

## Abstract

The phenomenon of *local coherences* has recently generated significant interest in the sentence processing literature. Tabor, Galantucci, & Richardson (2004) showed that participants are slower to read object-modifying reduced relative clauses (RCs) when the RC verb is part-of-speech ambiguous (tossed) than when it is umambiguous (thrown). This result is problematic for any fully incremental framework because the main-verb interpretation is incompatible with the global context and should thus be ignored by the processor. An unresolved issue with this result, however, is the extent to which such effects occur in natural reading. We test the extent to which such effects occur in natural reading, using eye-tracking data from free reading of newspaper text. Our results support both of the major theories of local coherences, and add ecological validity to the study of local coherences.

## 1. Background and Motivation

## background

Tabor, Galantucci, & Richardson (2004) varied verb ambiguity and relative clause reduction in object-modifying relative clauses, using sentences such as in (1):

- (1) a. The coach smiled at the player tossed a frisbee . . .
  - b. The coach smiled at the player who was tossed a frisbee.
  - c. The coach smiled at the player thrown a frisbee . . .
  - d. The coach smiled at the player who was thrown a frisbee.
- found an interaction such that [+amb,+red] was super-additively hard
- problematic result for any fully incremental framework because the main-verb interpretation is incompatible with the global syntactic context

## theories

### *Tabor et al. (2004):*

- difficulty arises because the 'locally coherent' parse of 'the player tossed a frisbee' competes with top-down syntactic expectations

*Gibson (2006)*:

- difficulty arises from word part-of-speech (POS) disambiguation
- difficulty inversely proportional to context-independent probability of tag given word multiplied by the smoothed syntactic expectations

## question

- both theories make broad claims about the parser from one example involving a very rare construction, a reduced relative of a passivized ditransitive
- how often do such effects arise in the reading of natural text?

## plan

- build a model of the reading times in a corpus of naturalistic eye-tracking data
- test factors representing each of the two major theories

References:

Kennedy, A., & Pynte, J. (2005). Parafoveal-on-foveal effects in normal reading. *Vision Research*, 45, 153–168. Tabor, W., Galantucci, B., & Richardson, D. (2004). Effects of merely local syntactic coherence on sentence processing. Journal of Memory and *Language*, 50, 355–370.

# Local Coherences in the Wild: An Eye-tracking Corpus Study Klinton Bicknell<sup>1</sup>, Vera Demberg<sup>2</sup>, and Roger Levy<sup>1</sup>

<sup>1</sup>University of California, San Diego <sup>2</sup>University of Edinburgh

	[+amb, +red]
• •	[+amb, -red]
	[-amb, +red]
•••	[-amb, -red]

## 2. Corpus, Model, and Control Factors

## corpus

- used the Dundee corpus (Kennedy & Pynte, 2005) of eye-movement data from 10 participants reading 51,000 words of *The Independent* (a British newspaper)

- to get syntactic category information, we parsed the corpus with the Charniak parser (Charniak, 2000)

## model

each of our factors was tested in a linear mixed-effect model (Baayen, Davidson, & Bates, In Press) of the first pass times on each word, with 11 fixed effect control factors and participant as a random effect, as in Demberg & Keller (To Appear)

## control factors

Demberg & Keller's control factors included:

linguistic properties:

- word length and frequency
- bigram probability
- position in the sentence
- lexicalized and unlexicalized syntactic
- surprisal [ =  $-\log P(W_i \mid context)$ ]

## 3. Factor One: $P(t_i|w_i)$

## motivation

Gibson (2006) predicts a word w to be read slower with a particular POS tag t as the context-independent probability  $P(t_i | W_i)$  decreases

## estimation

- for each word-tag pair in the Dundee cor estimated  $P(t_i | W_i)$  from a Charniak-parse of the British National Corpus (BNC)
- two versions of this factor:
- P<sub>m</sub>: maximum likelihood estimate (MLE)
- P<sub>s</sub>: MLE smoothed with a Bayesian prior, which is the type-averaged  $P(t_i | w)$  [see box]

## results

tested factor on all words in Dundee corpus that were found in the BNC

	coefficient estimates
MLE (P <sub>m</sub> ):	-0.8 [ <i>t</i> (199394) = -0.8;
Smoothed (P <sub>s</sub> ):	-2.3 [ <i>t</i> (199394) = -2.2;

## further analysis

- performed a natural spline regression on  $P_s(t_i | W_i)$  with 11 equally spaced knots (see graph; confidence intervals are bootstrapped)
- spline regression suggests that the significant effect is being driven by difference in the 0.0-0.2 range of probability



eye movement properties:

- landing position
- launch distance
- whether the previous word was fixated

Bayesian prior set to be:

 $\sum P_m(t_i|w)$  $P_{pr}(t_i) = \frac{w}{\|w\|}$ 

where *w* is a word *type*.

## smoothing

 $P_{s}(t_{i}|w_{i}) = \frac{c(t_{i}, w_{i}) + \beta [P_{pr}(t_{i})]}{c(t_{i}, w_{i}) + \beta [P_{pr}(t_{i})]}$  $c(w_i) + \beta$ where  $\beta$  is set to minimize Dundee corpus perplexity

p > .05] p < .05]

no effect (MLE too noisy?) small effect

### 4. Factor Two: $P(t_i|w_{i-1},w_i)$

### motivation

- conflicts with a top-down parse

## estimation

- two versions:  $P_m = MLE$

## results

	С
MLE (P <sub>M</sub> ):	-2
Smoothed (P <sub>c</sub> ):	-1

## further analysis

- performed a natural spline regression on  $P(t_i | W_{i-1}, W_i)$  as before
- spline regression shows evidence of an effect for all values of the factor

## 5. Caveat

## correlations

our factors were somewhat correlated with control factors:

- residualizing the covariates

## 6. Conclusion

## summary

## next step

test a larger subset of cases predicted by Tabor et al. to be difficult by using a modified parser to predict conflict between top-down and bottom-up parses



	word frequency	bigram probability
$P_{s}(t_{i} W_{i})$ :	r = .21	r = .19
$P_{s}(t_{i}   W_{i-1}, W_{i}):$	r = .28	r = .40

- because these control factors were explicitly included in the model, we have some confidence that our effect is 'above and beyond' them

0.2

08

0.0

- nevertheless, we plan to address this in the future by:

- constructing more estimates of these and other measures of conflict between top-down and bottom-up probabilities

- evidence for both Tabor et al. and Gibson's theories of local coherence effects - adds ecological validity to the study of local coherences

Baayen, R.H., Davidson, D.J., & Bates, D.M. (In Press). Mixed-effects modeling with crossed random effects for subjects and items. Journal of Memory and Language.

Charniak, E. (2000). A maximum-entropy-inspired parser. In *Proceedings of the 1<sup>st</sup> Conference of the North American Chapter of the Association for* Computational Linguistics, (pp. 132–139), Seattle, WA. Demberg, V., & Keller, F. (To Appear) Cognition.

Gibson, E. (2006). The interaction of top-down and bottom-up statistics in the resolution of syntactic category ambiguity. Journal of Memory and *Language*, 54, 363–388.