

Noun/Verb Entropy: an MEG Study of Word-level Syntactic Category Ambiguity

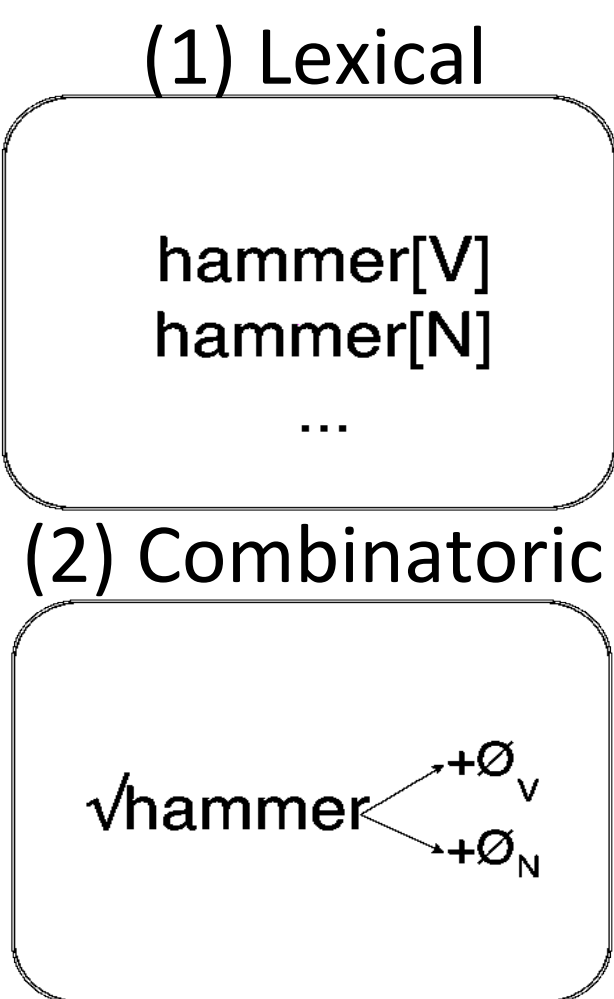
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Introduction

- How are grammatical categories such as *noun* and *verb* computed during word recognition? At least two options (for review, see: Vigliocco et al. 2011):

- (1) **Lexical**: category is a feature of the representation of each word, with distinct representations for each category
 - (2) **Combinatoric**: lexical category is assigned to a category-neutral root via affixation¹
- Lexical category as a feature of distinct words predicts category ambiguity correlations with 300ms MTL (middle temporal lobe) activity (see: e.g., meaning entropy effects for distinct meanings in Simon et al. 2012)
 - Lexical category as a product of combinatoric processes predicts earlier (before 300ms) LATL (left anterior temporal lobe) activity correlated with category ambiguity (see: verb subcategorization frame effect in Linzen et al. 2013 and cf. Bemis & Pykkänen 2013)
 - Does the brain response to category ambiguity of null-inflected words support (1) or (2)?



¹For linguistic support of (2), see: Barner & Bale (2002), Chomsky (to appear), Marantz (1997)

Stimuli Variables

Lexical Variables

- Derivational entropy**;
 $\left. \begin{matrix} \text{hammer...} \end{matrix} \right\} \begin{matrix} -able & 0.6 \\ -er & 0.2 \\ -ize & 0.2 \end{matrix}$
probability values for example only
- Number of senses (polysemes)**;
 - Number of distinct meanings (homographs)**;
 - bank (institution or river)
 - bat (animal or sports object)
 - row (line or paddle)

Combinatoric Variables

- Noun/Verb entropy**;
 $\left. \begin{matrix} \text{hammer...} \end{matrix} \right\} \begin{matrix} -\emptyset V & 0.33 \\ -\emptyset N & 0.67 \end{matrix}$
- Inflectional entropy**;
 $\left. \begin{matrix} \text{hammer...} \end{matrix} \right\} \begin{matrix} -s & 0.3 \\ -ing & 0.3 \\ -ed & 0.4 \end{matrix}$

Example N/V entropy
words >0.67:

veil, seat, risk, peel,
joke, heat, blur

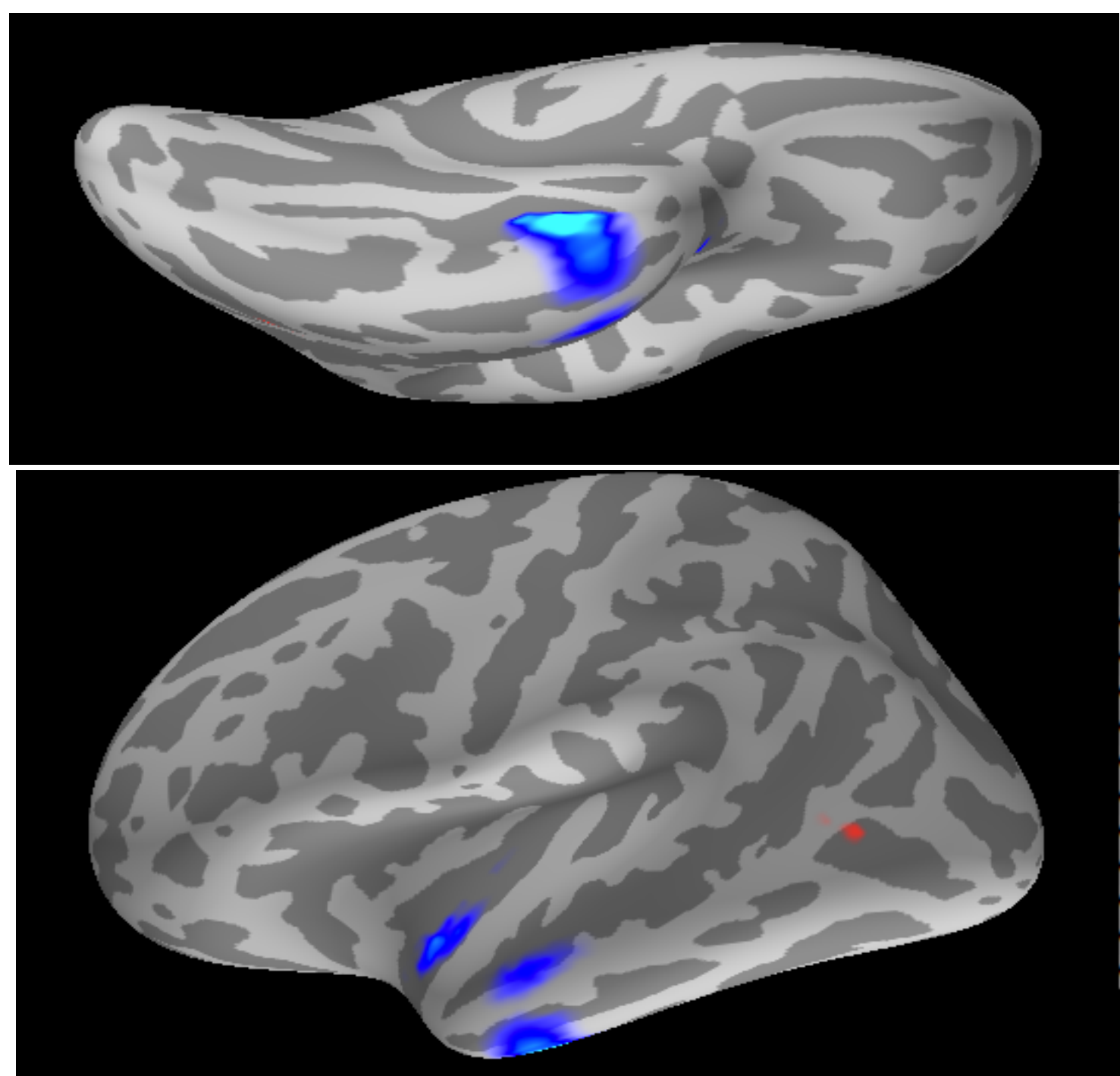
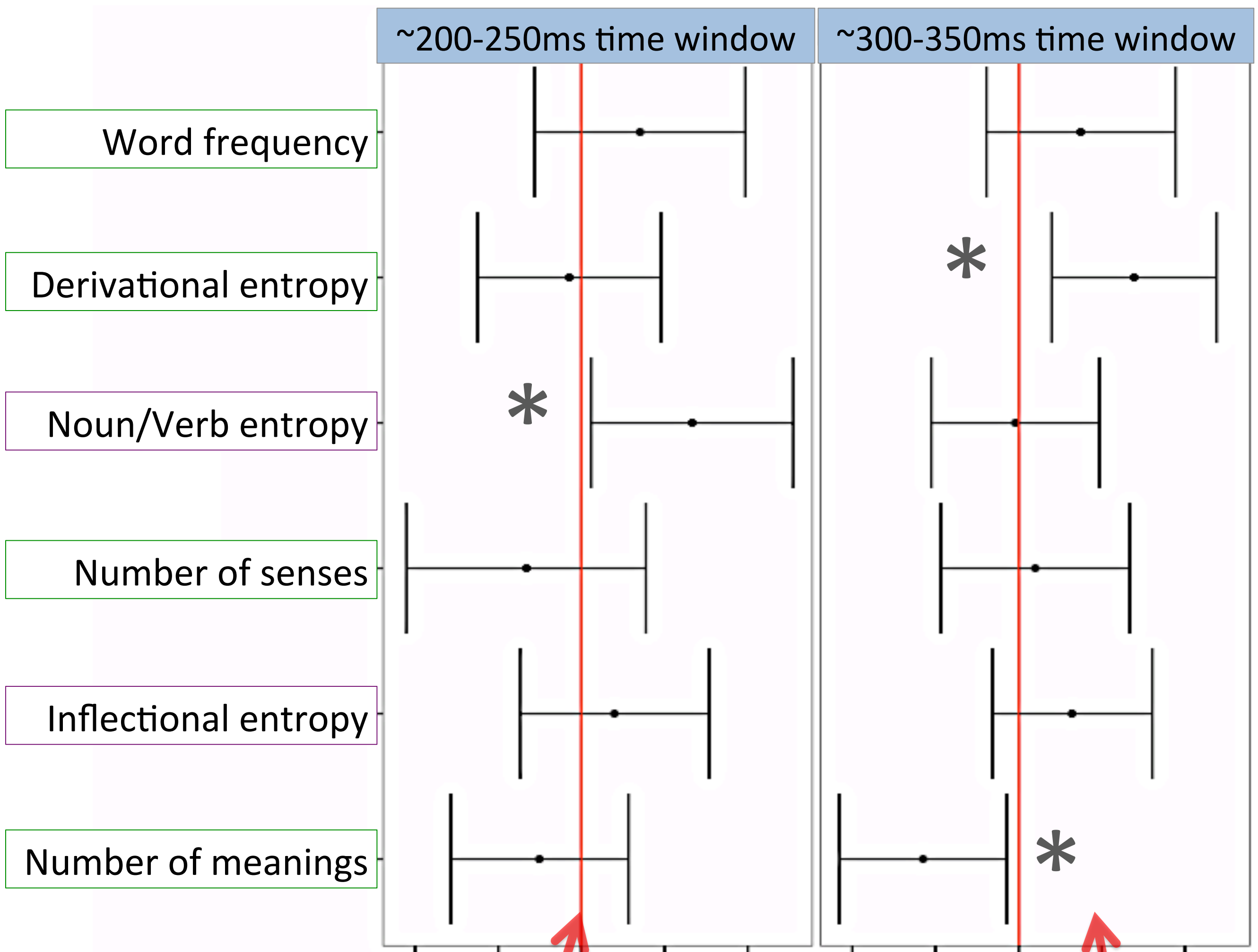
Noun/Verb entropy is given by:

$$H_{\text{noun/verb}} = -p_{\text{verb}} \log_2 p_{\text{verb}} - p_{\text{noun}} \log_2 p_{\text{noun}}$$

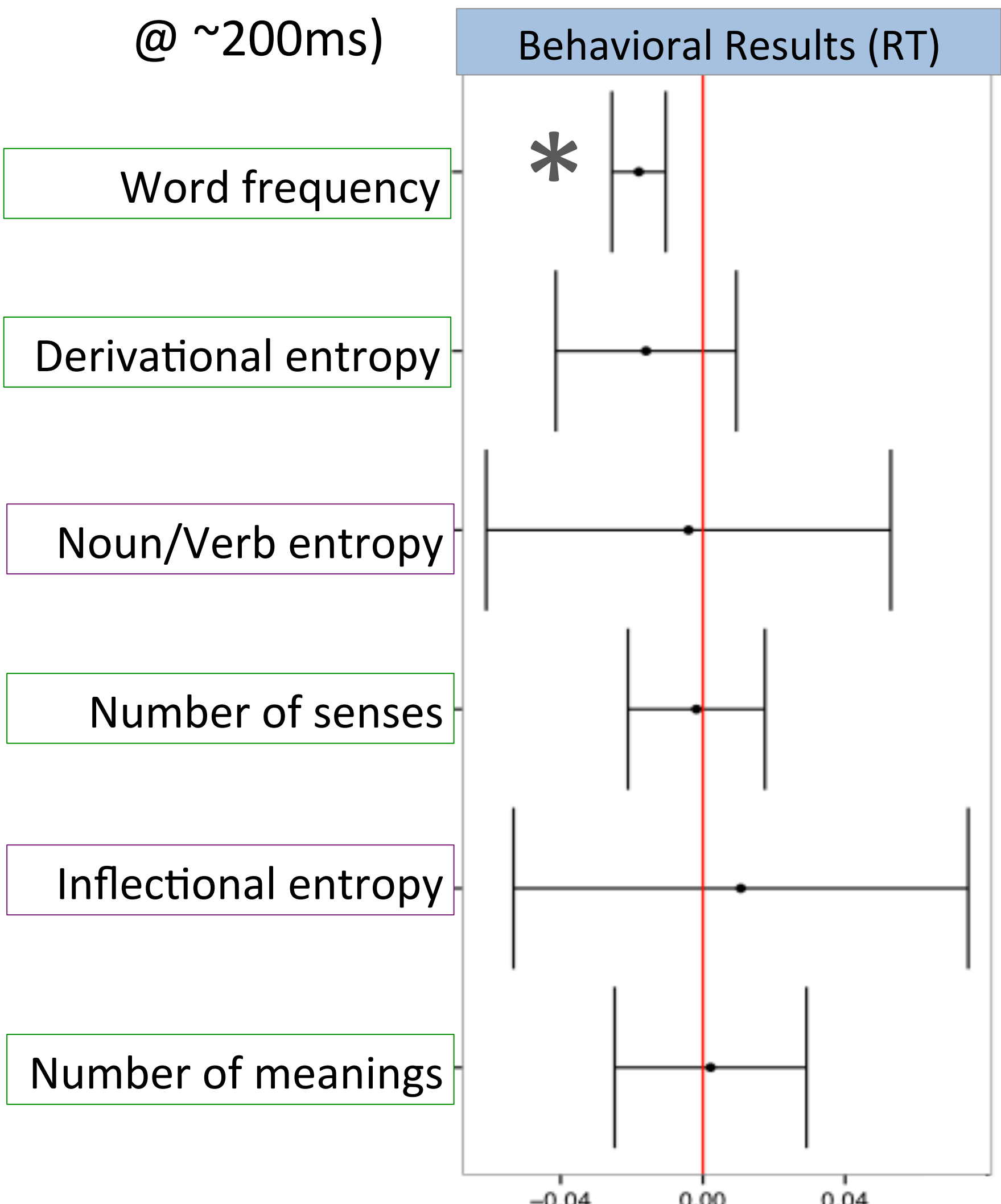
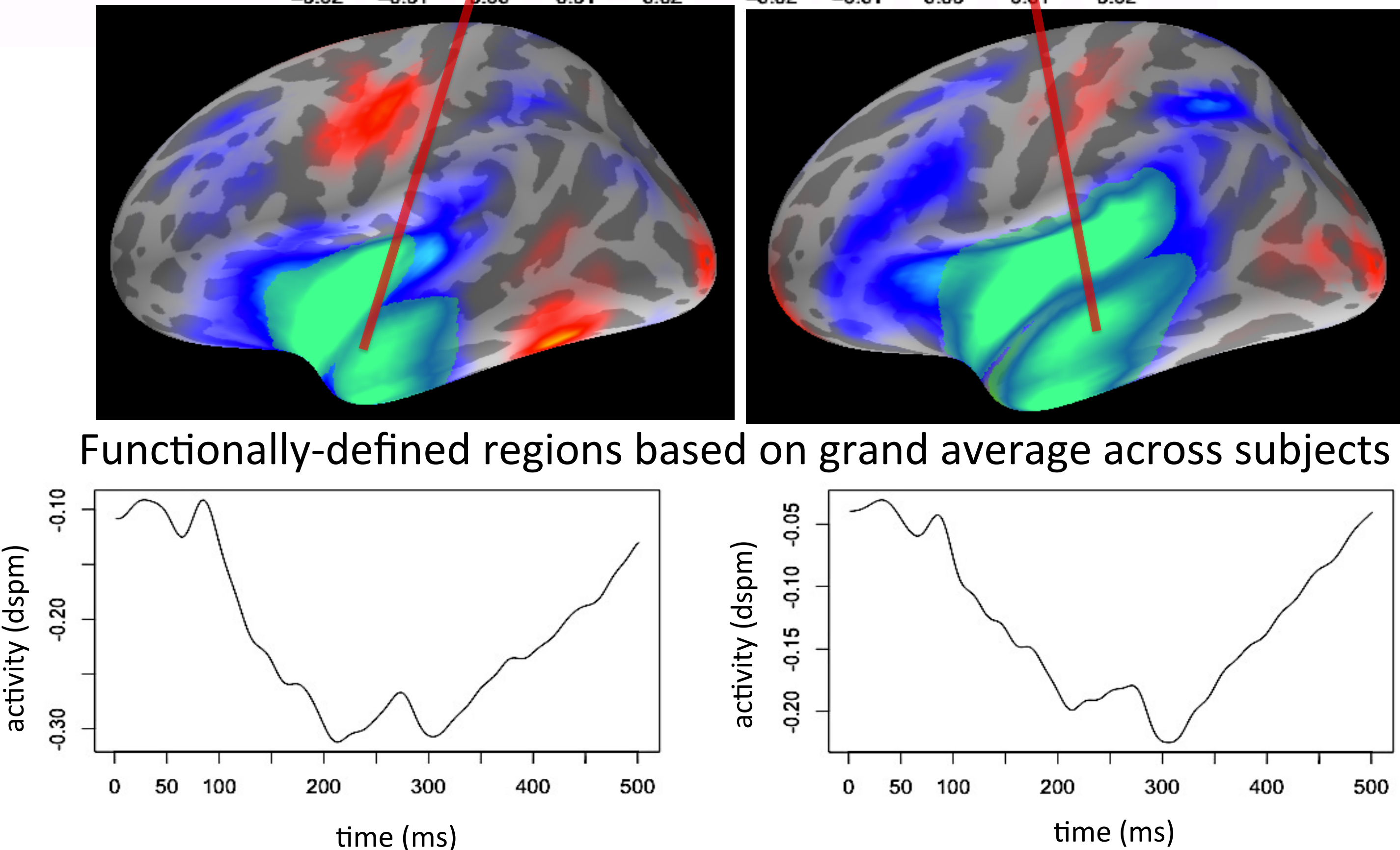
Materials & Methods

- Visual lexical decision experiment with concurrent MEG recording
- 12 right-handed native English speakers
- 208 sensor array
- Source solutions calculated with MNE (Gramfort et al. 2014)
- 313 words
- Effects of predictors were assessed using continuous linear mixed effects regression on single trial source activity

Results



Correlations between neural activity and N/V entropy (uncorrected t-maps @ ~200ms)



Conclusions

- No significant correlations were observed with number of senses or with inflectional entropy; however, stimuli were not selected optimally for observing the effects of these variables
- Number of meanings and derivational entropy correlate with activity in a broader temporal region and a later time window (~300-350ms), replicating experiments that associate these variables with lexical access
- As predicted by the combinatoric hypothesis, **N/V entropy correlates with activity in the LATL (left anterior temporal lobe) within the ~200-250ms time window**, parallel to the effects of subcategorization frame entropy for verbs when presented in isolation (Linzen et al. 2013)

References: Barner, D., & Bale, A. (2002). *Lingua*, 112, 771-791; Bemis, D., & Pykkänen, L. (2013). *PLoS ONE*, 8(9), e73949; Gramfort, A., Luessi, D., Larson, E., Engemann, D., Strohmeier, D., Brodbeck, C., Parkkonen, L., & Hämäläinen, M. (2014). *Neuroimage* 86, 446-460; Linzen, T., Marantz, A., & Pykkänen, L. (2013). *The Mental Lexicon* 8:2, 117-139; Marantz, A. (1997). *University of Pennsylvania Working Papers in Linguistics* 4.2: 201-225; Simon, D., Lewis, G., & Marantz, A. (2012). *Language and Cognitive Processes*, 27:2, 275-287; Vigliocco, G., Vinson, D., Druks, J., Barber, H., & Cappa, S. (2011). *Neuroscience & Biobehavioral Reviews* 35, 407-426. This work is supported by the NYU Abu Dhabi Research Council under grant G1001 from the NYUAD Institute, New York University Abu Dhabi.