

# The Significance of Errors to Parametric Models of Language Acquisition

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## **Classification of Language Examples**

Children become fluent despite lack of formal language teaching.

Not every utterance heard is a valid example of the environment language.

How can the child know which utterances are valid?

Every time a child mis-classifies an utterance as valid we get an error.

## Sources of Error

- ▶ Accidental Errors: lapses of concentration, slips-of-the-tongue, interruptions.
- ▶ Ambiguous Environments: bi-lingual environments, diglossia, language change
- ▶ Indeterminacy of Language:
  - ▶▶ Indeterminacy of meaning: “John kissed Kate” vs. “Kate was kissed by John”
  - ▶▶ Indeterminacy of parameter settings: SVO vs. SOV with v2

Require a learning model to attempt to learn from every utterance and be unaffected by misclassification errors.



## The Numbers Game

Game with 2 players:

- ▶ Player One: thinks of a set of numbers that can be defined by a rule.
- ▶ Player Two: attempts to discover the rule defining the set.

Only information available to player two is a stream of examples from player one.

# Deterministic Learners

Gibson and Wexler's Trigger Learner:

➡ Algorithm:

- ➡ attempt to parse with current parameters;
- ➡ change one parameter;
- ➡ adopt new settings if we can analyze an utterance that was previously not analyzable.

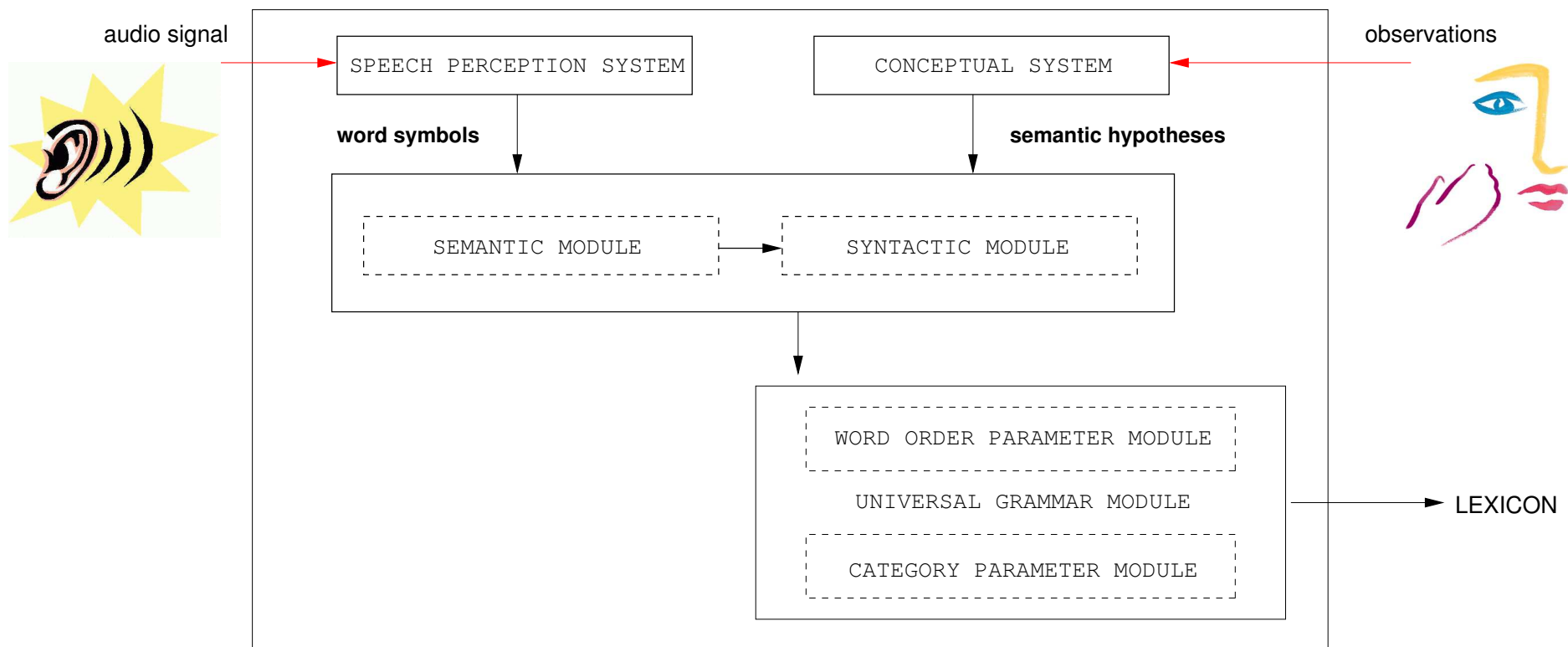
➡ Problems:

- ➡ local maxima;
- ➡ worse case scenario - last utterance seen is an error.

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Gibson E and Wexler K, 1994. Triggers. *Linguistic Inquiry* 25(3): 407-454

# A Robust Learning System



# Semantics Learning Module

Cross Situational Techniques:

➡ Constraining Hypotheses with Partial Knowledge:

If learner knows that: “cheese”  $\mapsto$  **cheese**

and on hearing “Mice like cheese” hypotheses:

**like(mice, cheese)**

**madeOf(moon, cheese)**

**madeOf(moon, cake)**

then we can rule out **madeOf(moon, cake)**

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Siskind J. 1996. A computational study of cross situational techniques for learning word-to-meaning mappings.

*Cognition* 61(1-2):39-91

# Syntactic Learning Module

Hypothesizes categorial grammar categories for a word:

➡ Forward Application ( $>$ )

$$X/Y \quad Y \rightarrow X$$

➡ Backward Application ( $<$ )

$$Y \quad X \backslash Y \rightarrow X$$

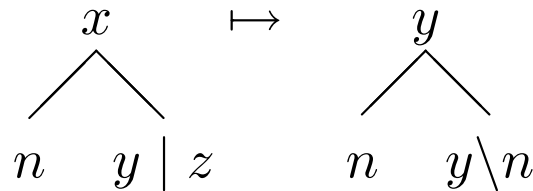
$$\begin{array}{c}
 \textit{Kim} \quad \frac{\textit{likes}}{\textit{(s}\backslash\textit{np})/\textit{np}} \quad \frac{\textit{Sandy}}{\textit{np}} \\
 \hline
 \textit{np} \quad \textit{s}\backslash\textit{np} \quad > \\
 \hline
 \textit{s} < \\
 \textit{-}
 \end{array}$$



# Syntactic Learning Module

Typing Assumption: the semantic arity of a word is usually the same as its number of syntactic arguments.

$\text{verb}(\text{arg1}, \text{arg2}) \mapsto a \mid b \mid c$





## The Universal Grammar

Underspecified inheritance hierarchy:

➡ **Categorial Parameters: 60 parameters**

➡ one per legal syntactic category

➡ **Word Order Parameters: 18 parameters**

➡ e.g. subject direction parameter (SVO,SOV vs. OVS,VSO)

Universal Grammar module consulted whenever syntactic learner returns a valid syntactic category for every word.



## The Sachs Corpus

Natural interactions of a child with her parents:

- ➡ Real child-directed utterances - child's utterances removed;
- ➡ Corpus modeled by Villavicencio;
- ➡ Annotated with semantic representations.

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Villavicencio A. 2002. *The acquisition of a unification based generalized categorial grammar*  
Ph.D Thesis, University of Cambridge.

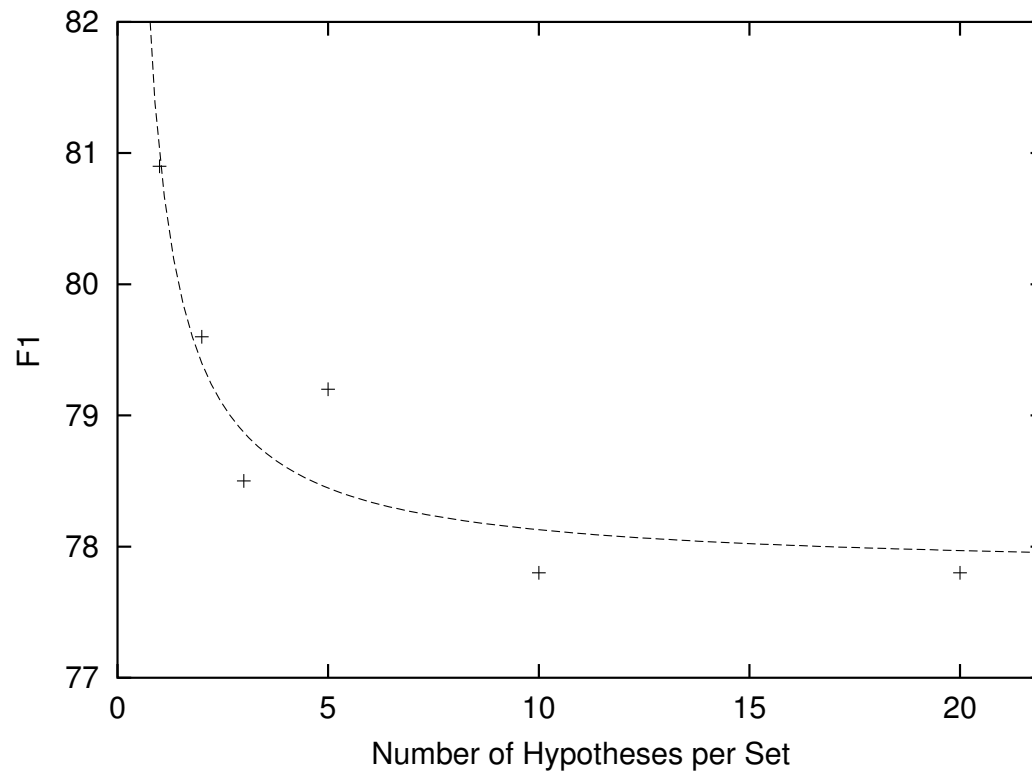


## Exp. 1: Indeterminacy of Meaning

Increasing numbers of semantic hypotheses per utterance:

- ➡ Extra hypotheses chosen randomly.
- ➡ Correct semantic expression was always present in the set.
- ➡ Hypothesis sets of sizes 2, 3, 5, 10 and 20.

## Exp. 1: Indeterminacy of Meaning



## Exp. 2: Indeterminacy of Parameter Settings

Misclassification due to thematic role: “He likes fish”

Possible interpretations:

likes(he, fish) - SVO

likes(fish, he) - OVS

- ➡ Learner was exposed to increasing amounts of misinterpreted thematic role (0% to 50% of all occurrences)



## Exp. 2: Indeterminacy of Parameter Settings

- ➡ mis-classification varied between 0% and 50% at 10% intervals:
  - ➡ 9 word-order-parameters set;
  - ➡ 13.5 word-order-parameters correct according to target (due to inheritance).
  - ➡ 45% difference in speed of convergence between error-free and maximum thematic-role-error case.



## Conclusions

Errors due to misclassification of language examples are likely.

Deterministic parametric learners have problems handling errors.

A statistical error-handling learner may be robust to errors.

Indeterminacy of language is just another case of misclassification.

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