

# A large-scale deconvolutional study of predictability and frequency effects in naturalistic reading

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## Question

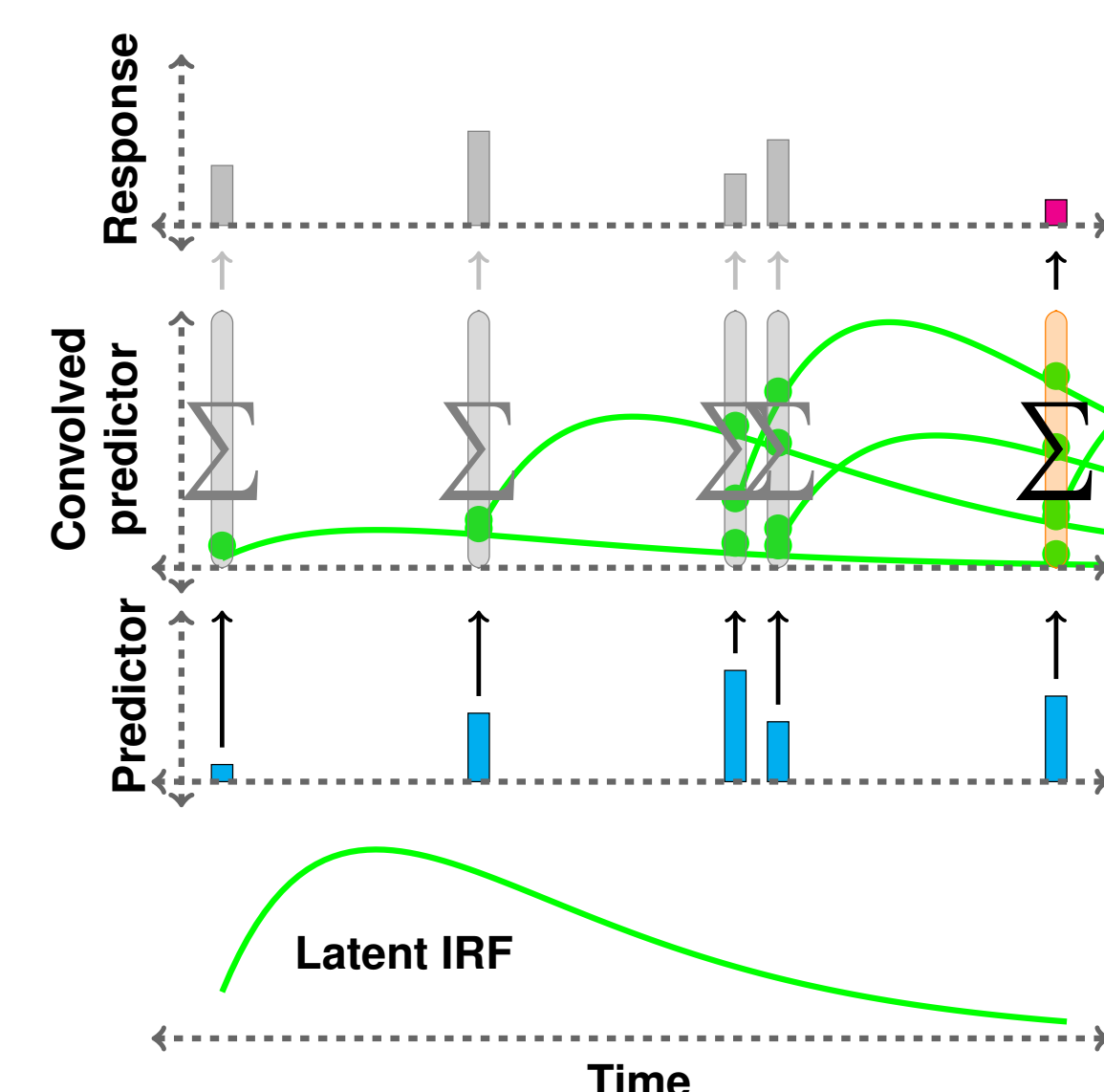
Are there distinct mechanisms for (1) predicting a word vs. (2) retrieving a word from the mental lexicon?

## Theoretical Background

- **Yes:**
  - Lexical retrieval cost depends on the strength of a word's representation in memory [18, 3, 10]
  - Retrieval cost is context-independent
  - **Prediction:** Separable effects of *predictability* and *frequency*
- **No:**
  - Comprehenders incur costs for incrementally reallocating resources among possible interpretations [15, 14, 16]
  - No context-independent lexical retrieval mechanism.
  - Frequency effects are subsumed into the probability model
  - **Prediction:** No separable effects of *predictability* and *frequency*

## Experimental Background

- Lots of experimental evidence for **Yes:**
  - Additive effects of corpus frequency and cloze predictability [17, 1, 8, 22]; see [21] for review.
- However:
  - Constructed stimuli may introduce task artifacts [4, 11, 2]
  - Cloze poorly differentiates low-probability words [20]
- Can be addressed by naturalistic stimuli with statistical probability estimates. However:
  - *Frequency* and *predictability* are naturally collinear [4]
  - Temporal diffusion may confound word-by-word modeling [5, 19]
- **This study:**
  - Naturalistic data address ecological validity
  - Large-scale data address collinear variables of interest
  - Deconvolutional modeling addresses diffusion of effects



Continuous-time deconvolution

## Experimental Design

- Evaluation on 3 large corpora containing over 1M events total:
  - **Natural Stories**, self-paced reading [7]
  - **Dundee**, eye-tracking [13]
  - **UCL**, eye-tracking [6]
- Deconvolutional time series regression [19]
- **Controls:** Sentence position, document position, word rate, word length, saccade length, whether the previous word was fixated
- **Predictors of interest:** unigram log probability, 5-gram surprisal
- Probabilities computed by KenLM models [12] trained on the Gigaword 3 corpus [9]
- **Response:** Log-ms (go-past for eye-tracking)
- By-subject random intercepts, slopes, and impulse response parameters

## Results

Comparison	p-value
5-gram only vs. baseline	<b>0.0001***</b>
Unigram only vs. baseline	<b>0.0001***</b>
5-gram + Unigram vs. Unigram-only	<b>0.0001***</b>
5-gram + Unigram vs. 5-gram-only	0.1440

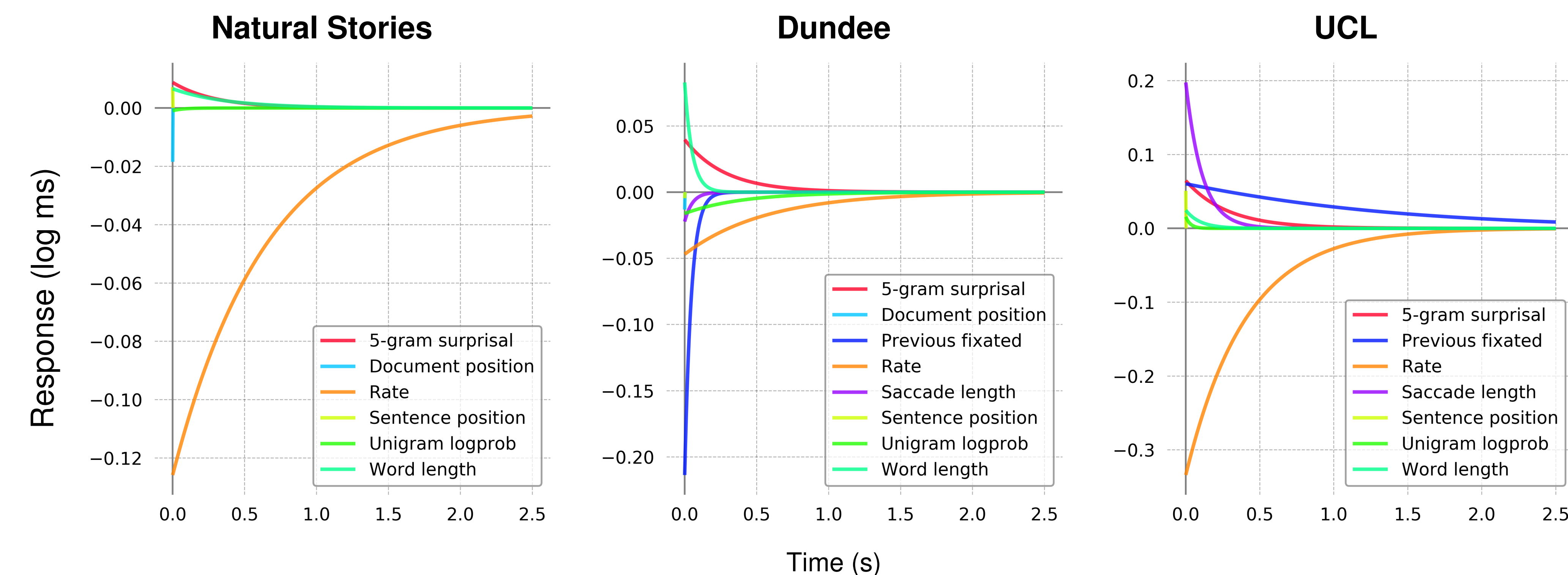
Pooled permutation testing results on out-of-sample data.

**Main result:** Significant effect of *frequency* over *predictability* but not vice versa, consistent with **No**.

Corpus	Effect estimate (log-ms)							
	SentPos	Trial	Rate	WordLen	SacLen	PrevFix	Unigram	5-gram
Natural Stories	0.0098	-0.0216	-0.3069	—	—	0.0158	<b>-0.0018</b>	<b>0.0174</b>
Dundee	-0.0085	-0.0052	-0.0277	0.0068	-0.0021	-0.0178	<b>-0.0067</b>	<b>0.0117</b>
UCL	0.0524		-0.1330	0.0023	0.0221	0.0778	<b>0.0005</b>	<b>0.0184</b>

Effect estimates (integrals of impulse response functions)

*Predictability* effects are larger magnitude than *frequency* effects.



Estimated impulse response functions for each predictor by corpus

## Conclusion

Results support **No**: no evidence of separable effects of *frequency* and *predictability*.

Finding is at odds with constructed experiments. Possible explanations:

- Frequency effects may exist in naturalistic reading but are too small to be detected.
- Constructed stimuli may introduce confounds:
  - Atypical word distributions
  - Lack of context
  - Suspension of normal communicative function of language
  - Comprehension → problem solving
- Cloze estimates may be too coarse, allowing frequency predictors to capture residual variance due to predictability

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