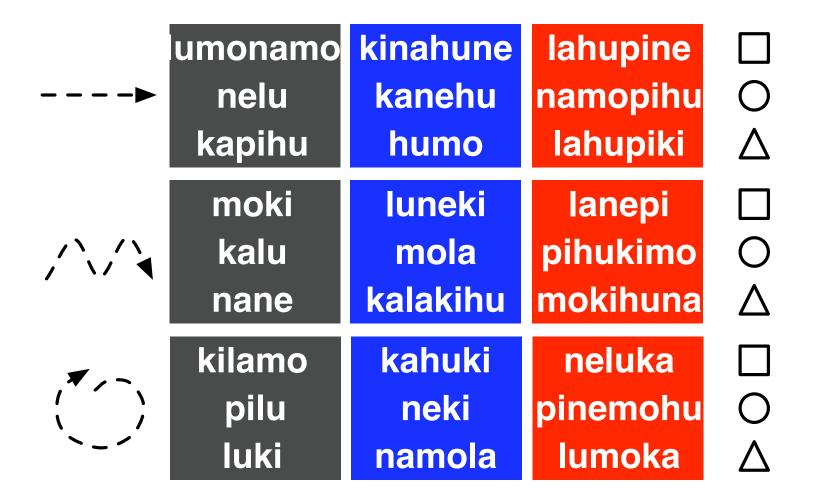
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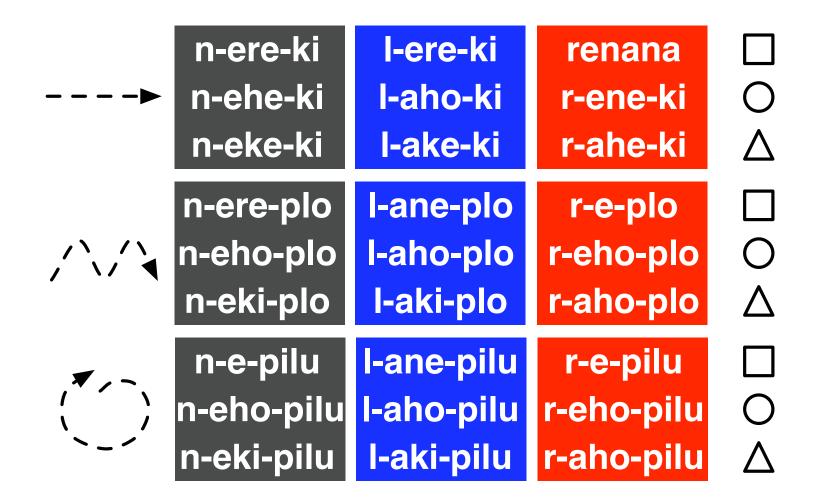
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Example initial language



Example final language (10 "generations" later)



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Adaptation again

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- It must be learned even though:
 - only a sub-sample is seen by learners
 - ambiguous signals are filtered out
- Morphological/syntactic structure is a solution to this problem. Mirrors proposed holistic to compositional protolanguage transition.
- Note: subjects cannot be aware of the filtering, but language structure is very different
 - Demonstrates that adaptation is *non-intentional*
 - Culture gives us design without a designer

Motivations for innateness: complex design

 "Evolutionary theory offers clear criteria for when a trait should be attributed to natural selection: complex design for some function, and the absence of alternative processes capable of explaining such complexity. Human language meets these criteria." (Pinker & Bloom 1990)





Motivations for innateness: complex design

- "Evolutionary theory offers clear criteria for when a trait should be attributed to natural selection: complex design for some function, and the absence of alternative processes capable of explaining such complexity. Human language meets these criteria." (Pinker & Bloom 1990)
- But iterated learning provides just such a process





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 - Natural induction: "right" answer defined by outside forces - potentially difficult problem
 - Cultural induction: problem determined by the very biases that learners themselves have. Any biases you have are guaranteed to be helpful because target is shaped by similar learners
- The less data learners have, the better the situation gets because data increasingly reflects biases (Kirby et al 2007)
- "The poverty of the stimulus solves the poverty of the stimulus" (Zuidema 2003)





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- Huge role for biological evolution to explain the origins of transmissible signal-meaning mappings:
 - "Vocal" learning of complex sequential signals
 - Inference of complex meanings
- Ongoing co-evolutionary pressures to maintain and refine suite of cognitive abilities that support cultural transmission (cf. Lachlan's *cultural trap*, and Deacon's *redistributed selection*)

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- The most poorly understood of these is cultural evolution, yet it has much explanatory power
- To understand this better, we can move from models of learning to models of iterated learning
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- Ultimately, changes our perspective on the biological evolution of our species capacity for language