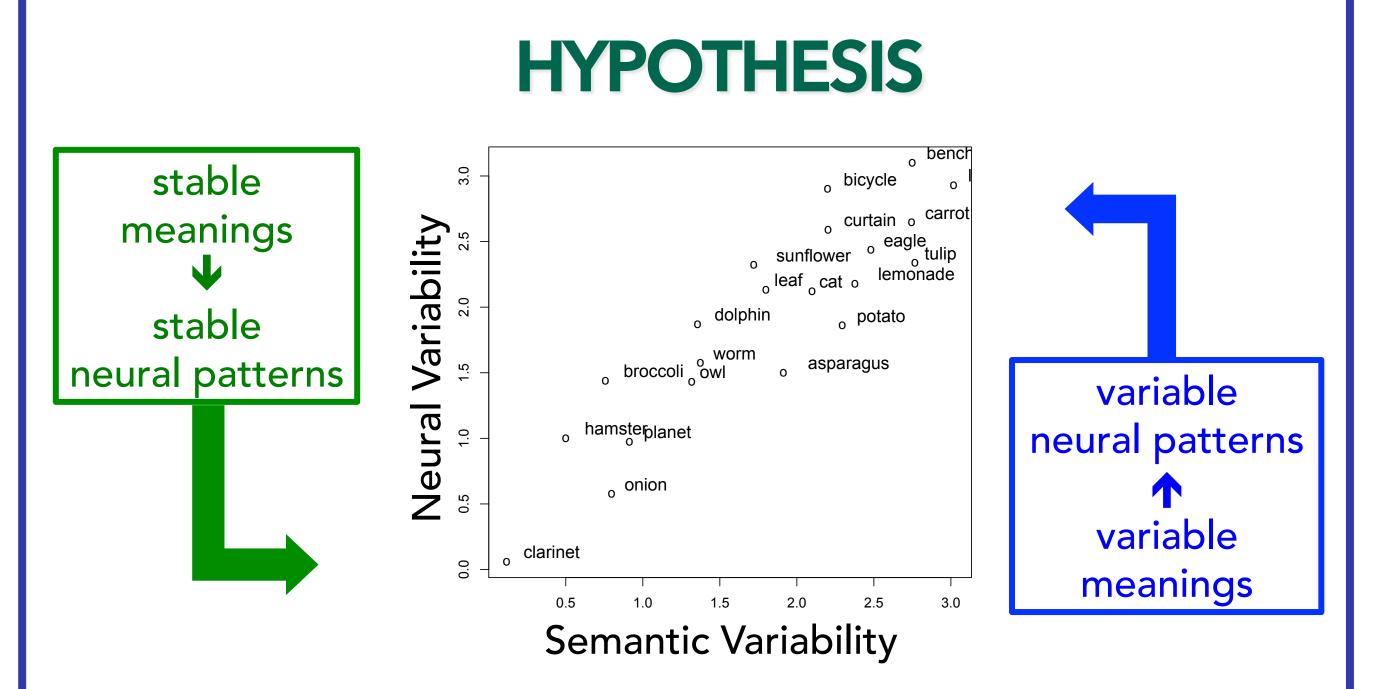


#### **CONTEXT-DEPENDENT THEORY OF CONCEPT REPRESENTATION**

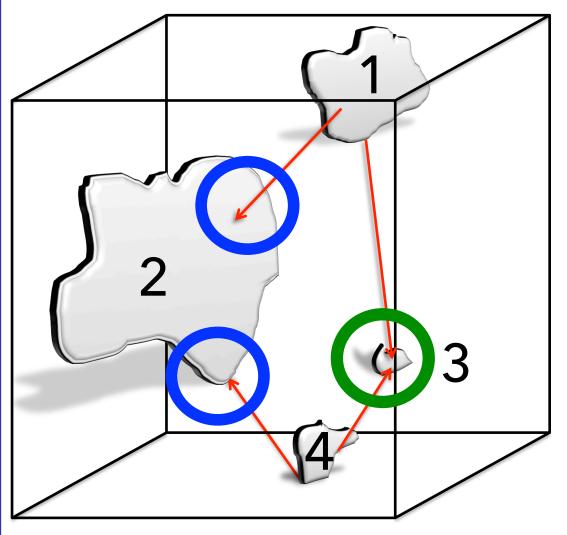
- A Meaning is generated by the dynamic interaction between a concept and the context in which it is accessed.
- ♦ Concepts are not represented as context-invariant, static entities retrieved in isolation.
- $\diamond$  Neuroscientists often treat these representations as fixed. - common practices to reduce "noise" in signal: averaging across stimulus presentations; limiting analyses to voxels with the most stable activation profiles

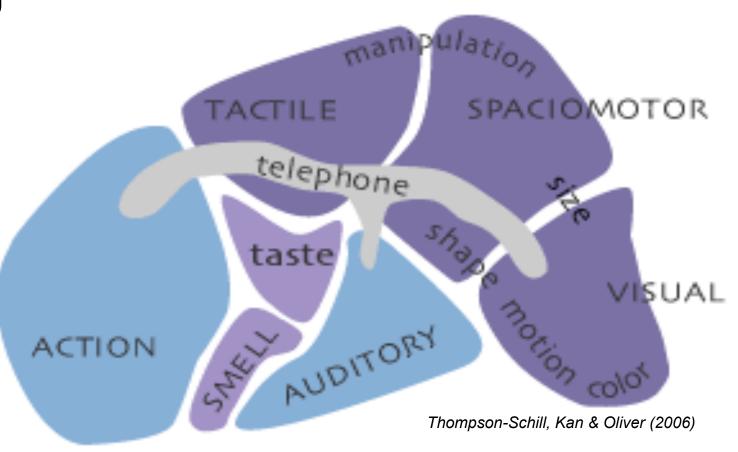
#### Objective

- $\diamond$  Compare neural patterns elicited by conceptual processing of the same stimulus item as it appears in different contexts
- ♦ Relate within-item, cross-context neural variability to measures of semantic/contextual variability



- ♦ Corresponding neural variability: physical manifestation of concept-context coupling
- ♦ Semantic features are neurally distributed & dynamically activated depending upon current task/context





- Projected into high-dimensional semantic space: a concept's meanings in its various uses
- ♦ Traverse from one concept to another
- $\diamond$  Concept #2 has more diverse meanings than Concept #3
- ♦ The 2 instantiations of Concept #2 are more variable than the 2 instantiations of Concept #3

# Semantic Variability Predicts Neural Variability of Object Concepts

## Elizabeth Musz & Sharon L. Thompson-Schill **University of Pennsylvania**

### **QUANTIFYING SEMANTIC VARIABILITY**

♦ Concept: word appearances in large linguistic corpora ♦ Context: paragraph of text in which the word appears ♦ How many contexts does each concept appear in, and how similar are these contexts to one another?

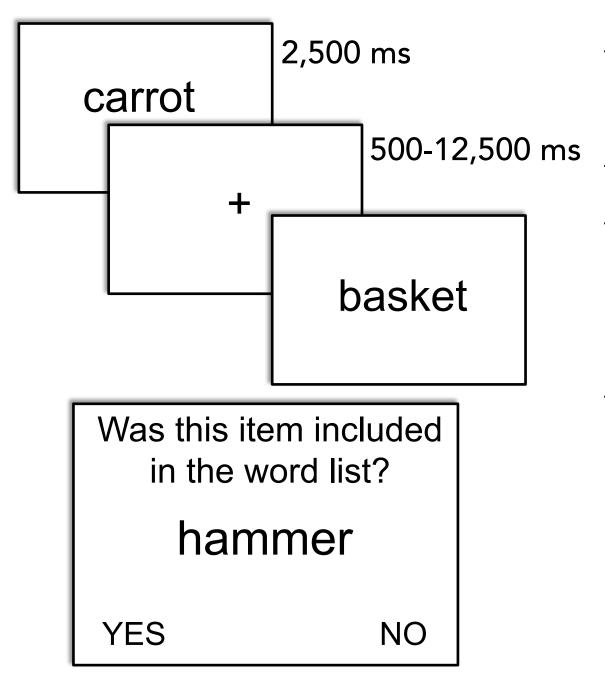
 $\diamond$  SemVar: a composite score for each concept, computed using PCA on results from topic modeling<sup>3</sup>, LSA<sup>2</sup>, and context frequency counts<sup>1</sup>  $\diamond$  a measure of diversity amongst a concept's contexts

#### **CREATING VARIABLE CONTEXTS**

#### Stimuli

- $\diamond$  160 single-sense, concrete nouns: 30 "target" & 130 "filler" words 15 polysemous & homonymous words ω Words assigned to 9 *unique, randomly* ordered lists, each with: • 10 targets
  - 15 fillers
  - 5 "poly/homs"
- ♦ Each target & poly/hom word appears in 3 different lists
- ♦ Unique and unrepeated fillers added to lists, to increase list variability

#### Procedure

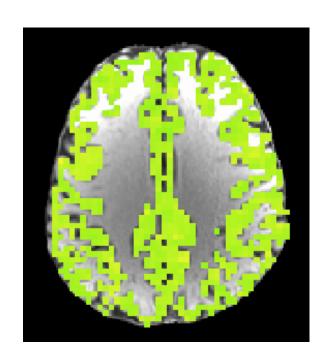


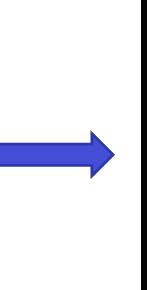
- 1 scan per list, each 4 minutes long
- Sequential word presentations  $\diamond$ ♦ Task during scanning: memory encoding
  - Purposefully left open-ended, to avoid constraining subjects' semantic interpretations
- ♦ Task after each run: recognition memory tests
  - Probes: 5 foils & 5 fillers
  - Memory for target words never tested between lists

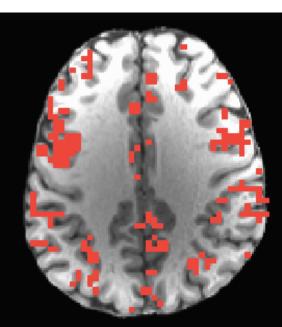
#### WHOLE-BRAIN VOXEL SELECTION

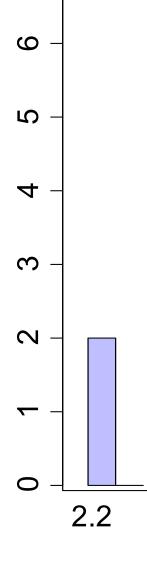
 $\diamond$  Gray matter voxels ranked by test statistic, those most responsive to both: (1) words vs. fixation and (2) differences across word presentations  $\diamond$  Measured neural patterns in 12 voxel sets of varying sizes: top 25-10,000 voxels  $\diamond$  Contiguity constraint: each voxel must share a face with 1+ other included voxels

Voxel test statistics in gray matter voxels

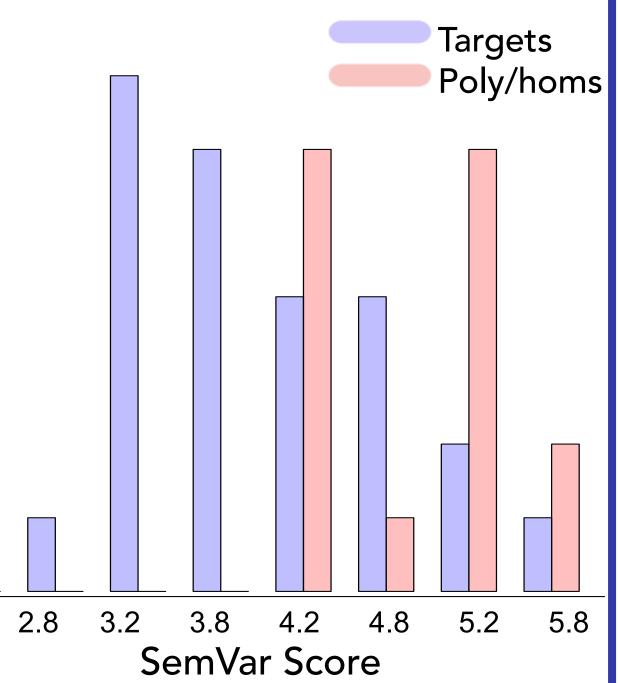






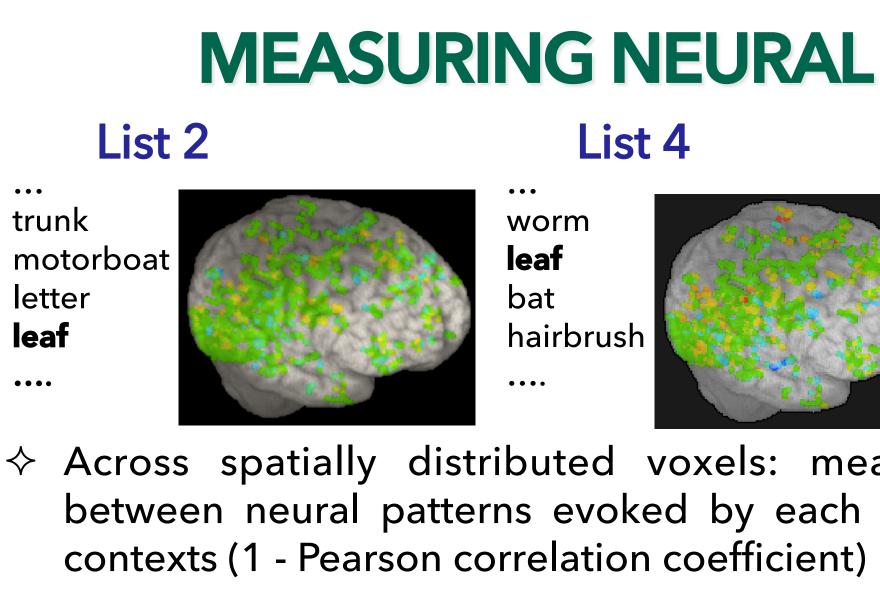


#### Histogram of SemVar Values



Subjects (n=19) completed 9 fMRI scans

Mask of voxels with top 2,000 test statistics



#### **Semantic-Neural Correlation**

- ♦ Positive correlation between target words' SemVar score and corresponding neural variability, *t*(18)= 3.1, *p*= .006
- Correlations significantly positive across subjects, when patterns measured in sets of 250-2,000 voxels.

#### **Effects of Lexical-Semantic Ambiguity**

- ♦ Polysemous and homonymous words: 2+ different meanings or senses share the name same (e.g., chicken meat & chicken bird)
- $\diamond$  these words should exhibit especially variable patterns, since they denote multiple concrete meanings.
- ♦ Polyhoms elicit less neural similarity than target words: t(19) = -2.2, p = .04
- $\diamond$  Robust at voxel set sizes: 25-750
- $\diamond$
- $\diamond$
- $\diamond$ semantic representation

#### REFERENCES

1. Brysbaert, M., & New, B. (2009). Moving beyond Kučera and Francis: A critical evaluation of current word frequency norms and the introduction of a new and improved word frequency measure for American English. Behavior Research Methods, 41, 977-990.

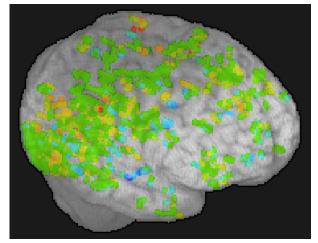
2. Hoffman, P., Lambon Ralph, M. A., & Rogers, T. T. (2012). Semantic diversity: A measure of semantic ambiguity based on variability in the contextual usage of words. *Behavior Research Methods*, 45, 718-730. 3. Pereira, F., Botvinick, M., & Detre, G. (2013). Using Wikipedia to learn semantic feature representations of concrete concepts in neuroimaging experiments. Artificial Intelligence, 194, 240-252.



### **MEASURING NEURAL VARIABILITY**

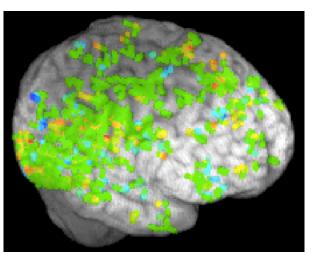
List 4

leaf hairbrush

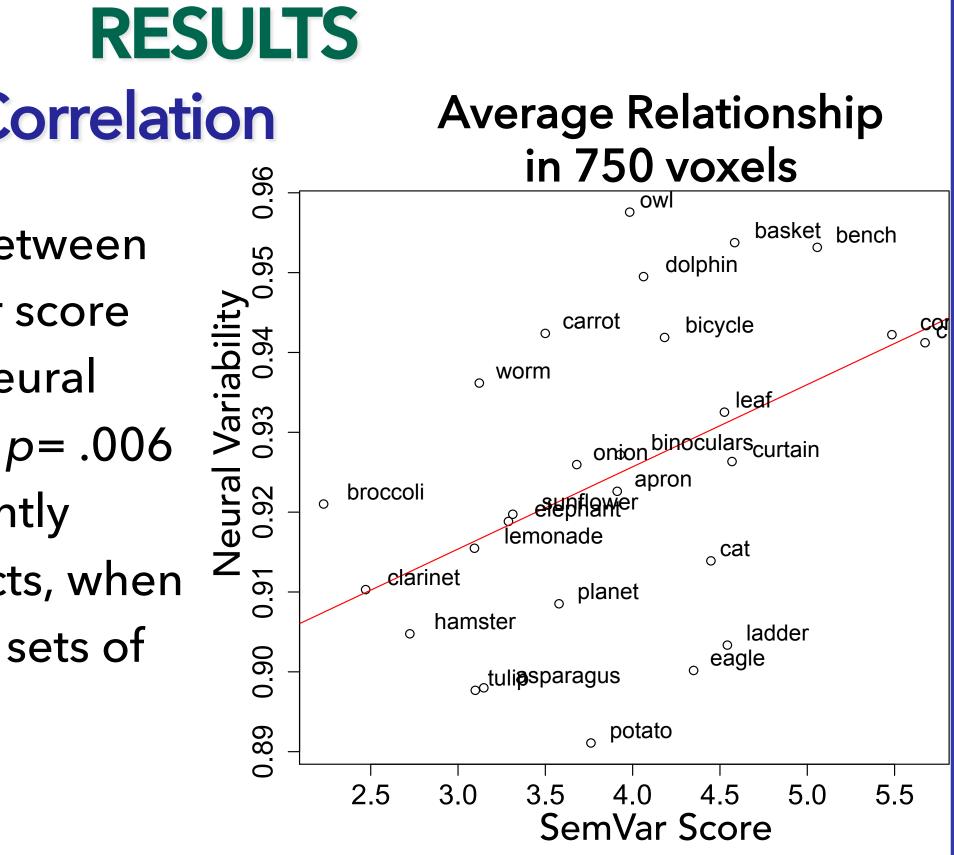


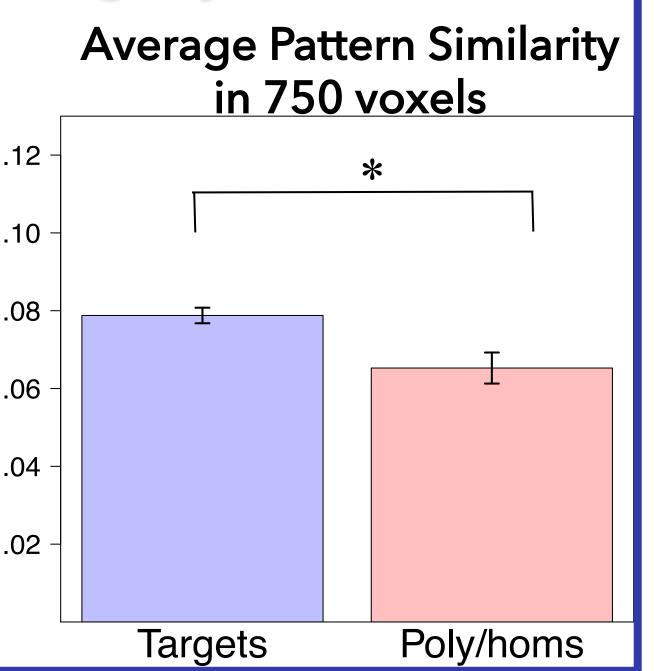
chicken zucchini leaf asparagus

List 9



 $\diamond$  Across spatially distributed voxels: measured average dissimilarity between neural patterns evoked by each concept in its three different





#### DISCUSSION

Neural activity varied across repeated stimulus presentations, and this variation was reliably predicted by measures of semantic variability.

Supports a flexible, distributed theory of semantic memory organization, in which a concept's meaning varies continuously as a function of its context.

Within-stimulus "noise" can reflect context-modulated variation in a concept's

