# Approximations of Predictive Entropy Correlate with Reading Times

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Approximate Entropy

# A child XXXXXXX the fish

A child XXXXXXX the fish A child annoyed XXX fish

A child XXXXXXX the fish A child annoyed XXX fish A child annoyed the XXXX

Lexical frequency of the upcoming masked word affects processing

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$$H(X) \stackrel{def}{=} -\sum_{x \in X} P(x) \log P(x)$$
(1)

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Roark et al. (2009) distinguishes two kinds of entropy (over words and preterminals)

$$LexH(w_{1..i-1}) \stackrel{def}{=} -\sum_{w_i \in V} P_G(w_i \mid w_{1..i-1}) \log P_G(w_i \mid w_{1..i-1})$$
(2)  
$$SynH(w_{1..i-1}) \stackrel{def}{=} -\sum_{p_i \in G} P_G(p_i \mid w_{1..i-1}) \log P_G(p_i \mid w_{1..i-1})$$
(3)

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### Roark et al. (2009) showed

- SynH predicts self-paced reading times
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## But

- Small training corpus (V is poor)
- Small test corpus:

 $\sim$  200 sentences,  $\sim$  4000 words, 23 subjects

Natural Stories self-paced reading corpus (Futrell et al., in prep)

- 181 subjects
- 10 narrative texts
- 485 sentences (10256 words)
- Each text followed by 6 comprehension questions
- Events removed if <100 ms or >3000 ms

Parsed using Roark (2001) parser

Fitted with *lmer* 

#### SPACES WERE MASKED

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Α -----

- child -----

----- annoyed ------

----- the -----



Predictor	$\hat{eta}$	$\hat{\sigma}$
Syntactic H	4.53*	0.54
Lexical H	-1.05	0.41

Replication of Roark et al. (2009)

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Replication of Roark et al. (2009)

But Angele et al. (2015) found a lexical frequency effect

$$S_{G}(w_{i}, w_{1..i-1}) \stackrel{def}{=} -\log P_{G}(w_{i} \mid w_{1..i-1})$$
(4)

$$LexH_{G}(w_{1..i-1}) \stackrel{def}{=} \sum_{w_{i} \in V} -P_{G}(w_{i} \mid w_{1..i-1}) \log P_{G}(w_{i} \mid w_{1..i-1})$$
(5)  
$$= \sum_{w_{i} \in V} P_{G}(w_{i} \mid w_{1..i-1}) S_{G}(w_{i}, w_{1..i-1})$$
(6)  
$$= E[S_{G}(w_{i}, w_{1..i-1})]$$
(7)

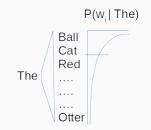
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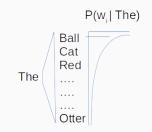
We can use a corpus instead of explicitly computing the expectation

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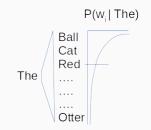
Approximate Entropy



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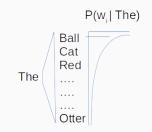


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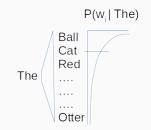


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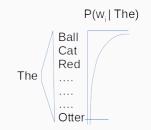
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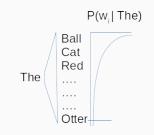
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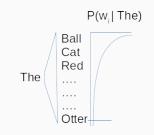
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#### Ex: The boy annoyed the fish.

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We can treat large corpora as our samplers.

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We can try:

• Future Roark surprisal (same distribution as SynH)

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- Future categorial grammar surprisal (tests how specific syntactic prediction is)

Predictor	$\hat{eta}$	$\hat{\sigma}$
Syntactic H	4.62*	0.53
Future Roark Surprisal		0.40
Future N-gram Surprisal	4.05*	0.58
Future Categorial Grammar Surprisal	4.10*	0.74

• Better encoding of  $w_i$  to help with  $w_{i+1}$ 

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- A kind of Uniform Information Density (UID; Jaeger, 2010)
  - Optimizes per-millisecond informativity

### CONCLUSIONS

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- Syntactic uncertainty is fine-grained

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