

NLPitch - ILLC - 26 October 2021

Analysing Human Strategies of Information Transmission as a Function of Discourse Context

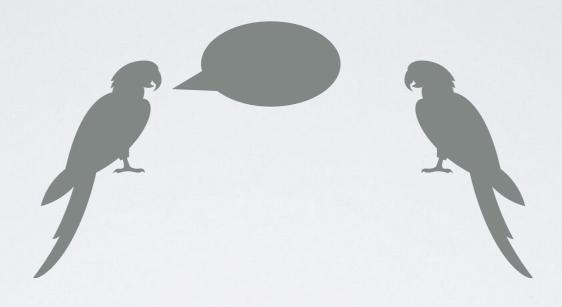
Mario Giulianelli and Raquel Fernández

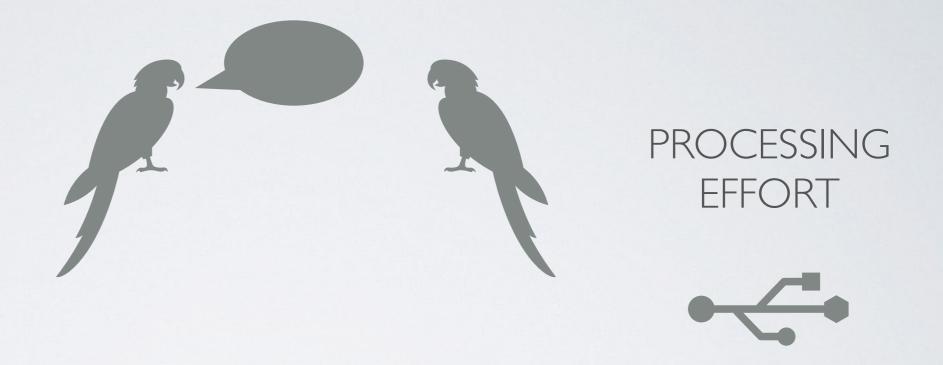
Institute for Logic, Language and Computation University of Amsterdam

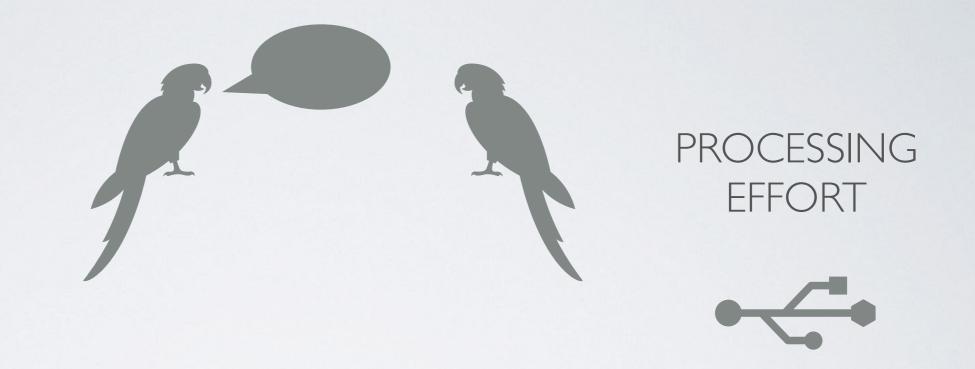
{m.giulianelli|raquel.fernandez}@uva.nl







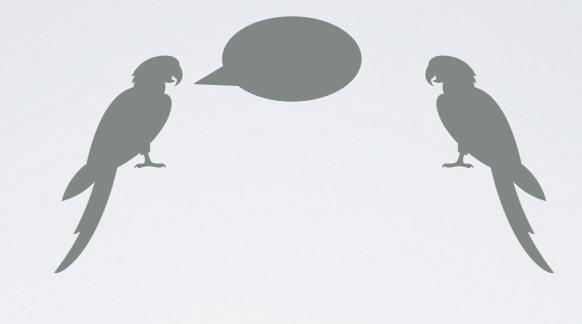




Amcore Financial Inc. said it agreed to acquire Central of Illinois Inc. in a stock swap.

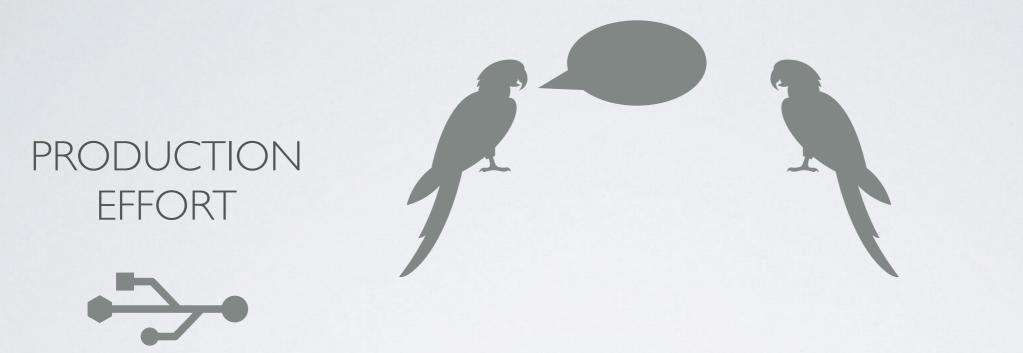
Shareholders of <u>Central</u>, a bank holding company based in Sterling, will receive Amcore stock equal to 10 times <u>Central</u>'s 1989 earnings, Amcore said.

For the first nine months of 1989, <u>Central of Illinois Inc.</u>, <u>a bank holding company</u> <u>based in Sterling</u>, earned \$2 million.



PRODUCTION EFFORT





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EFFORT

PROCESSING



Information Theory (Shannon, 1948)

H(*S*) Shannon information content

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Mutual information

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I(*S*; *C*) Context informativeness

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I(S; C) Context informativeness

$H(S | C) \equiv H(S) - I(S; C)$ In-context information content

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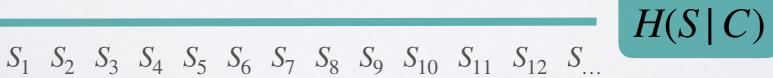
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$H(S \mid C) \equiv H(S) - I(S; C)$

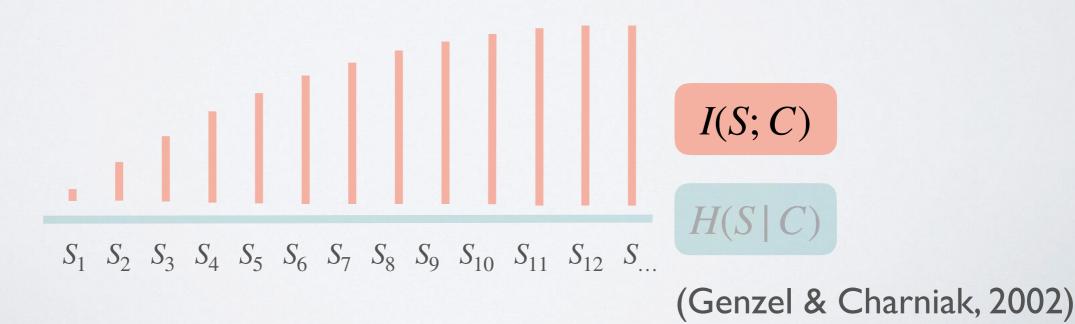
In-context information content



(Genzel & Charniak, 2002)

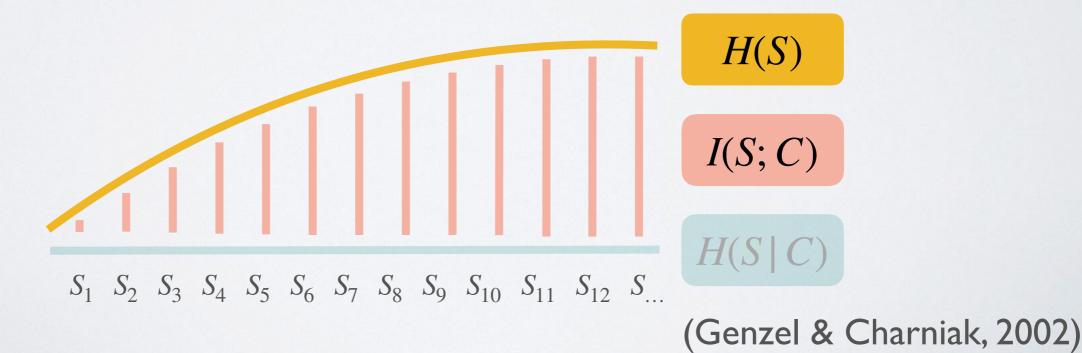


I(*S*; *C*) Context informativeness



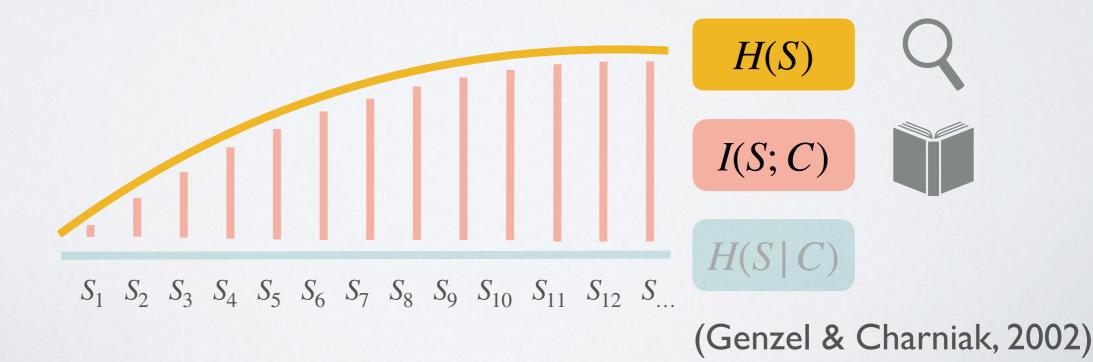


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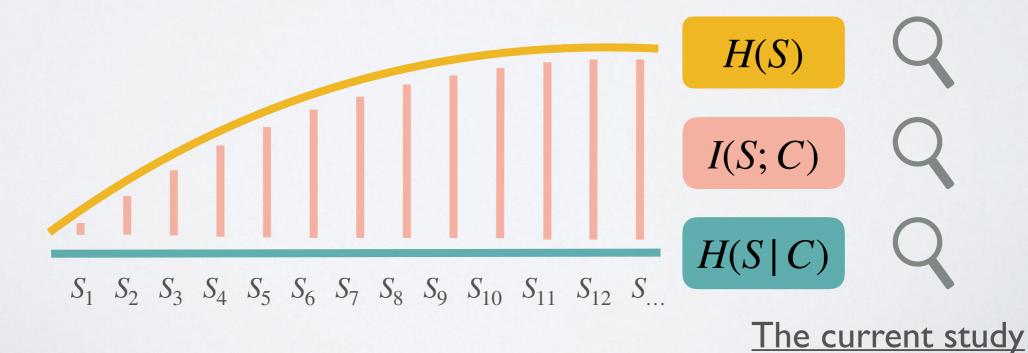
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$$H(S) = -\log_2(S) = -\frac{1}{|S|} \sum_{w_i \in S} \log_2 P(w_i | w_1, \dots, w_{i-1})$$

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 $I(S; C) \equiv H(S) - H(S \mid C)$

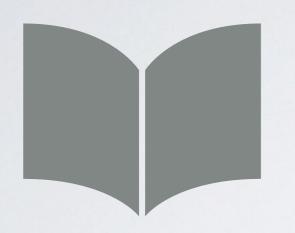
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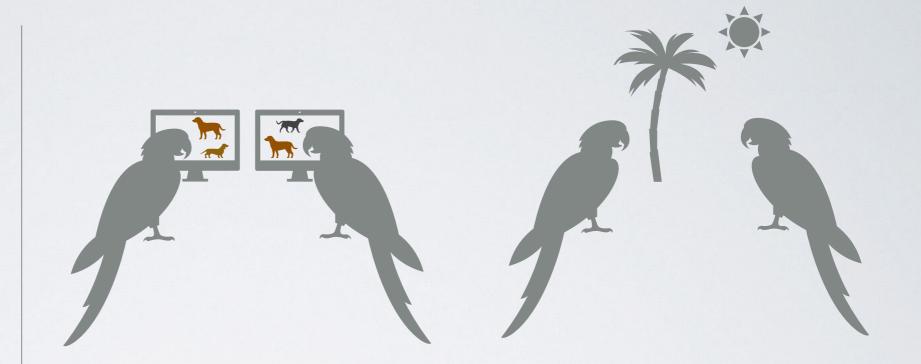
$$I(S; C) = H(S) - H(S | C)$$

 $P(w_i | ...)$ estimates obtained with GPT-2 fine-tuned on 70% of each target corpus (30% held-out for analysis)

Experimental data

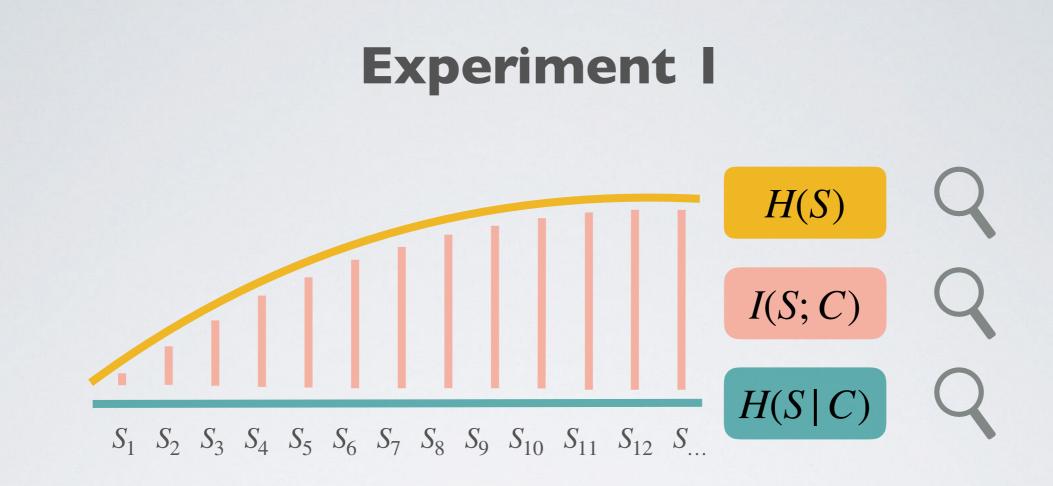


Penn Treebank newspaper articles (Mitchell et al. 1999)

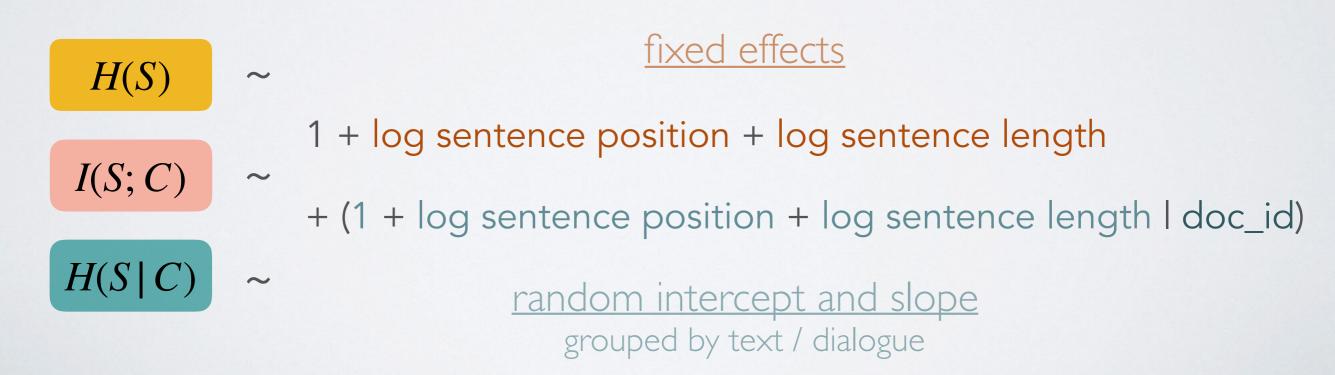


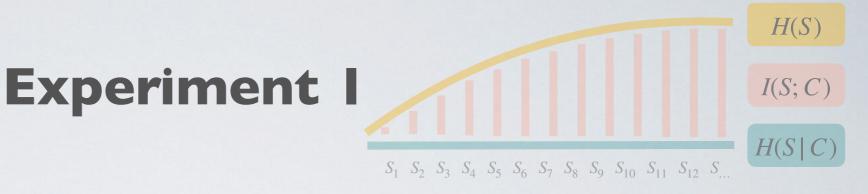
PhotoBook task-oriented written dialogues (Haber et al., 2019)

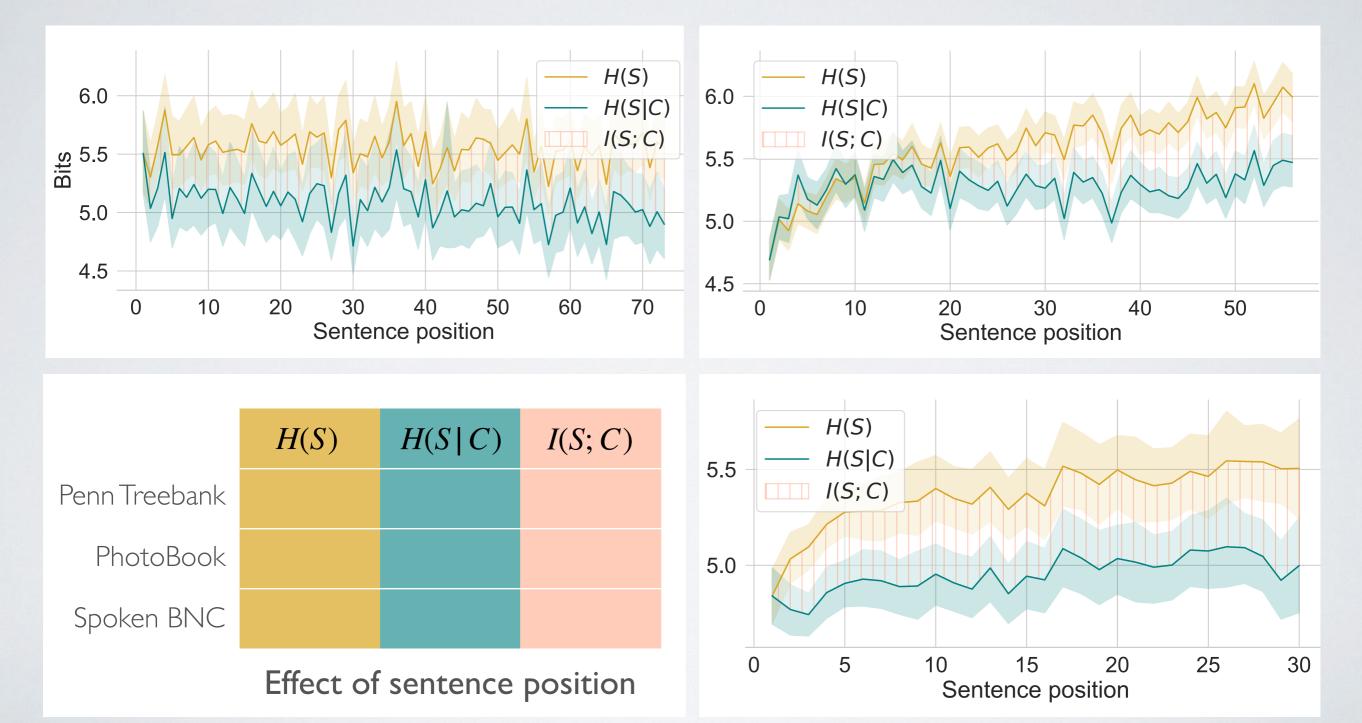
Spoken BNC open domain dialogues (Love et al., 2017)

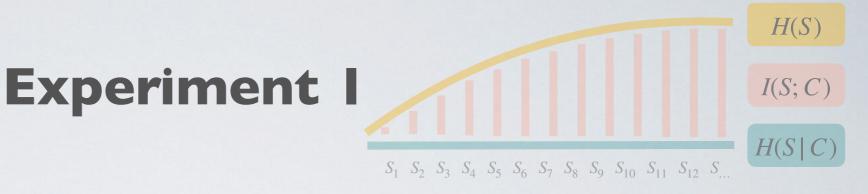


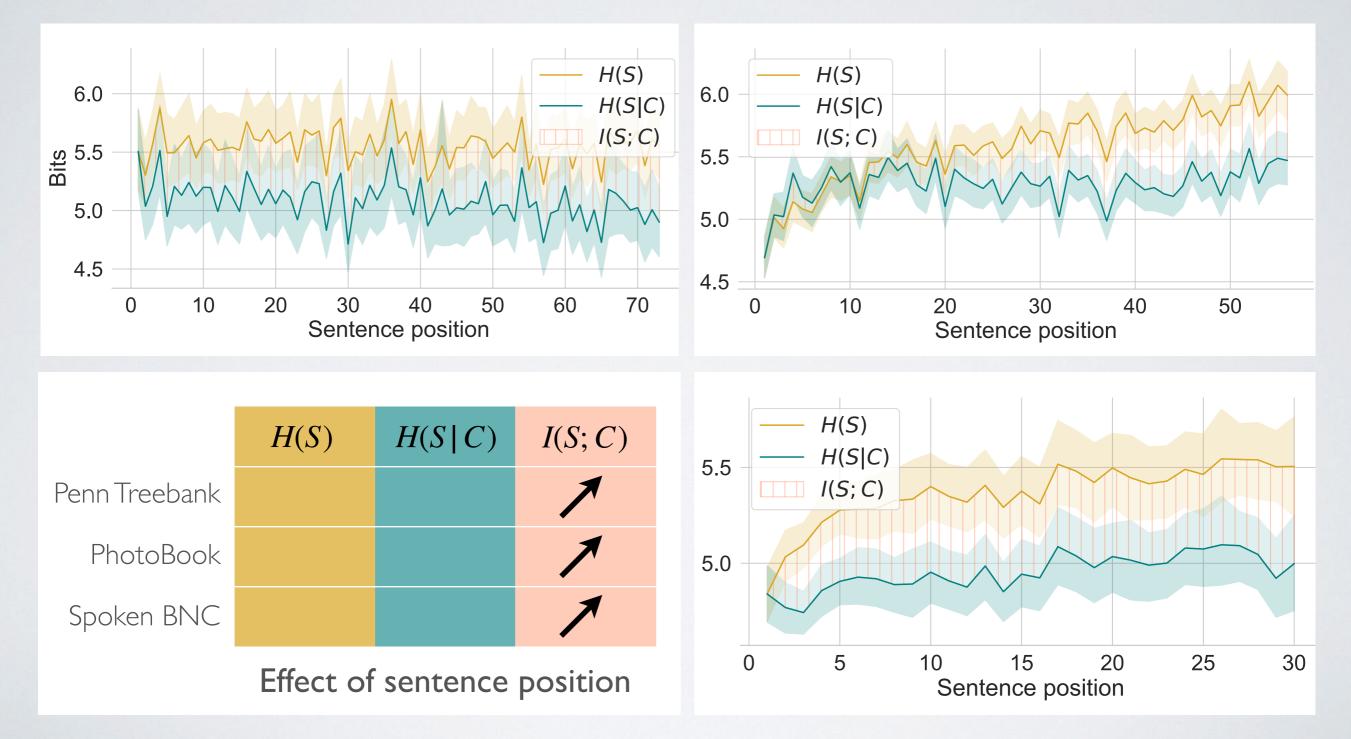


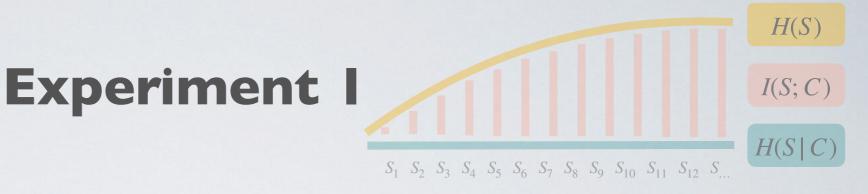


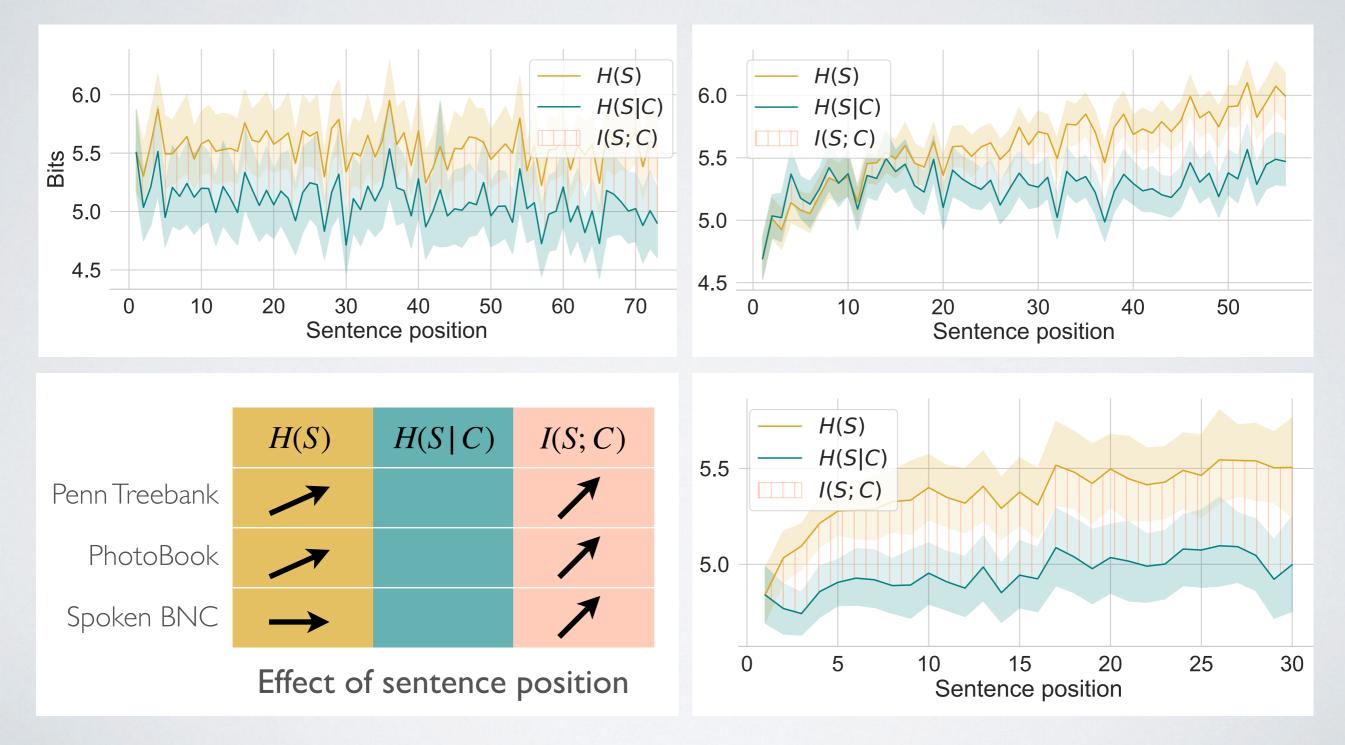


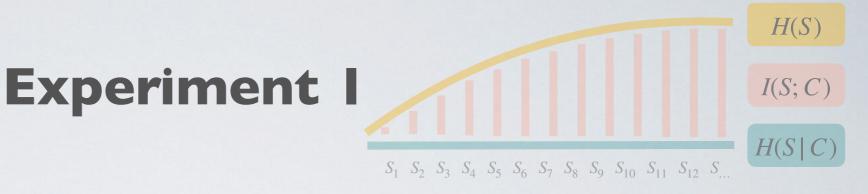












15

Sentence position

20

25

30

H(S)H(S)6.0 6.0 H(S|C) H(S|C)I(S; C)I(S; C) Bits 2.5 5.5 5.0 5.0 4.5 4.5 10 20 30 40 50 60 70 10 50 20 30 40 0 0 Sentence position Sentence position H(S) $H(S \mid C)$ I(S; C)H(S)H(S|C)5.5 I(S; C) Penn Treebank PhotoBook 5.0 Spoken BNC

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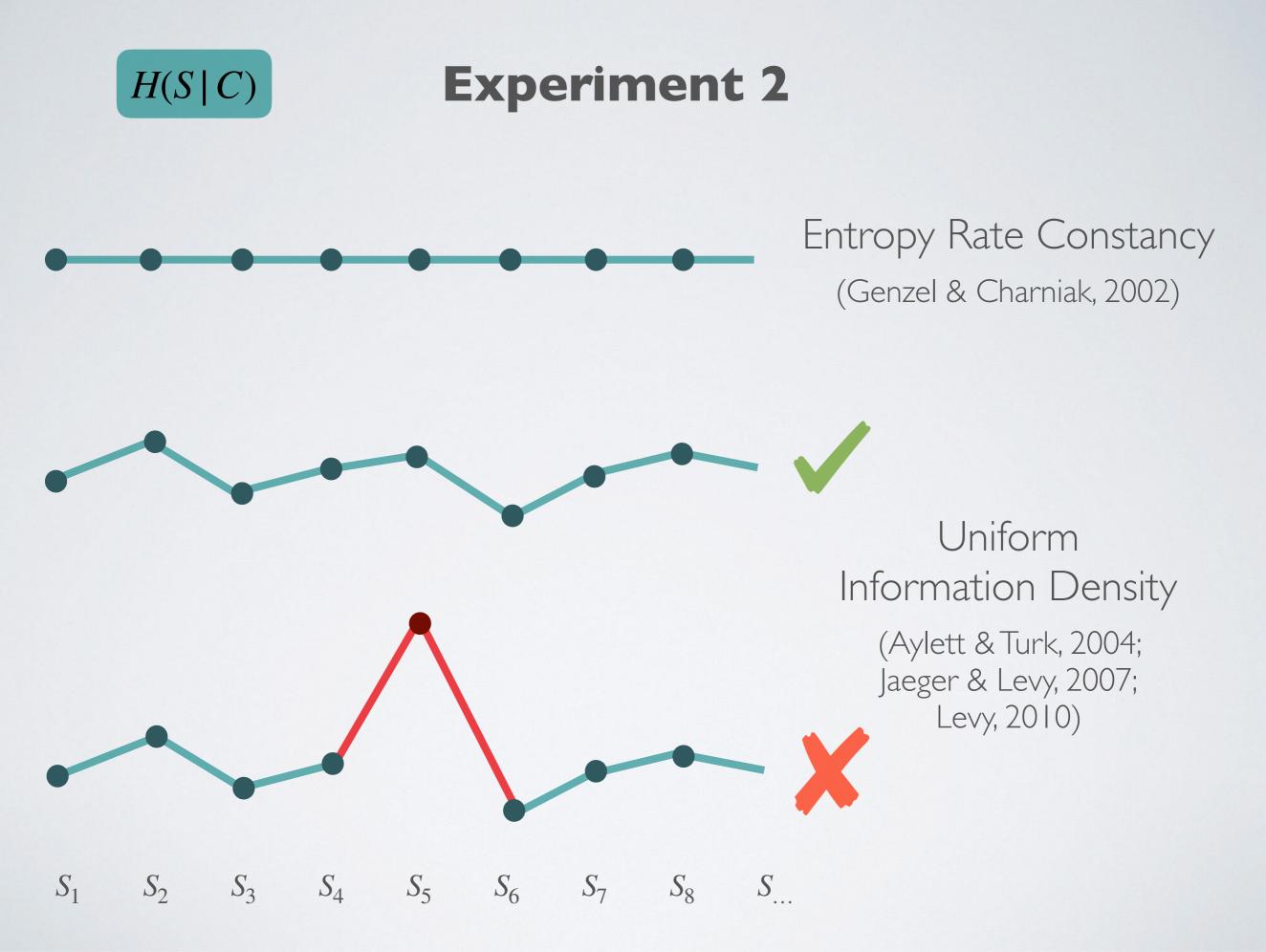
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10

Effect of sentence position

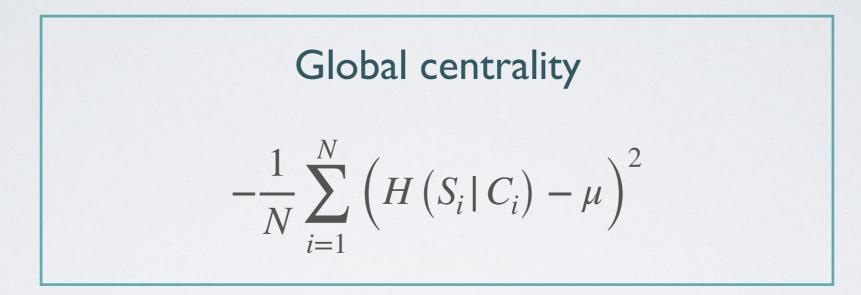


 $S_1 \quad S_2 \quad S_3 \quad S_4 \quad S_5 \quad S_6 \quad S_7 \quad S_8 \quad S_{...}$



Experiment 2

Criteria of uniformity (Collins, 2014)

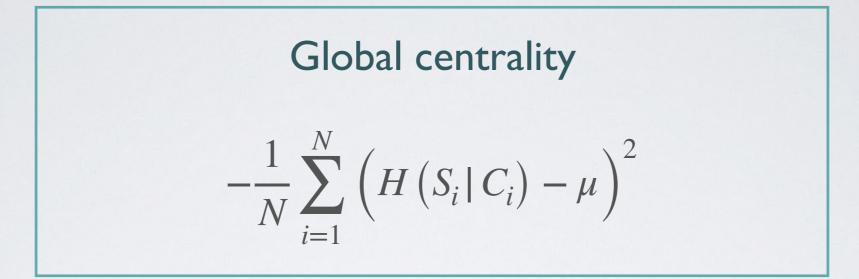


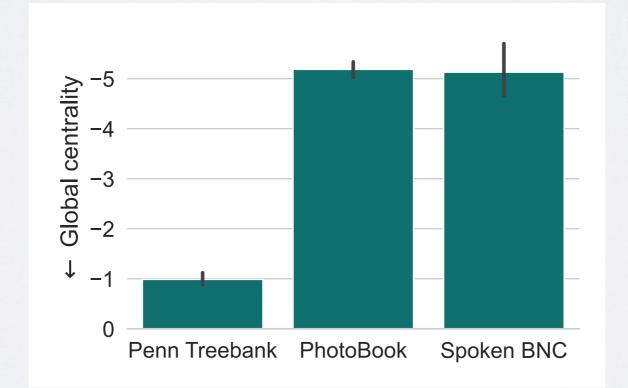
Local predictability

$$-\frac{1}{N}\sum_{i=2}^{N} \left(H\left(S_{i} \mid C_{i}\right) - H\left(S_{i-1} \mid C_{i-1}\right) \right)^{2}$$

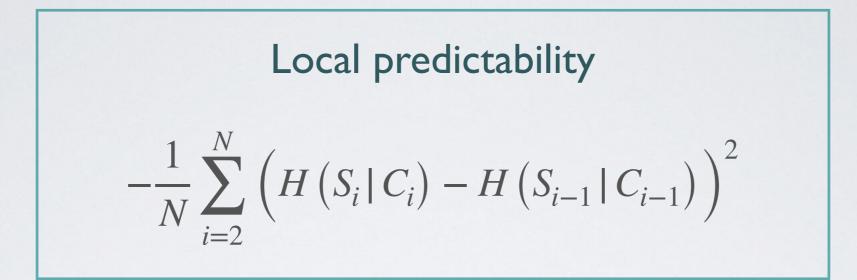
N number of sentences in the text / dialogue μ average information content in the text / dialogue

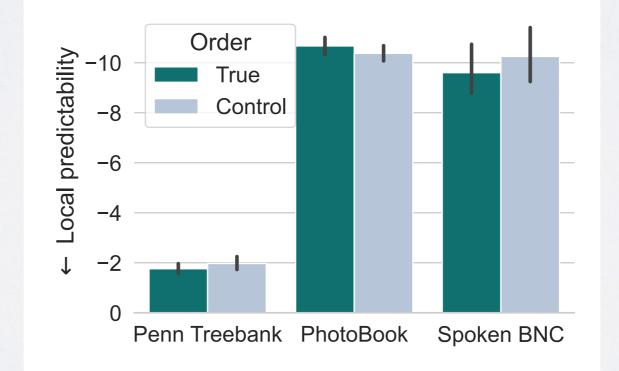
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Summary

We have examined some central tenets of the classic information-theoretic model of communication.

We have used language models to obtain information content estimates for sentences within their discourse context.

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Findings

In newspaper articles information content remains stable, as predicted by the Entropy Rate Constancy principle.

This is not the case for spoken open domain dialogues, nor for written task-oriented dialogues.

Global uniformity is a more faithful criterion than local uniformity.

Discussion

Identifying the relevant components of discourse context \rightarrow Giulianelli, Sinclair, Fernández - EMNLP 2021

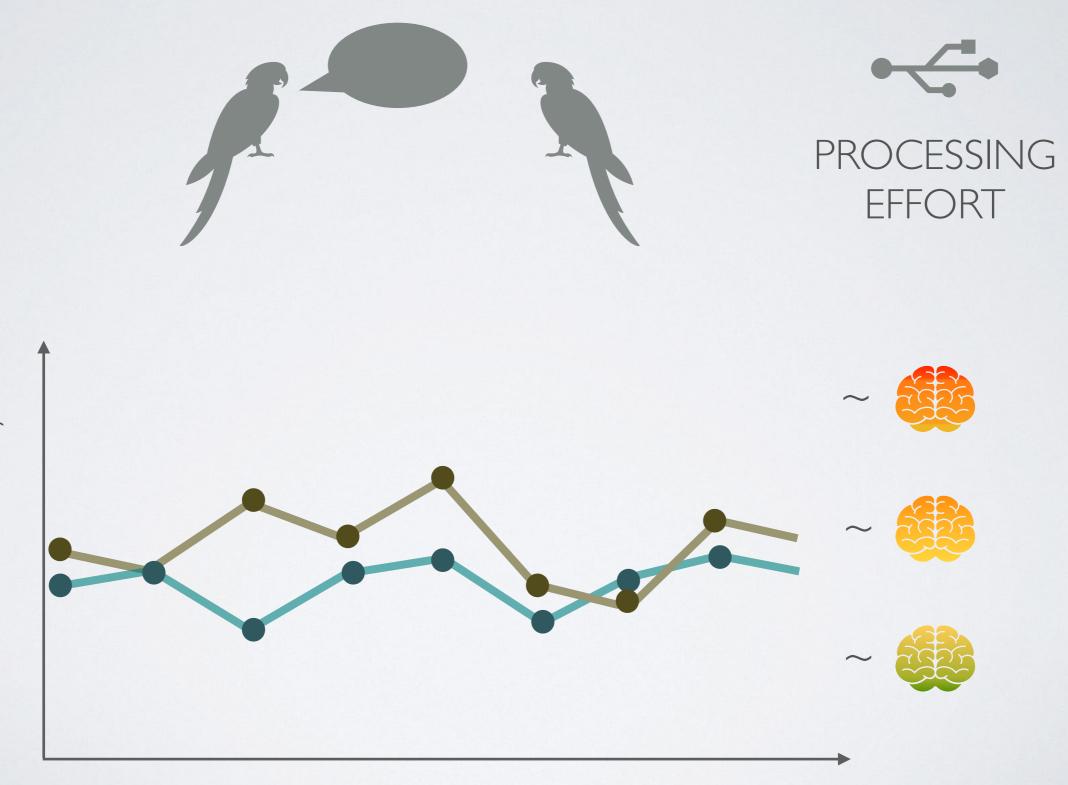
Single vs. multiple addressees

Production and comprehension effort conflated in a single estimate

Architectures and training objectives to emulate the organisation of information density found in human data

APPENDIX

Uniform Information Density (Aylett & Turk, 2004; Jaeger & Levy, 2007)

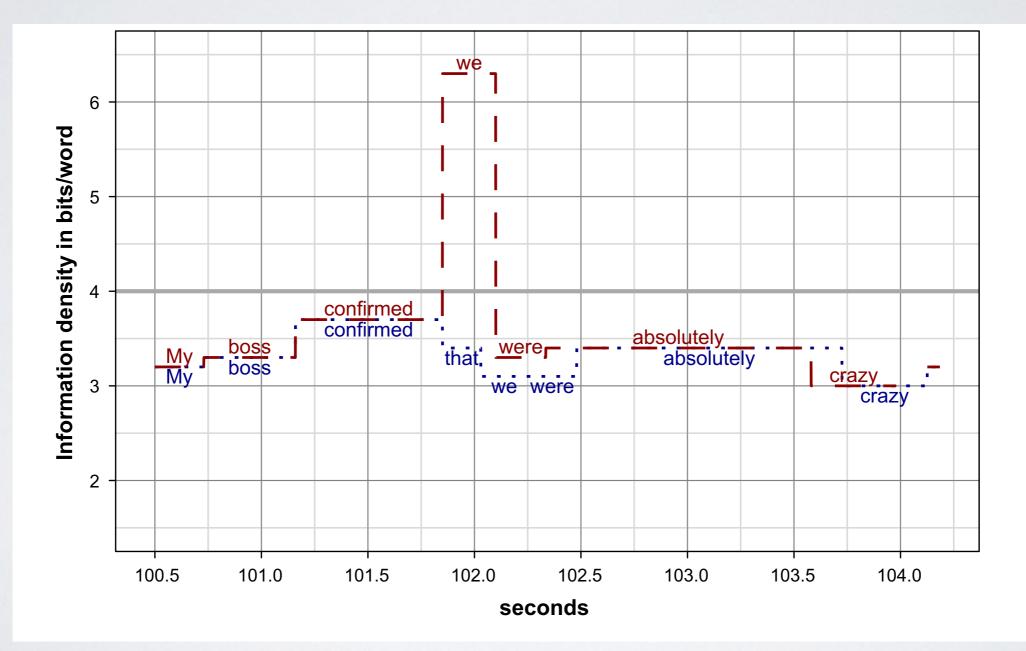


Information density

Uniform Information Density

Example I: Syntax

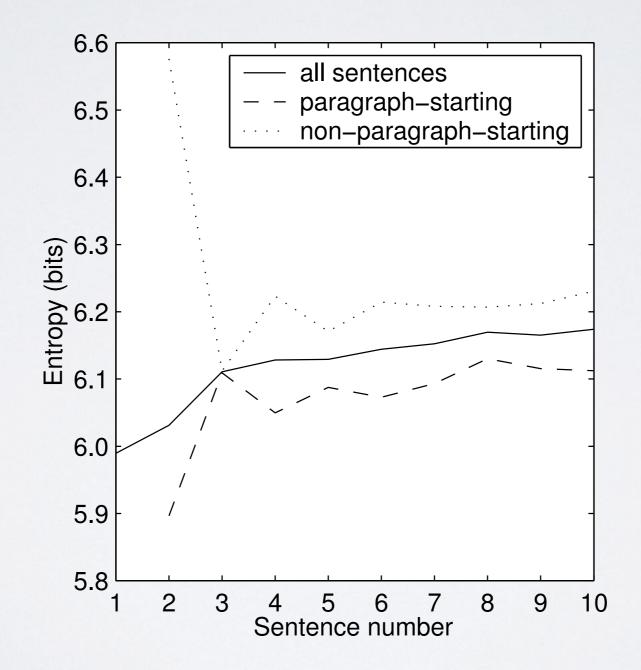
Complementiser that-mentioning (e.g., Jaeger, 2010)



(Jaeger, 2010; Figure 1a)

Uniform Information Density

Example 2: Discourse Entropy rate constancy (Genzel & Charniak, 2002, 2003)



(Genzel & Charniak, 2003; Figure I)

Task-oriented dialogue corpora

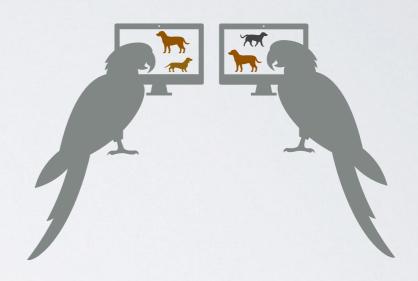
Round 4

В	pink sweater woman and man w/umbrella
Α	yes
В	statue man with umbrella
Α	У
В	guy in black suit with 2 plaid blue umbrellas
Α	no

B guy in black suit with 2 plaid blue umbrellas

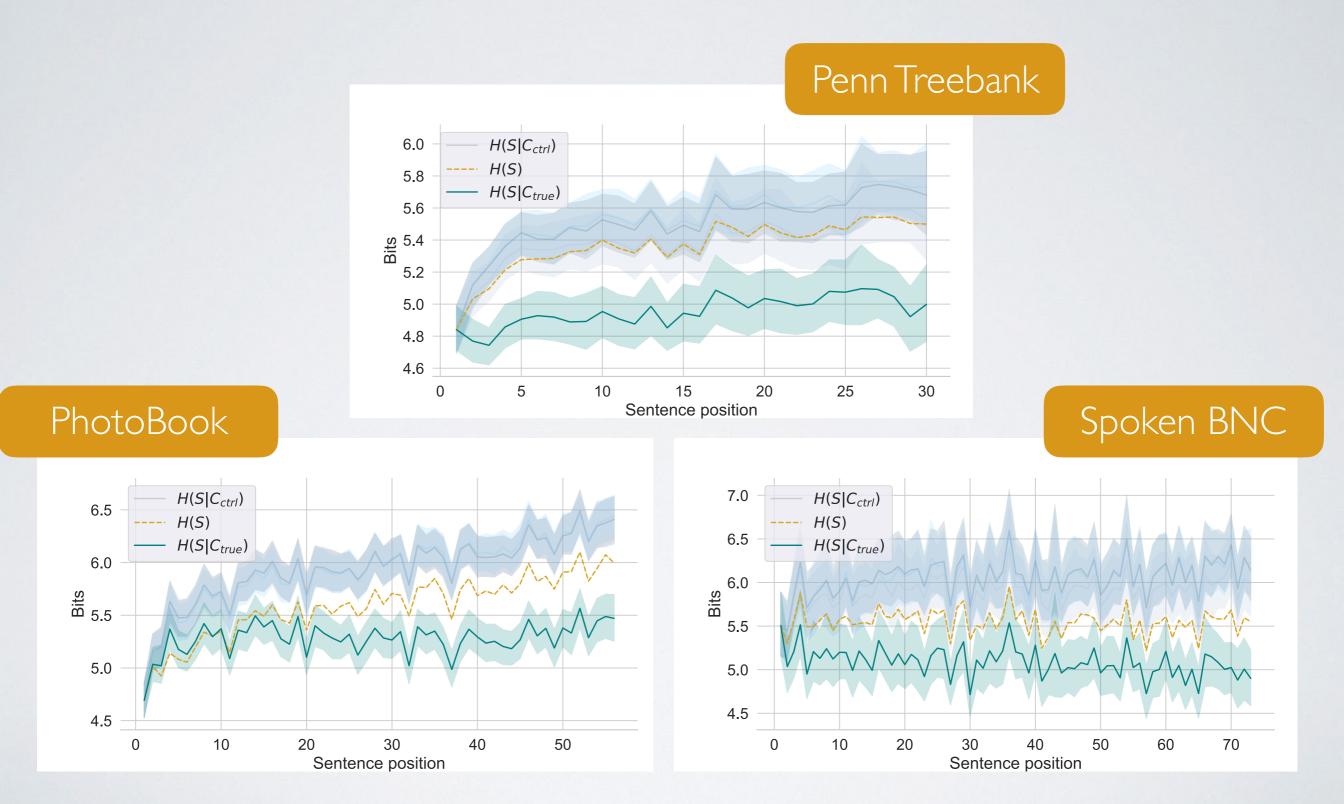
A no

- B lady in pink and guy with umbrella
- A no
- B statue/umbralla again
- A no



PhotoBook written cooperative reference game (Haber et al., 2019)

Control runs



Context informativeness

