

## INTRODUCTION

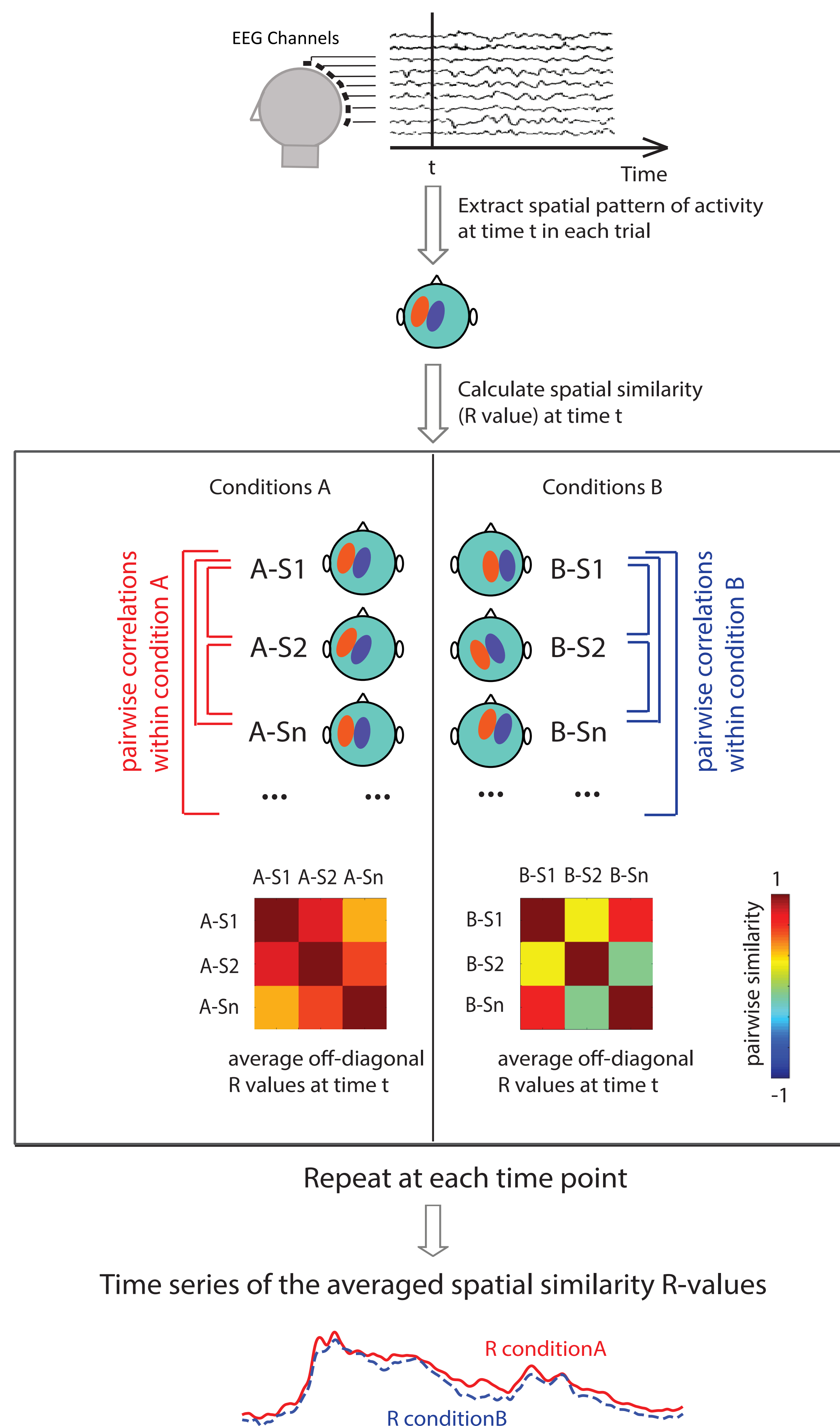
- Negativity bias: When attending to emotional valence, we devote more attentional resources to evaluating negatively-valenced stimuli than positive or neutral stimuli (Baumeister et al., 2001; Ito et al., 1998).
- ERPs: Negativity bias manifests as a larger late positivity evoked by negative stimuli, from 500ms after stimulus onset (Citron, 2012; Delaney-Busch et al., 2016).
- Question: does the neural negativity bias reflect deeper (re)-analysis of semantic representations during this late of evaluative processing?
- A large body of work suggests that the semantic representation of a word can be inferred “by the company it keeps” (Firth, 1957).
- Therefore, we used computational models, which build semantic representations (high-dimensional vector spaces) through a statistical analysis of the contexts in which words occur, in combination with EEG and Representational Similarity Analysis (RSA) to probe the nature of the neural negativity bias.

## METHOD

- Semantic similarity measures of pairs of words: word2vec (Mikolov, 2013) and Latent Semantic Analysis (LSA; Landauer et al., 2007).
  - Random sample of 4637 words from the Warriner, Kuperman, & Brysbaert (2013) corpus: varied in emotional valence (negative, neutral, positive).
  - 468 words that were used in an EEG study: varied in emotional valence (negative, neutral, positive) and were matched on arousal and lexical properties (word length, word frequency, and concreteness).
- EEG recorded as 22 healthy adults read and judged the valence of the matched set of 467 words.
  - Analysis of ERPs: late positivity (500-800 ms; previously reported by Delaney-Busch et al., 2016).
  - Spatial similarity analysis
    - Cluster-based permutation ANOVA to test similarity of patterns of neural activity elicited by emotionally valenced words.

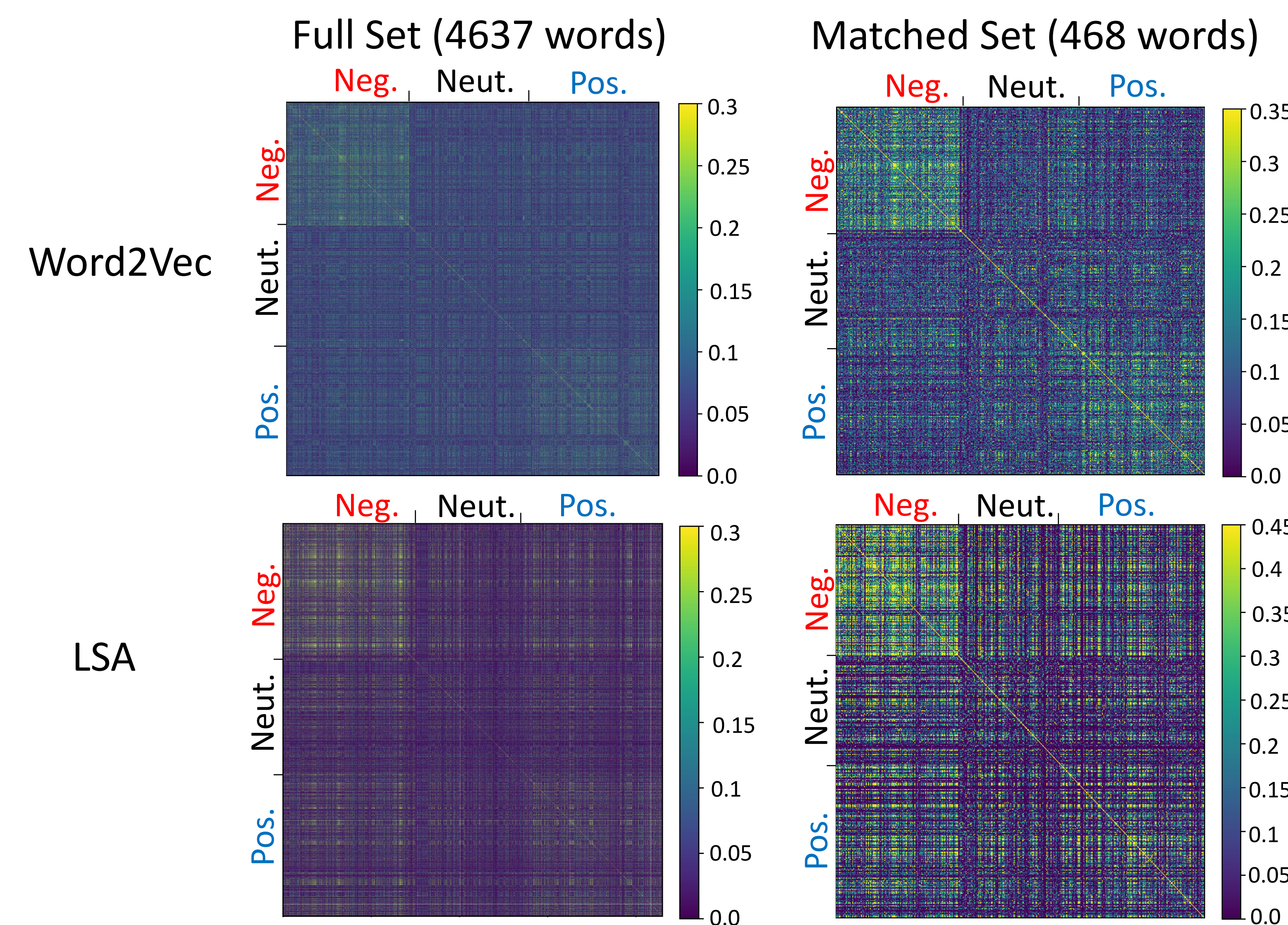
### Spatial Similarity Analysis Pipeline

(see also Wang et al., 2020)

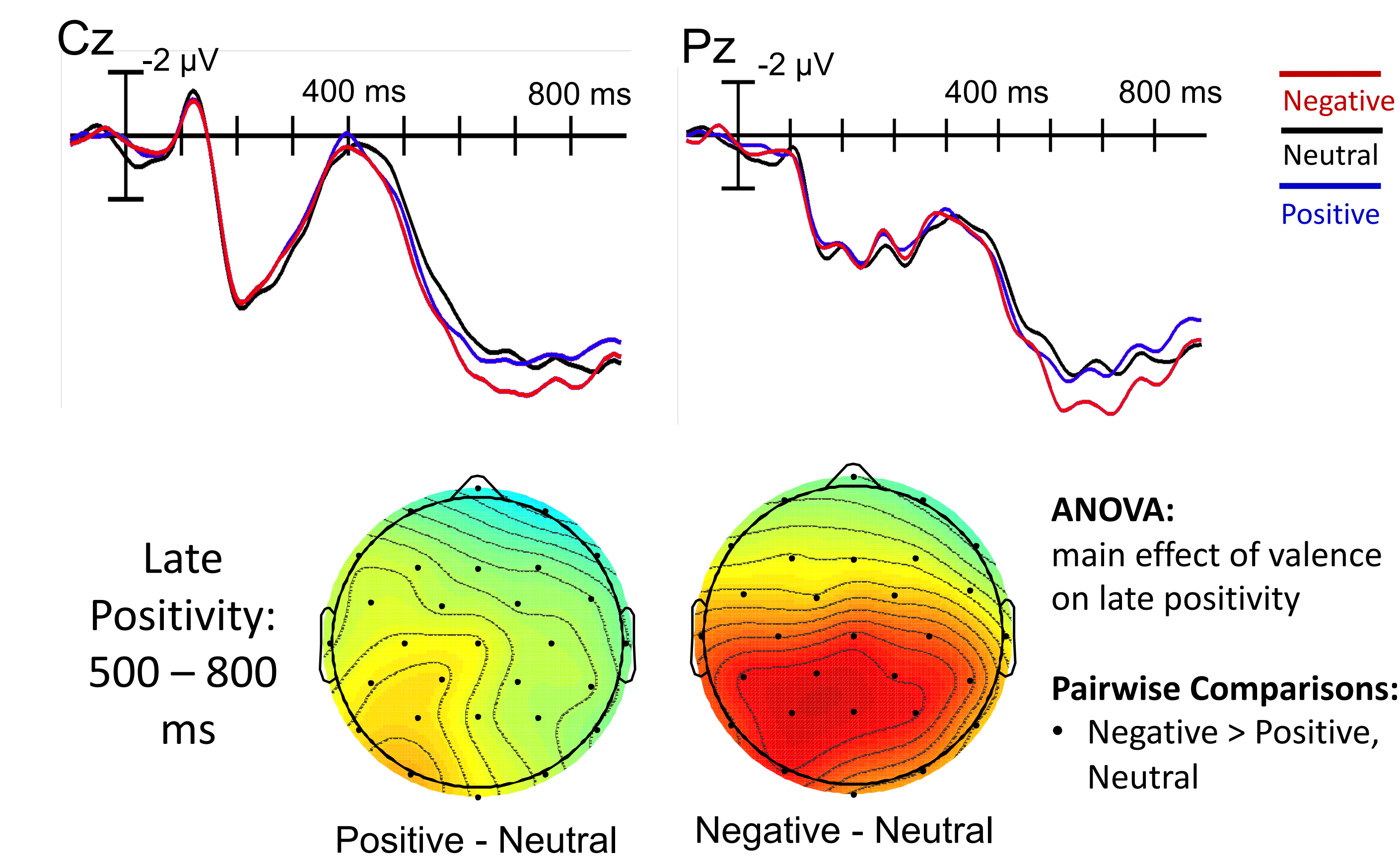


## RESULTS

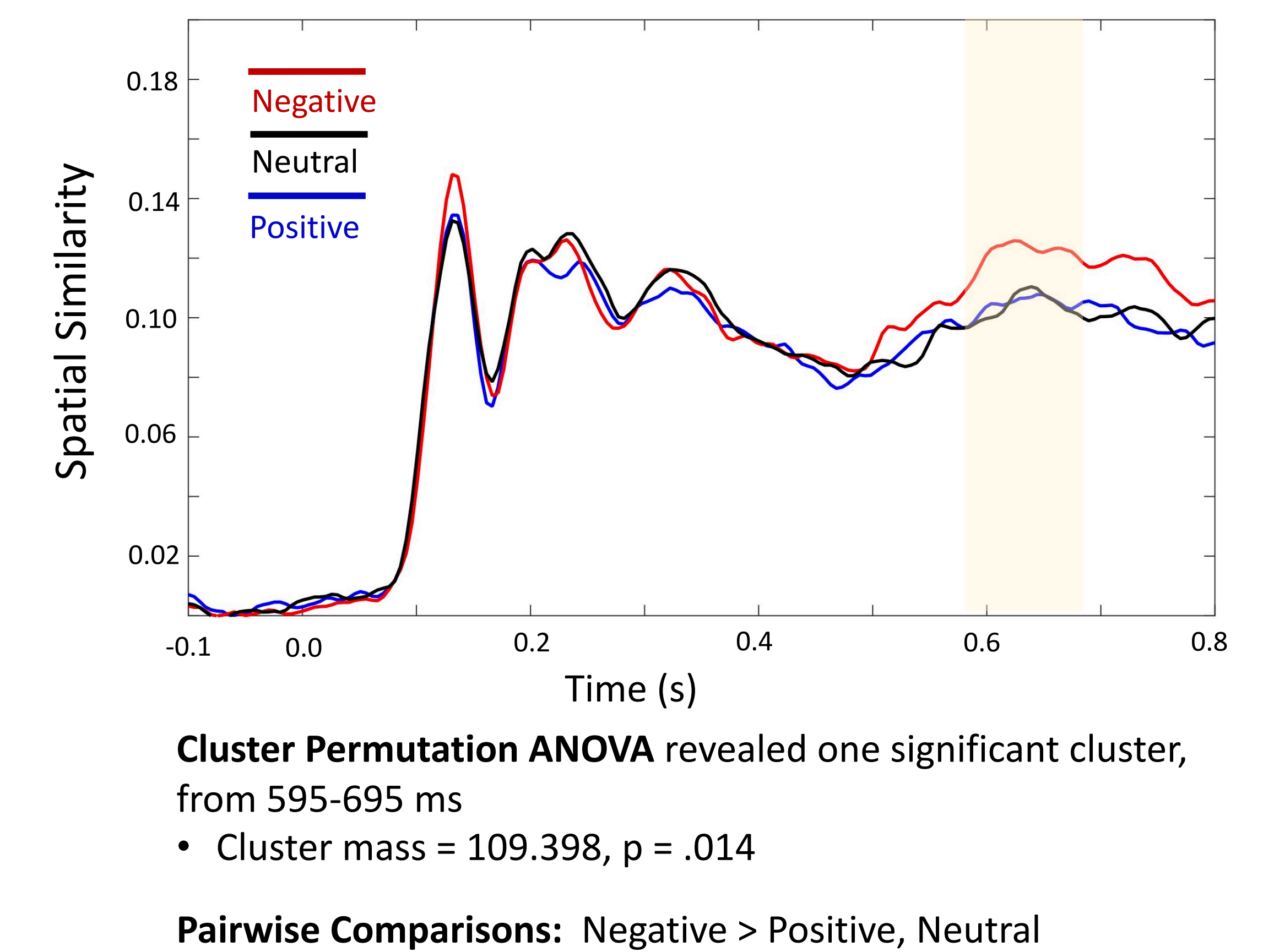
### Pairwise Semantic Similarity



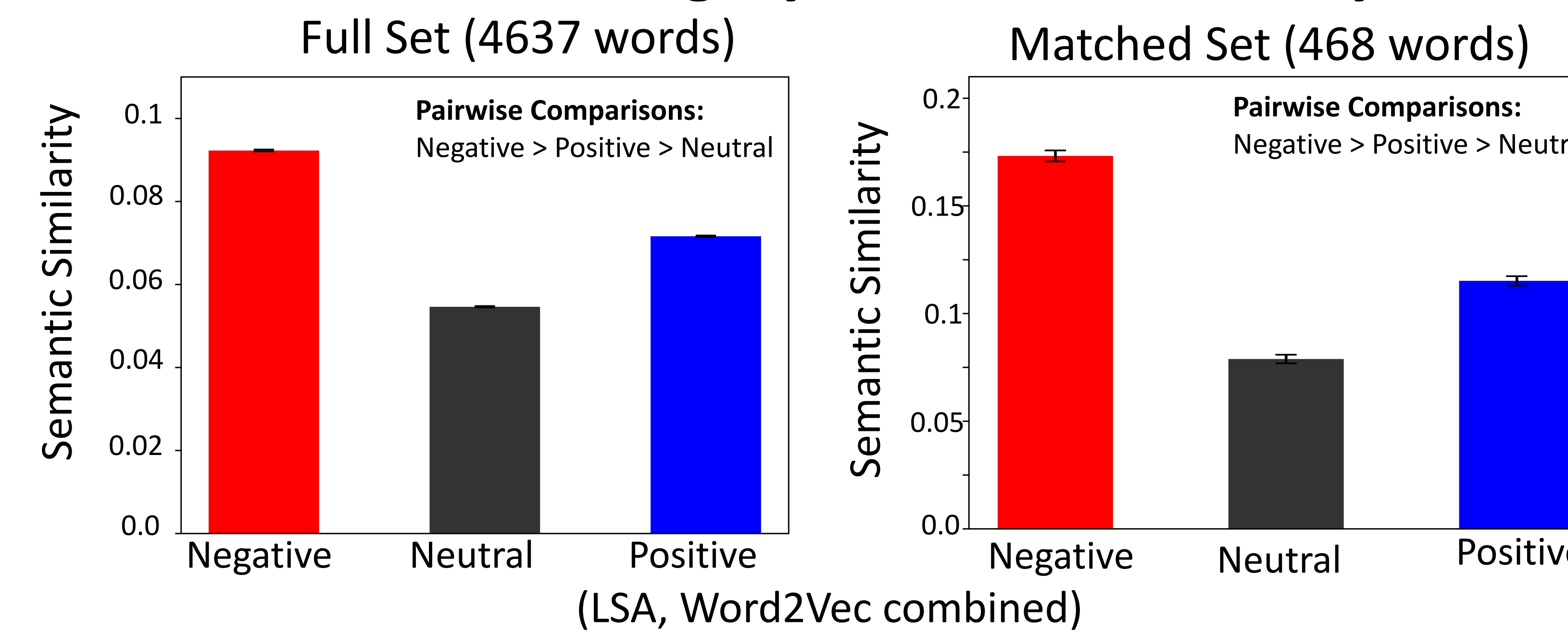
### ERP Analysis



### EEG Spatial Similarity Analysis



### Within-Category Semantic Similarity



## CONCLUSIONS

- Computational model-based similarity measures showed greater semantic similarity amongst negative words than amongst positive words, and greater similarity among positive words than neutral words.
- ERPs confirmed the negativity bias: Negative words evoked a larger late positivity between 500-800ms than neutral or positive words (see also Delaney-Busch et al., 2016).
- Spatial similarity analysis: within the late positivity time window, there was greater similarity between patterns of neural activity produced by negative words than patterns of activity produced by positive or neutral words.
- Together, these data suggest that the prolonged neural processing associated with negatively-valenced stimuli may reflect a re-evaluation of their underlying semantic representations.
- Future: Trial-by-trial spatial similarity analysis to confirm that greater neural similarity in the late positivity time window is linked to greater semantic similarity on a trial-by-trial basis.

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### ACKNOWLEDGMENTS

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