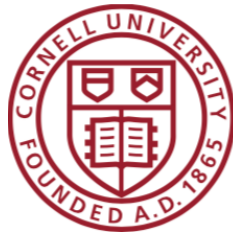


Diathesis Alternations and Selectional Restrictions in Sentence Processing: A fMRI Study

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55th Meeting of the Chicago Linguistic Society (CLS 55)
May 17, 2019

Introduction

- ▶ Human study of natural language comprehension
- ▶ Brain areas that correspond to different aspects of sentence processing, as exemplified through verbs
- ▶ Present neuroimaging study using fMRI

Background

- ▶ Verbal argument structure
- ▶ Diathesis alternations

Background

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- ▶ Diathesis alternations

break

1. Fred broke the window

AGENT THEME

2. Fred broke the window with a rock

AGENT THEME INSTRUMENT

3. The rock broke the window.

INSTRUMENT THEME

4. The window broke.

THEME

5. The window was broken by Fred.

THEME AGENT

Background

- ▶ Verbal argument structure
 - ▶ Diathesis alternations
 - ▶ Selectional restrictions
 - ▶ *pour, sing* vs. *make, give*

General Approach

- ▶ Incorporating computational linguistics:
 - ▶ Operationalize cognitive hypotheses
 - ▶ Often based on incremental, expectation-based theories of sentence comprehension (Hale, 2001; Levy, 2008)
 - ▶ Statistical/probabilistic language models (along with information-theoretic complexity measures)
 - ▶ Estimate word-by-word comprehension difficulty during real time language processing,

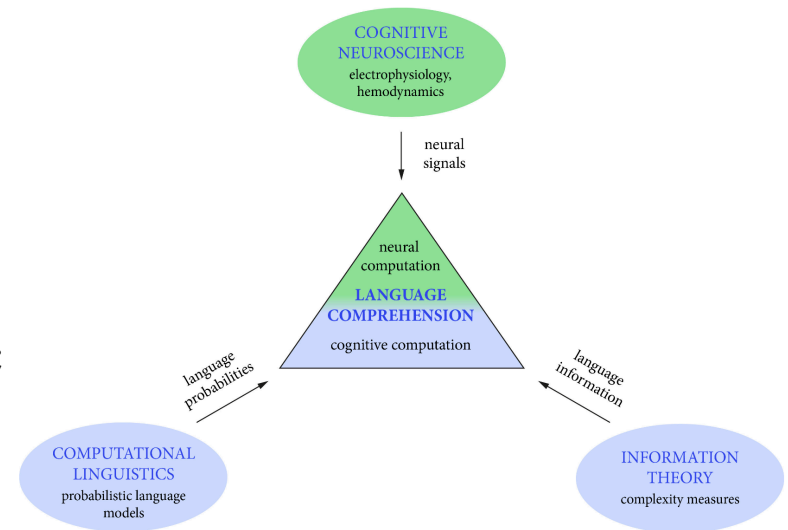


Fig. 1: Schematic depiction from Armeni et al., (2017)

General Approach

- ▶ Using naturalistic stimulus:
 - ▶ Computational modeling makes it easier to study the brain responses to naturalistic stimuli (Brennan, 2016)
 - ▶ Ecologically valid stimuli
 - ▶ Complement experimental approaches with controlled task-based designs
 - ▶ Easily reusable and shareable for different research questions

Research Question

- ▶ Do diathesis alternations and selectional restrictions on a verb have different neural bases?

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- ▶ Do diathesis alternations and selectional restrictions on a verb have different neural bases?
- ▶ Use PropBank (Kingsbury, 2002) & Resnik (1996)'s selectional preference strength metric

fMRI Experiment

- ▶ Dataset:
 - ▶ The audio stimulus was Antoine de Saint-Exupéry's *The Little Prince*, translated by David Wilkinson and read by Nadine Eckert-Boulet.
 - ▶ 1970 verbs attested in the story (401 unique); excluding modals, auxiliaries, and gerunds using the NLTK tagger & Stanford POS tagger

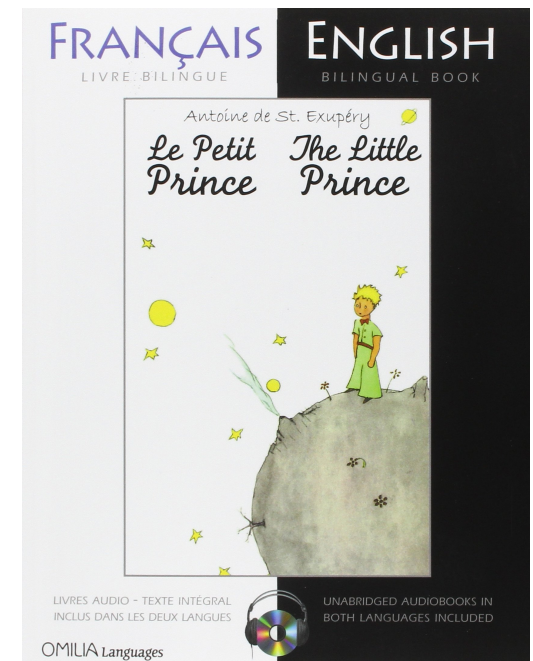


Fig. 2: Cover of *The Little Prince*

fMRI Experiment

- ▶ Experimental Design:
 - ▶ Participants (n=51, 32 female) were college-aged, right-handed, native English speakers
 - ▶ Listened to *The Little Prince's* audiobook for 1 hour 38 minutes across nine sections. (15,388 words total; 1,453 sentences)
 - ▶ Comprehension was confirmed through multiple-choice questions (90% accuracy, SD 3.7%).

fMRI Experiment

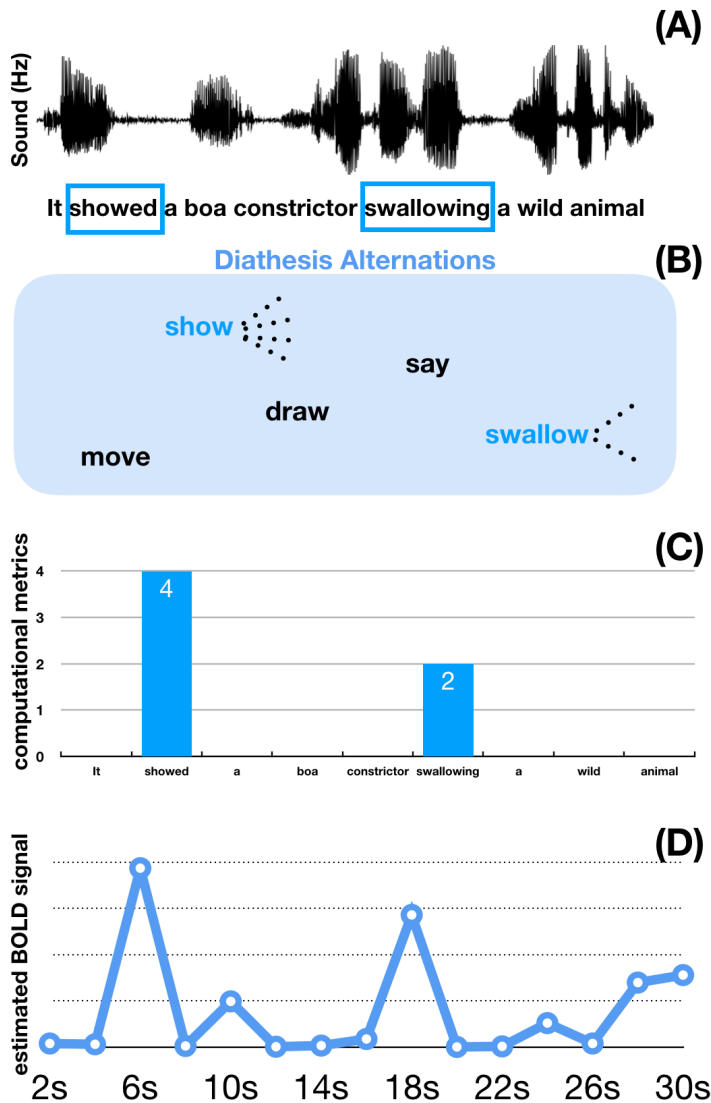


Fig. 3: Pipeline adapted from Bhattasali et al. (2018)

► Overview of analysis:

- The General Linear Model (GLM) typically used in fMRI data analysis is a time series linear regression (Poldrack et al., 2011).
- The regressors are convolved with the canonical HRF to create the estimated fMRI signal (BOLD), which is compared against the observed BOLD signal during passive story listening.

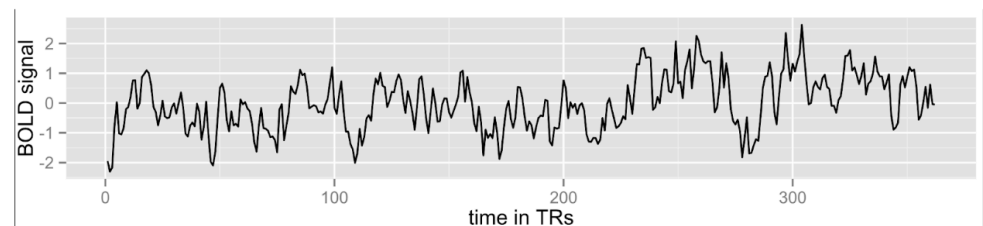


Fig. 4: Sample BOLD signal

fMRI Experiment

▶ Data Analysis:

- ▶ Preprocessing was carried out with AFNI version 16 and ME-ICA v3.2 (Kundu et al., 2011)
 - ▶ Images were normalized to the MNI-152 template
 - ▶ ME-ICA denoises T2* signal using ICA into BOLD and noise components from physiology, motion, scanner artifacts
- ▶ Statistical analyses carried out in SPM12 (Friston et al., 2007).
- ▶ 8 mm FWHM Gaussian smoothing kernel was applied on the contrast images from the first-level analysis to counteract inter-subject anatomical variation.
- ▶ Group-level results reported underwent FWE voxel correction for multiple comparisons which resulted in T-scores > 5.3

Statistical Analysis

- ▶ GLM Regressors:
 - ▶ **PropBank score**: Represents the number of diathesis alternations for a verb
 - ▶ **Selection preference strength**: Represent selectional restrictions on a verb
 - ▶ Word rate: Indicator of spoken word offset
 - ▶ Word frequency: log-frequency in movie subtitles (Brysbaert & New, 2009)
 - ▶ f0: fundamental frequency of the narrator's voice, reflects pitch
 - ▶ RMS amplitude: intensity, an acoustic correlate of volume

Statistical Analysis

- ▶ **PropBank scores:** Calculated from PropBank (Kingsbury, 2002), which consists of all the sentences from the Penn Treebank annotated with semantic roles with higher scores indicating more diathesis alternations.

Statistical Analysis

▶ PropBank scores

▶ *hang* ⇒ 8

▶ **hang**, *suspend, suspending*

▶ **hang**, *exist, be*

▶ **hang_on**, *wait*

▶ **hang_on**, *maintain possession of*

▶ **hang_up**, *terminate a phone call*

▶ **hang_up**, *stuck on*

▶ **hang_out**, *spend time socially*

▶ **hanging**, *execution*

Statistical Analysis

- ▶ **Selectional preference strength**: Calculated according to Resnik (1996) by estimating verb-direct object pairs from the Gigaword (Ferraro et al., 2014) & WaCkypedia (Baroni et al., 2009) corpora and then calculating the number of different WordNet semantic classes a given verb's direct objects falls into.

$$\Pr(v,c) = \frac{1}{N} \sum_{n \in \text{words}(c)} \frac{1}{|\text{classes}(n)|} \text{freq}(v,n)$$

Statistical Analysis

- ▶ Selectional preference strength:
- ▶ *pour*: <pour, juice>, <pour, milk>, <pour, water>, ...

<i>{act, action, activity}</i>	<i>{natural object}</i>
<i>{animal, fauna}</i>	<i>{natural phenomenon}</i>
<i>{artifact}</i>	<i>{person, human being}</i>
<i>{attribute, property}</i>	<i>{plant, flora}</i>
<i>{body, corpus}</i>	<i>{possession}</i>
<i>{cognition, knowledge}</i>	<i>{process}</i>
<i>{communication}</i>	<i>{quantity, amount}</i>
<i>{event, happening}</i>	<i>{relation}</i>
<i>{feeling, emotion}</i>	<i>{shape}</i>
<i>{food}</i>	<i>{state, condition}</i>
<i>{group, collection}</i>	<i>{substance}</i>
<i>{location, place}</i>	<i>{time}</i>
<i>{motive}</i>	

Table 1: The 25 noun semantic classes in WordNet (Miller, 1993)

Statistical Analysis

► Diathesis alternations vs Selectional restriction

Verb	PropBank scores	Selectional Preference Strength
pour	0.22	1
hang	0.33	0.91
call	1	0.52
catch	0.89	0.27
read	0.11	0.49
open	0.33	0.48
make	0.33	0.54
give	0.44	0.28

Table 2: Comparing diathesis alternation and selectional restriction metrics

Results

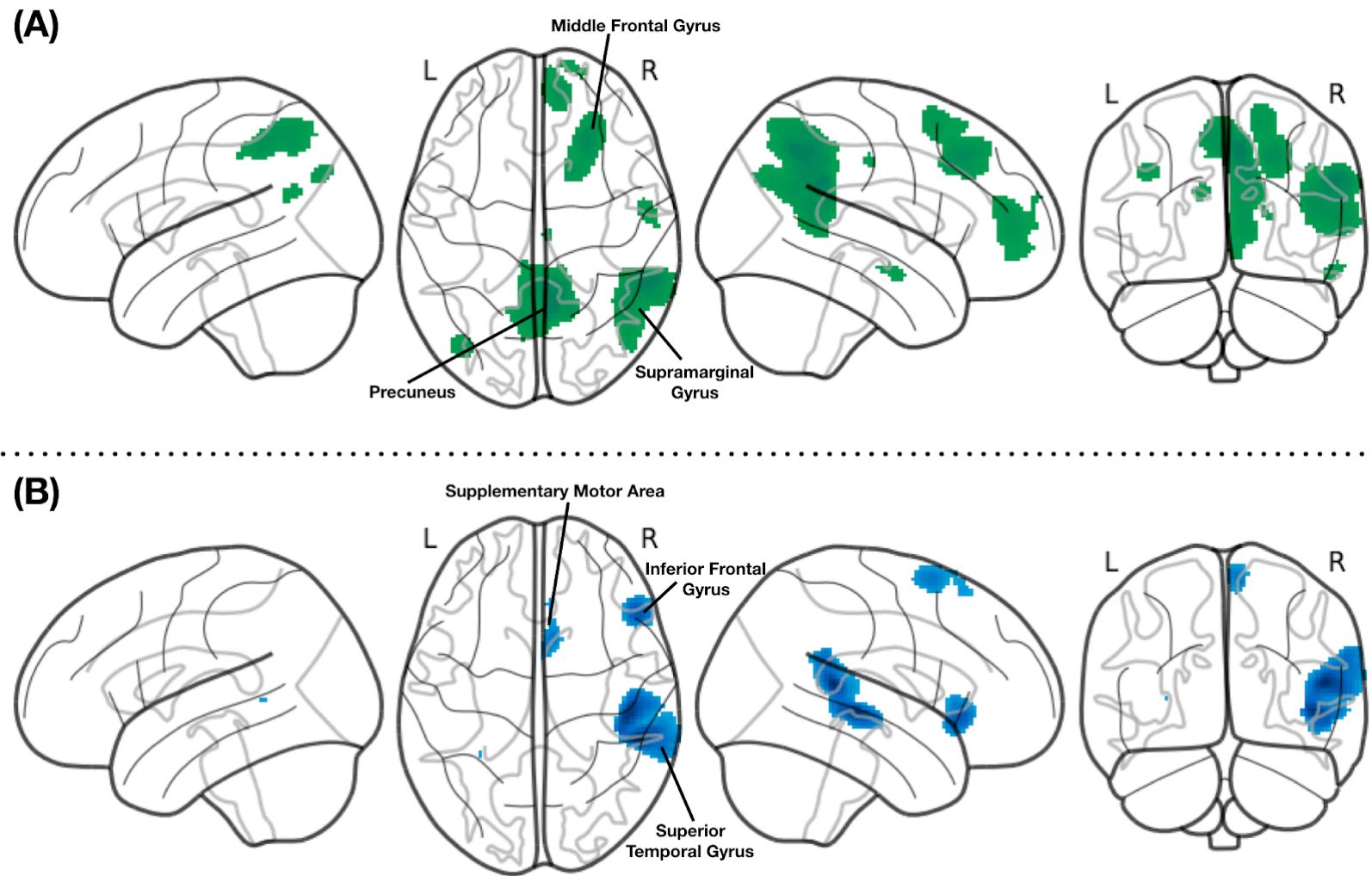
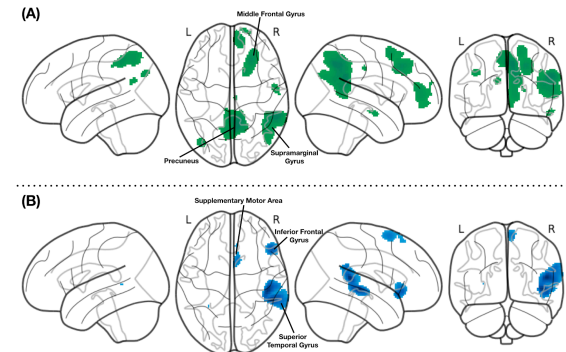


Fig. 5: (A): Whole-brain contrasts for diathesis alternations in green
(B): Whole brain contrasts for selectional restriction in blue

Results

► Results:

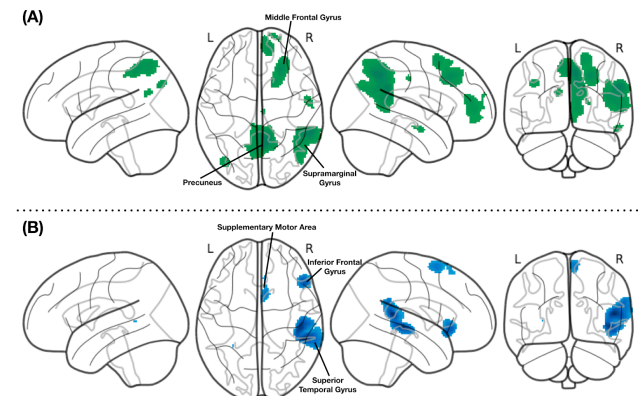
- Significant clusters for the diathesis alternations were observed in the right Supramarginal Gyrus and Middle Frontal Gyrus and bilateral Precuneus
- Significant clusters for selectional restrictions were observed in the right Superior Temporal Gyrus, Inferior Frontal Gyrus, and Supplementary Motor Area



Results

► Results:

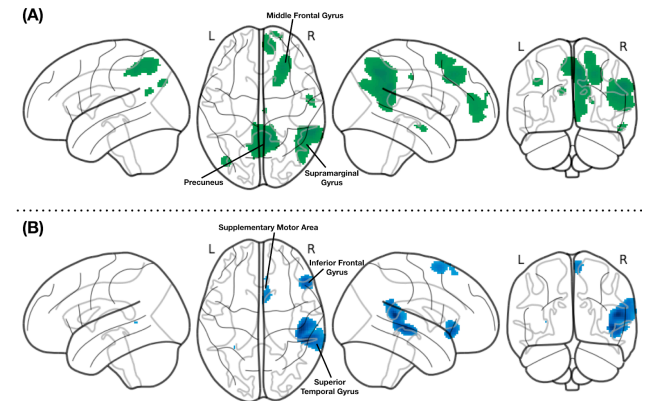
- Diathesis alternations results corroborate previous neuroimaging studies related to semantic roles and subcategorization (Shetreet et al., 2006; Thompson et al., 2010; Thompson et al., 2007; Meltzer-Asscher et al., 2013)
- Previous studies were controlled, task-based, block design experiments (e.g., lexical decision); our results replicate the findings with ecologically valid stimulus



Results

► Results:

- Selectional restrictions are consistent with other neuroimaging studies related to lexical-semantic processing (Kuperberg et al., 2000; Baker et al., 2001; Zempleni et al., 2007)



Conclusion

- ▶ Diathesis alternations and selectional restrictions evoke different pattern of activation in the brain
- ▶ Both metrics operationalize a degree of constraint: form and meaning
- ▶ Suggests that C-selection and S-selection have different neurobiological correlates.

Acknowledgements



Murielle Fabre



Christophe Pallier



Jonathan Brennan



This material is based upon work supported by the National Science Foundation under Grant Number 1607441.

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Thank you