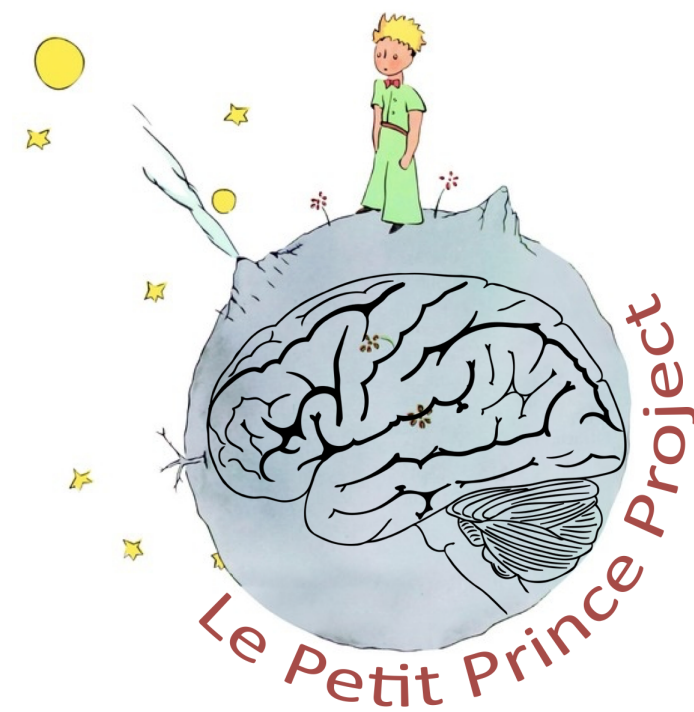


# Modeling Conventionalization and Predictability in Multiword Expressions at the brain level



Murielle Popa-Fabre<sup>1</sup>

Shohini Bhattachali<sup>2</sup>

John Hale<sup>3</sup>

Christophe Pallier<sup>4</sup>

<sup>1</sup>CNRS Université Paris Diderot–Paris 7

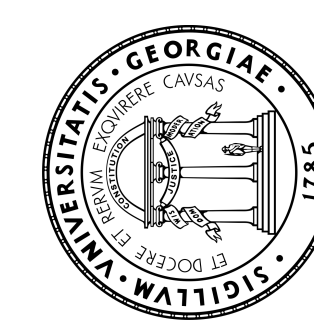
<sup>2</sup>University of Maryland

<sup>3</sup>University of Georgia

<sup>4</sup>CEA Inserm Cognitive Neuroimaging Unit



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## Introduction

Multiword Expressions (MWEs) demonstrate finer-grained distinctions such as degrees of conventionalization and predictability.

Examples: *break the ice*, *boa constrictor*, *safe and sound*, *see to it*, *in spite of*

These gradients could be reflected during cognitive processing and are captured using computational metrics, Association Measures.

## Question

Are these computational measures and their cognitive instantiations are discernible at the cerebral level during sentence processing?

## Data Collection

Participants (n=51) were college-aged, right-handed, native English speakers.

Listened to a spoken recitation of *The Little Prince* for 1 hour and 38 minutes across nine separate sections; 15,388 words in total.

Comprehension was confirmed through multiple-choice questions at the end of each section (90% accuracy, SD = 3.7%).

## Multiword Expressions

742 MWEs were identified in the dataset through a transition-based MWE analyzer (Al Saied et al., 2017) trained on Children's Book Test dataset (Hill et al., 2015).

Association Measures calculated using corpus frequency counts from COCA:

① Pointwise Mutual Information (PMI):

$$PMI = \log_2 \frac{c(w_n^1)}{E(w_n^1)} \quad (1)$$

② Dice's Coefficient:

$$Dice = \frac{n \times c(w_n^1)}{\sum_{i=1}^n c(w_i)} \quad (2)$$

## Group-level Results

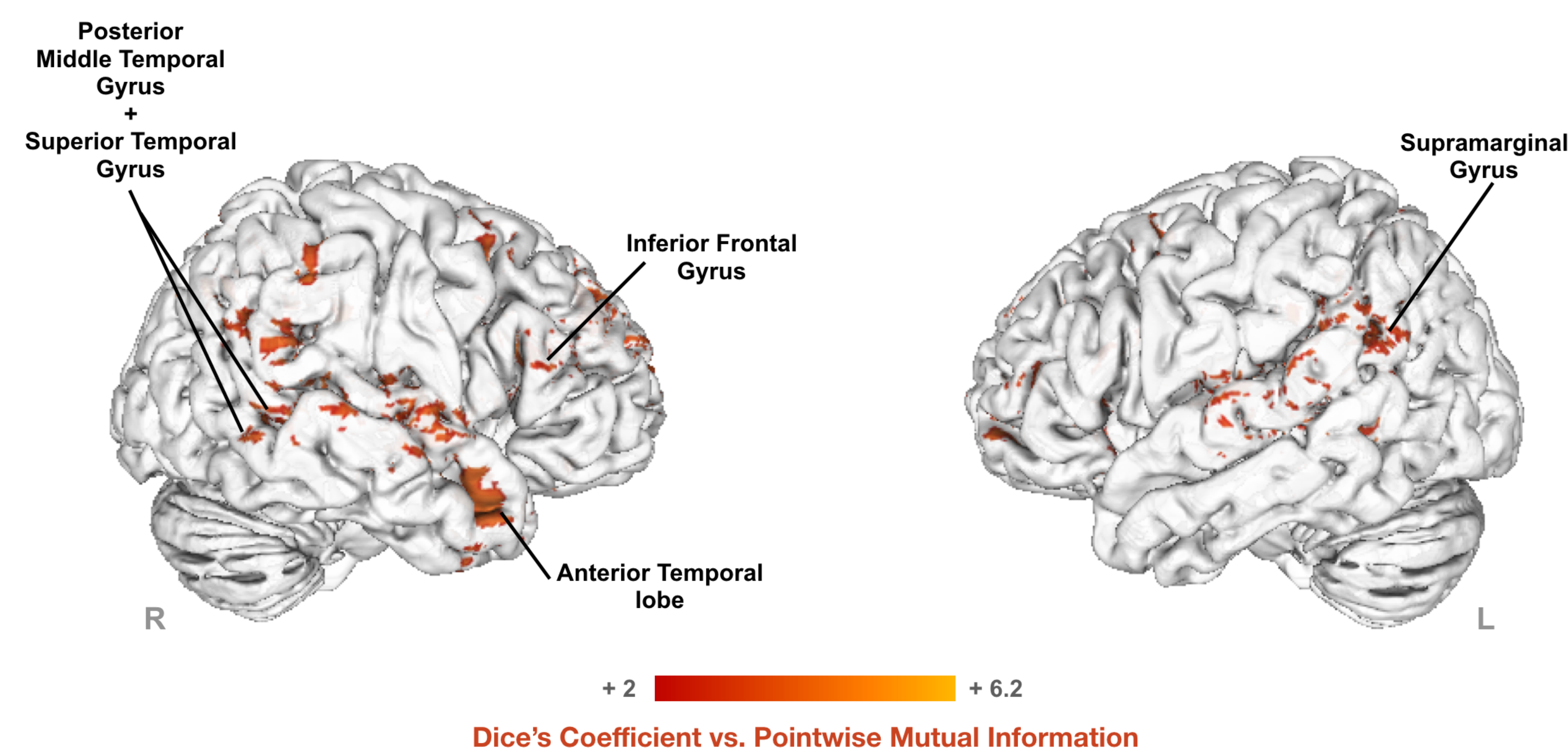


Figure 1: Z-map showing regions having a significant effect for Dice's coefficient versus Pointwise Mutual Information after Bonferroni correction with  $p < 0.05$

Regions for Dice > PMI	Cluster size MNI Coordinates z-scores				
	(in voxels)	x	y	z	
R Superior Temporal Gyrus (BA 38)	47	48	10	-26	5.80
R Middle Temporal Gyrus	84	54	-18	-10	6.09
R Middle Temporal Gyrus (BA 22)	98	48	-36	2	5.85
R Superior Temporal Gyrus (BA 22)	70	48	-12	2	5.83
R Middle Temporal Gyrus (BA 22)	16	58	-46	2	5.14
L Superior Temporal Gyrus	13	-62	-18	6	5.64
R Superior Frontal Gyrus	10	20	56	12	5.53
R Inferior Frontal Gyrus (BA 45)	10	48	20	14	5.64
L Supramarginal Gyrus	22	-56	-56	22	5.37
R Inferior Parietal Lobule/ Superior Temporal Gyrus (BA 40)	10	62	-46	22	5.44
R Inferior Parietal Lobule/ Superior Temporal Gyrus (BA 40)	16	54	-46	22	5.45
R Superior Frontal Gyrus	35	20	42	34	5.69
R Cingulate Gyrus	17	2	-34	34	5.85
R Precuneus	22	32	-72	36	5.76
L Inferior Parietal Lobule	12	-34	-58	46	5.17

Table 1: Significant clusters for Dice's Coefficient versus Pointwise Mutual Information after Bonferroni correction with  $p < 0.05$

## fMRI Analysis

Preprocessing was carried out with AFNI version 16 and ME-ICA v3.2 (Kundu et al., 2011).

GLM Analysis:

- MWE predictors convolved with SPM12's canonical HRF, regressed against observed BOLD signal during passive story listening.
- To account for sentence-level compositional processes, a regressor formalizing syntactic structure building included (Hale, 2014).
- Includes four regressors of non-interest: word offset, frequency, pitch, intensity

$r^2$  Model Comparison: For every subject, we compute how much the inclusion of each variable of interest (i.e. Dice and PMI) increases the cross-validated  $r^2$ .

## Conclusion

- Dice's Coefficient, formalizing the degree of predictability, is a better predictor of cerebral activation for processing MWEs.
- Indicates that Dice is a more cognitively plausible computational metric.
- Suggests that a bimodal distribution of gradient is more suitable for processing MWEs

## Acknowledgements

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