# On Usage and Grammar 

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## Seeing vs. Believing

- Mismatch between the input and the output
- Different kinds of input do different things (Yang 2002, 2004, 2005; see also Fodor \& Sakas 2002, Pearl 2007)
- Data, model, and inference, for both the child and the scientist
- what's the grammar like based on the sample data?
- what to do with the grammar given the sample data?


## Usage/Item/Constructivist Language

- "the 2-year-old child's syntactic competence is comprised totally of verb-specific constructions with open nominal slots. (Tomasello 2000, p214, Cognition, 2000, TICS, etc.)
- Verb Island Hypothesis (Tomasello 1992): "Of the 162 verbs and predicate terms used, almost half were used in one and only one construction type, and over two-thirds were used in either one or two construction types."
- Determiners (Pine \& Lieven 1997): far below chance overlap in "a-N" and "the-N" combinations, suggesting that determiner is not mastered early on (contra Valian 1986)
- morphology (Pizutto \& Caselli 1994): 47\% of all verbs were used in 1 person-number agreement ( 6 forms are possible), $\mathbf{4 0 \%}$ were used in 2 or 3 forms, and only $\mathbf{1 3} \%$ were used in 4 or more.
- Statistical validations?


## George Kingsley Zipf

- Few words are used frequently, and they are very frequent
- Most words are used very rarely, exactly only once
- More precisely, the rank and the frequency of words multiple to a constant
- this can be visualized by plotting log(rank) against log (frequency): you'll get a straight line

$$
f=\frac{C}{r} \text { where } C \text { is some constant }
$$

## Plotting the Brown corpus

Top: words, Bottom: pseudowords


## But why?

- Zipf (1949): Principle of Least Effort (more frequent words tend to be shorter)
- cf. the debate between Simon \& Mandelbrot in the 196 os
- Chomsky (1958): define words as strings between, say, "e"
- the_manager_is_late
- th
- _manag
- r_is_lat


## Zipfian presence



- Zipf like distributions can be observed all over the place
- In language, the slope of log-log fit is very close to 1 (Baroni 2008)


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- Determiners (Pine \& Lieven 1997): far below chance overlap in "aN " and "the-N" combinations, suggesting that determiner is not mastered early on (contra Valian 1986)
- morphology (Pizutto \& Caselli 1994): $47 \%$ of all verbs were used in 1 person-number agreement ( 6 forms are possible), $40 \%$ were used in 2 or 3 forms, and only $\mathbf{1 3} \%$ were used in 4 or more.
- But what's the Null Hypothesis?


## Diversity of Usage

- Valian (1986): the knowledge of the category determiner fully productive by $2 ; 0$, virtually no errors
- Pine \& Lieven (1997), Pine \& Martindale (1996): No, because overlap is much lower than, say, even $50 \%$

$$
\text { overlap }=\frac{\# \text { of nouns with BOTH the AND } a}{\# \text { of nouns with EITHER the OR } a}
$$

- The same logic behind Tomasello's Verb Island Hypothesis
- But Valian, Solt \& Stewart (2008, J. Child Language) found no difference between kids and their mothers!
- Brown corpus: overlap for the and a is $25.2 \%$


## The Productivity Hypothesis

- Assume DP DN is completely productive: combination is independent
- D a/the, N cat, book, desk, ...
- substitute DP for VP, PP, inflections ...
- Given the Zipfian distribution of words, overlap is necessarily low
- Most nouns will be sampled only once in the data: zero overlap
- If a noun is sampled multiple times, there is still a good chance that it is paired with only one determiner, which also results in zero overlap
- If the determiner frequencies are Zipfian as well, this makes the overlap even lower


## Determiner-Noun Usage

- "the bathroom" » "a bathroom"
- "a bath" > "the bath"
- Brown corpus: $75 \%$ of singular nouns occur with only the or a
- $25 \%$ of the remainders are balanced
- favored vs. less favored $=2.86: 1$
- This is also true of CHILDES data, for both children and adults ( 12 samples)
- $22.8 \%$ appear with both, favored vs. less favored $=2.54: 1$


## Zipfian Probabilities

- Assume that there are N words and their frequencies are Zipfian
- In the child production data, singular nouns have the slope of -1.08 (very close to perfect Zipfian fit)
- ist word has frequency of C
- 2 st word has frequency of $\mathrm{C} / 2$
- rth word has frequency of $\mathrm{C} / \mathrm{r}$


## Zipfian Probabilities

- The $r$ th word has probability of $\mathbf{P}_{\mathbf{r}}$

$$
\begin{gathered}
\frac{C / r}{\frac{C}{1}+\frac{C}{2}+\ldots+\frac{C}{N}} \\
\frac{1}{r H_{N}} \text { where } H_{N}=\sum_{i=1}^{N} \frac{1}{i}
\end{gathered}
$$

- In a sample size of $\mathbf{S}$, it has an expected occurrence of

$$
S P_{r}=\frac{S}{r H_{N}}
$$

- S and N can be directly obtained from CHILDES data


## Expected overlap of the rth Noun

The expected overlap for $N_{r}$ is 1 - (expected probability of $N_{r}$ appearing with exactly one determiner for all $S P_{r}$ trials), or

$$
\begin{equation*}
1-\sum_{d=1}^{D} p_{d}^{S P_{r}} \tag{1}
\end{equation*}
$$

If the determiners are also Zipfian, we have

$$
p_{d}=\frac{1}{d H_{D}} \text { where } H_{D}=\sum_{i=1}^{D} \frac{1}{d}
$$

We add up (1) for all N nouns and divide that by $N$ : that is expected overlap

## $\mathrm{N}=50, \mathrm{~S}=100$



- The expected probability of a noun having overlap


## Empirical Data

- Children: Adam, Eve, Sarah, Nina, Naomi, Peter
- All children in CHILDES that started at one/two word stage and with reasonably large longitudinal samples
- Used a variant of the Brill tagger (1995) with statistical information for disambiguation (gpostt1.sourceforge.net), which has an accuracy of about $97 \%$
- Standard procedure in data processing:
- remove annotation markers
- repetitions count only once ("a doggie! a doggie! a doggie!")
- extract D- $\mathrm{N}_{\text {singular }}$ pairs


## Empirical/Theoretical Results

| Child | Sample <br> Size $(S)$ | $a \&$ the Noun <br> types | $a$ or the Noun <br> types $(N)$ | Overlap <br> (expected) | Overlap <br> (empirical) | $S$ <br> $\overline{\mathrm{~N}}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Naomi $(1 ; 1-5 ; 1)$ | 884 | 60 | 349 | 19.0 | 19.8 | 2.53 |
| Eve $(1 ; 6-2 ; 3)$ | 831 | 61 | 283 | 22.7 | 21.6 | 2.94 |
| Sarah $(2 ; 3-5 ; 1)$ | 2453 | 187 | 640 | 26.4 | 29.2 | 3.83 |
| Adam $(2 ; 3-4 ; 10)$ | 3729 | 252 | 780 | 32.0 | 32.3 | 4.78 |
| Peter $(1 ; 4-2 ; 10)$ | 2873 | 194 | 480 | 43.0 | 40.4 | 5.99 |
| Nina $(1 ; 11-3 ; 11)$ | 4542 | 308 | 660 | 47.2 | 46.7 | 6.88 |
| First 100 | 600 | 53 | 243 | 19.6 | 21.8 | 2.47 |
| First 300 | 1800 | 141 | 483 | 26.7 | 29.1 | 3.73 |
| First 500 | 3000 | 219 | 640 | 32.3 | 34.2 | 4.68 |
| Brown corpus | 20650 | 1175 | 4664 | 23.8 | 25.2 | 4.43 |

also considered the first 100, 300, 500 tokens of the six children
paired t- and Wilcoxon tests reveal no difference

## Empirical vs. Expected



## Why Variation

- Some children have higher overlap than others (and Brown)
- Overlap is determined by how many nouns (out of N ) can be expected to be sampled more than once, or

$$
\begin{gathered}
S \frac{1}{r H_{N}}>1 \\
r=\frac{S}{H_{N}} \approx \frac{S}{\ln N}
\end{gathered}
$$

- Overlap is a monotonically increasing function of

$$
\frac{S}{N \ln N} \text { or } \approx \frac{S}{N}
$$

## Analysis of Variation

| Child | Sample <br> Size $(S)$ | $a$ \& the Noun <br> types | $a$ or the Noun <br> types $(N)$ | Overlap <br> (expected) | Overlap <br> (empirical) | $\overline{\mathrm{N}}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Naomi $(1 ; 1-5 ; 1)$ | 884 | 60 | 349 | 19.0 | 19.8 | 2.53 |
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## $r=0.986, \mathrm{p}<0.00001$

## Interim Conclusion

- Children's determiner usage is consistent with the hypothesis of fully productivity.
- We need a theory for how the child gets there (Yang 2002, 2005)
- It is premature to conclude, based on low overlap data, that child language is item-based
- Item-based learning needs to make some quantitative predictions about what to expect


## An attempt at item-based learning

- central tenet of frequency and memorization
- model the learner as a list of joint D-N pairs with their associated frequency
- sample from the list with their joint frequencies
- BIG learner: list consists of 1.1 million child-directed utterances
- small learner: list consists of the child-directed utterance for each particular child
- calculate the overlap for the sampled D-N pairs, averaging over 1000 trials


## item-based learners

| Child | Sample <br> Size $(S)$ | Overlap <br> (BIG learner) | Overlap <br> (small learner) | Overlap <br> (empirical) |
| :---: | :---: | :---: | :---: | :---: |
| Eve | 831 | 16.0 | 17.8 | 21.6 |
| Naomi | 884 | 16.6 | 18.9 | 19.8 |
| Sarah | 2453 | 24.5 | 27.0 | 29.2 |
| Peter | 2873 | 25.6 | 28.8 | 40.4 |
| Adam | 3729 | 27.5 | 28.5 | 32.3 |
| Nina | 4542 | 28.6 | 41.1 | 46.7 |
| First 100 | 600 | 13.7 | 17.2 | 21.8 |
| First 300 | 1800 | 22.1 | 25.6 | 29.1 |
| First 500 | 3000 | 25.9 | 30.2 | 34.2 |

- paired t- and Wilcoxon tests show significant differences ( $\mathrm{p}<$ 0.005)


## A brief look at verbs

- Tomasello (2000, p214, inter alia)
- Verb Island Hypothesis (Tomasello 1992): "Of the 162 verbs and predicate terms used, almost half were used in one and only one construction type, and over two-thirds were used in either one or two construction types."
- data not available in public
- morphology (Pizutto \& Caselli 1994): 47\% of all verbs were used in 1 person-number agreement ( 6 forms are possible), $40 \%$ were used in 2 or 3 forms, and only $13 \%$ were used in 4 or more.
- data not available in public
- Data from CHILDES data


## Islands all over the map ...



- 1.1 million adult sentences, top 15 transitive verbs (put, tell, see, want, let, give, take, show, got, ask, make, eat, like, bring, hear)
- extracted top 10 "sentence frames" (Tomasello 1992) with nominal objects


## Romance morphology

- $(1,2,3$ person $) \times($ singular, plural $)=6$ possible forms
- Data: entire Italian, Spanish, and Catalan child data and childdirected data
- Part-of-speech tagging preprocessing (freeling)
- thanks to Erwin Chan for his help
- Only looking at finite forms (infinitives do not mark agreement)


## Results

| Subject | I form | 2 forms | 3 forms | 4 forms | 5 forms | 6 forms | token/type |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Italian children | $81.8 \%$ | $7.7 \%$ | $4.0 \%$ | $2.5 \%$ | $1.7 \%$ | $0.3 \%$ | 1.533 |
| Italian adults | $63.9 \%$ | $11.0 \%$ | $7.3 \%$ | $5.5 \%$ | $3.6 \%$ | $2.3 \%$ | 2.544 |
| Spanish children | $80.1 \%$ | $5.8 \%$ | $3.9 \%$ | $3.2 \%$ | $3.0 \%$ | $1.9 \%$ | 2.233 |
| Spanish adults | $76.6 \%$ | $5.8 \%$ | $4.6 \%$ | $3.6 \%$ | $3.3 \%$ | $3.2 \%$ | 2.607 |
| Catalan children | $69.2 \%$ | $8.1 \%$ | $7.6 \%$ | $4.6 \%$ | $3.8 \%$ | $2.0 \%$ | 2.098 |
| Catalan adults | $72.5 \%$ | $7.0 \%$ | $3.9 \%$ | $4.6 \%$ | $4.9 \%$ | $3.3 \%$ | 2.342 |

No major difference between Spanish \& Catalan kids and adults
The main predictor is also $\mathrm{S} / \mathrm{N}$

## A view from afar

- Sparse data problem (Jelinek 1993)
- The biggest challenge in computational linguistics
- Statistical models of language, even simple ones such as bigrams/trigrams, require parameter values: many will have zero occurrence in the sample, however large
- Sparse data problem = Poverty of Stimulus
- This is true at the level of morphology (Chan 2008)


## Zipf Morph



## No full paradigms in data

|  | Millions <br> of words | \# possible <br> verb forms | Max \# forms <br> for any <br> lemma |
| :--- | :---: | :---: | :---: |
| Basque | 0.6 | 22 | 16 |
| Catalan | 1.7 | 45 | 33 |
| Czech | 2.0 | 72 | 41 |
| Greek | 2.6 | 83 | 45 |
| Hungarian | 1.0 | 76 | 48 |
| Hebrew | 2.5 | 33 | 23 |
| Slovene | 2.5 | 32 | 24 |
| Spanish | 2.7 | 51 | 34 |

From Chan (2008)

## More Zipfian than Zipf



- Most things we hear are repetitions: this becomes more significant when we look at linguistic combinations
- Words, bigram, and trigram frequencies
- $40 \%$ of words occur only once, $\sim 80 \%$ of word-pairs occur only once, $>90 \%$ of word triples occur only once


## The view from (not too) afar

- Statistical language learning and parsing
- Input: most of Penn Treebank, a collection of trees manually annotated,
- Training: essentially tally the frequencies of usage for a predefined set of rules in the form of a lexicalized Context-Free Grammar (Collins 1997, Bikel 2004)
- Testing: Try parsing the rest of Treebank VsawP $\rightarrow$ Vsaw NP $S \rightarrow N P \sim V P$ $\mathrm{NP} \rightarrow \mathrm{D} \sim \mathrm{N}$ AuxP $\rightarrow$ AuxP~NP $\mathrm{VP} \rightarrow \mathrm{V} \sim \mathrm{NP}$


Vattack P $\rightarrow$ Vattack NP
VrecommendP $\rightarrow$ Vrecommend NP

## The law of diminishing returns



- The measurement of success is the $\%$ of phrase structure brackets inserted correctly
- The "learner" can quickly learn the general rules of the grammar with a small fraction of the data
- Lexicalized learning pays very little dividend (Bikel 2004)


## Zipf in the Penn Treebank



- No. 1 rule: $\mathrm{PP} \rightarrow \mathrm{P}$ NP
- No. 2 rule: $S \rightarrow$ NP VP


## Grammar despite Usage



Grammar must overcome Usage

