ysis of Frequencyand Memory-Based rocessing

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Introduction

working memory fit with current theories of language processing? What influence do memory limitations have a times? Frequency has a huge effect on causes frequency? Perhaps cognitive limitations (e.g. in memprocessing speeds, but what

Frequency Effects

Surprisal

text [Hale, 2001]: How unpredictable a word is given the preceding con-

$$surprisal(x_t) = -\log_2\left(\frac{\sum_{s \in S(x_1...x_t)} P(s)}{\sum_{s \in S(x_1...x_{t-1})} P(s)}\right)$$
 (1)

leaves have $x_1 \dots x_t$ as a prefix. $S(x_1 \dots x_t)$ is the set of syntactic trees whose

Parsing Operations:

L = Word is the Last element of a connected component

Word is the First element of a connected component

Entropy Reduction

is a measure of uncertainty:

$$H(x_{1...t}) = \sum_{s \in S(x_1...x_t)} -P(s) \cdot \log_2 P(s) \tag{2}$$

The reduction in uncertainty caused by observing x_t 2003]:

 $\Delta H(x_{1...t})$ $= \max(0, H(x_{1...t-1}) H(x_{1...t}))$

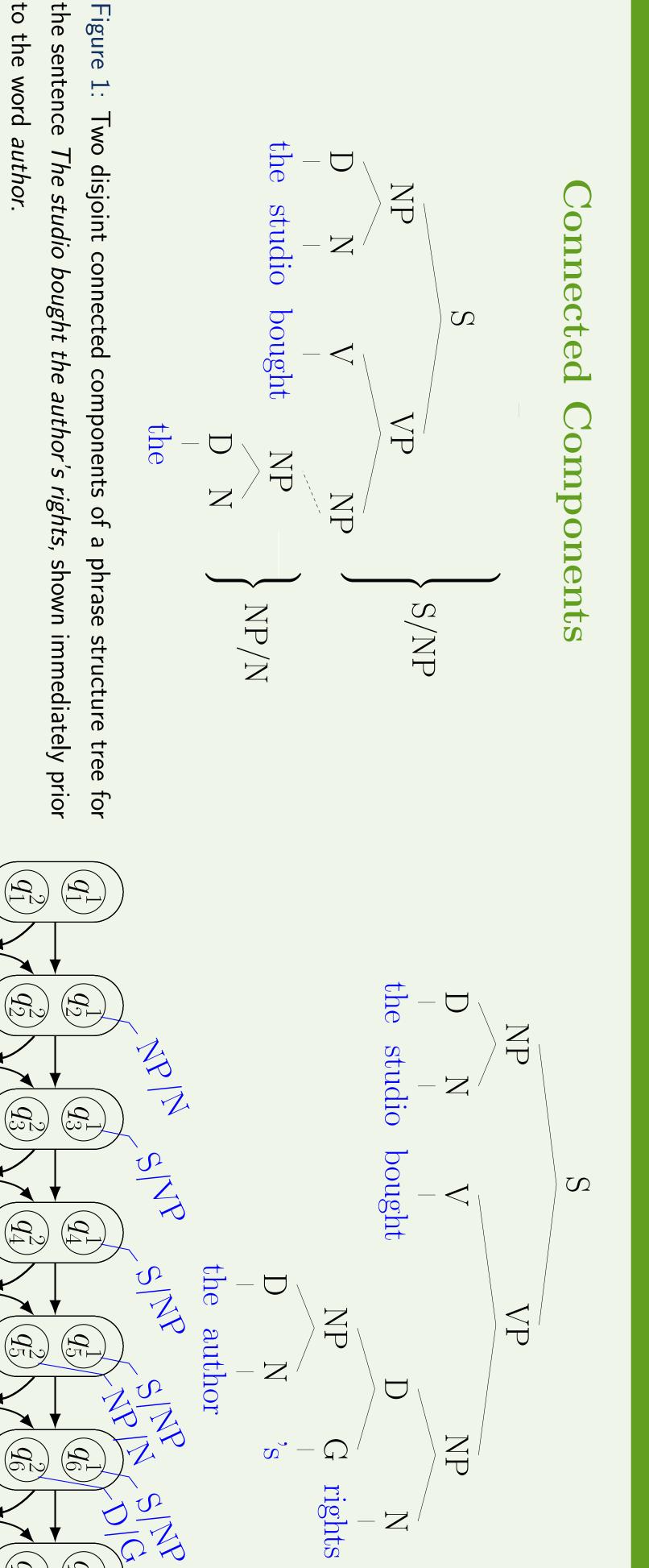
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Memory and Pr ocessing



Memory Theories

the

studio

bought

the

author

rights

Dependency Locality Theory

[Gibson, 2000]

Integration cost {Resolving dependencies Storage cost Difficulty Unresolved dependencies Maintaining dependencies Memory operations Beginning dependencies

ACT-R

[Lewis et al., 2006]

Encoding cost {Beginning a new dependency Retrieval cost {Resolving a dependency Difficulty { Activation decay Similarity interference

Hierarchic Sequential Prediction

Retrieval can be facilitated by re-activations

[Botvinick, 2007]

- Learned sequential associations
- Contextual temporal associations

Difficulty { Temporal cueing Temporal cueing { Resolving embedded dependencies

times

under

each theory of working memory.

Hierarchic sequential

p

ediction is agnostic about the processing speed of F+L- operations

Dynamic Recruitment

[Just and Varma, 2007] Difficult constructions -extra processing resources

Difficulty {Center embeddings Recruitment {Beginning embeddings Release {Completing embeddings

Embedding Difference

[Wu et al., 2010

Difficulty {Dependency maintenance

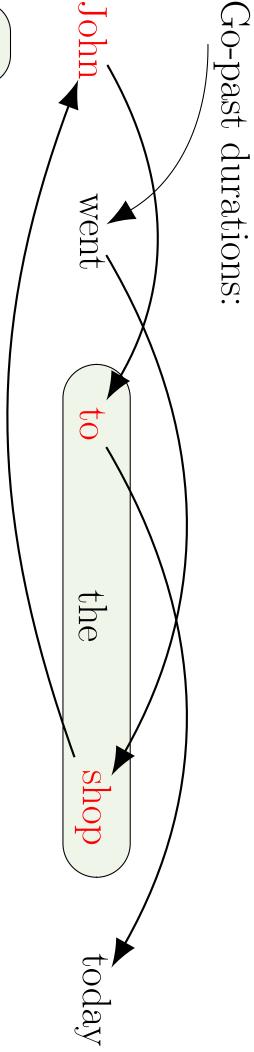
- Increased embedding depth
- Reduced embedding depth

Predictions

h operation and reading	n each opera	Table 1: Predicted correlation between eacl
positive negative	positive	Embedding Difference
positive negative	positive	Dynamic Recruitment
positive		Hier. Sequential Prediction
positive	positive positive	ACT-R
positive	positive	Dependency Locality Theory positive positive
F-L+	F+L-	Theory

Eye Tracking

- Slower reading difficulty
- How much can be processed up to a given point?



Iraining

Fixation in go-past

duration

-past

region

Parser and Lexicon: WSJ 02-21

950,028 Accuracy words Berkeley Parser 39,832 sentences [van Schijndel et al.,

N-grams:

rams: Brown, WSJ 02-21, BNC, Dundee 87,302,312 words • 5,052,904 sentences Smoothed with modified Kneser-Ney

Evaluation

Dundee Eye-tracking Corpus [Kennedy et al., 2003]

- 10 subjects 154,168 go-past durations 2,388 sentences

Omit:

- Unknown words \mathcal{O} times in WSJ)
- First and last of each line
- Fixations after long saccades (> 4 words)
- Mixed Effects Model:

Word length

- Cum.
- Surprisal surprisal
- Prev/Next fixated? Unigram and bigram Cum. entropy reduction
- Joint interactions

Sentence

position

- Factors from prev. region
- Subject/Item random intercepts Length of region alculate

nonlation w	otrica anmon		3.	
umuiative n	dimulative metrics summed over regional	d over re	510N	
Factor	Operation	Coeff	t-score	p-value

0.12	-1.55	-0.0031	Cue Awaited	F+L+
$5.23 \cdot 10^{-8}$	-5.44	-0.0154	Integrate	F-L+
$2.22 \cdot 10^{-14}$	7.10	0.0224	Initiate	F+L-
0.55	0.60	0.0026	Cue Active	F-L-
p-value	t-score	Coeff	Operation	Factor

Conclusions

Support for dynamic recruitment:

- Initiation difficulty due to load causes recruitment
- Returns processing to average difficulty
- Int egration facilitation due to reduced load causes release
- Returns processing to average difficulty

fects, which perhaps causes infrequency of embeddings Initiation inhibition suggests difficulty beyond frequency