

Cross-situational learning of word-pseudosign pairs in children and adults: a behavioral and event-related potential study

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BACKGROUND

- In human communication, both auditory and visual inputs are essential¹
- In child and adult communication, **gestures** often persist to accompany, emphasize, and complement speech^{2,3}.
- In conjunction with speech, gestures can be used to support **word learning** with pre-verbal children (*baby signing*⁴) and clinical populations⁵.
- Word learning can occur implicitly through the recognition of patterns in language input (*statistical learning*⁶).
- The **cross-situational learning**⁷ paradigm is used to replicate statistical learning mechanisms, mostly using **pseudowords** as referents for novel objects to investigate noun learning.
- Cross-situational learning** consists of **ambiguous learning trials**, useful to mimic real-life learning scenarios in experimental settings.
- Despite the role of gestures in language and communication, there is a lack of research on **cross-situational learning** of **visual language** and **speech**.

RESEARCH QUESTIONS

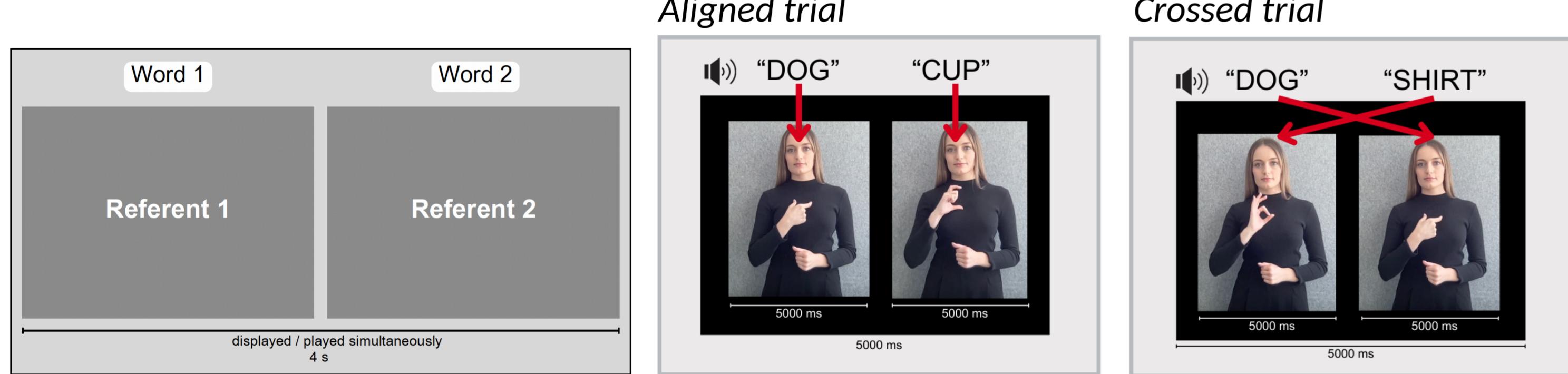
- Is it possible to map **familiar spoken words** to **novel pseudosigns**?
- Is this possible to do this **mapping** rapidly through statistical learning?
- Is it possible to build **semantic categories** of novel pseudosigns?
- In case of category violation, do pseudosigns elicit electrophysiological responses similar to spoken words (e.g., **N400** response)?

MATERIALS AND METHODS

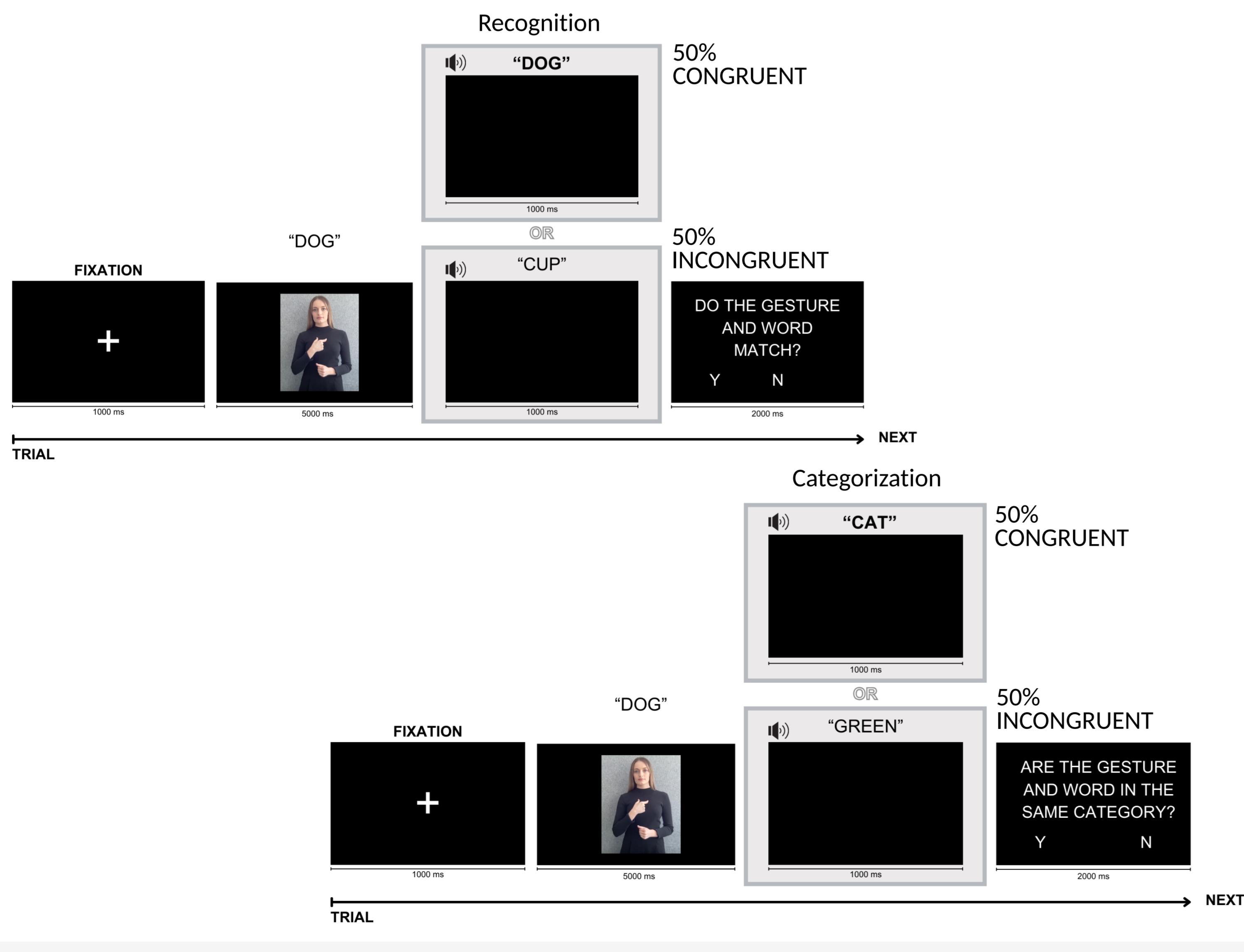
- participants: 25 children ($M = 9.11$ years, $SD = 14.8$) and 19 adults ($M = 27$ years, $SD = 4.9$)
- stimuli: 8 words (8 semantic categories) matched with 8 pseudosigns (non-iconic, phonotactically legal)
- dependent variables: response accuracy (%), d-prime, mean N400 amplitude (μ V)
- data analysis: signal detection theory⁸, cluster-based permutation analysis, linear mixed models

PROCEDURE

- Familiarization phase: cross-situational learning paradigm



- Recognition and categorization tasks: yes/no task (response button press)



BEHAVIORAL RESULTS

- Both groups performed above chance (accuracy and d-prime scores)
- Adults** performed significantly better than children in both tasks (d -prime ~ (Task * Group))
- No significant effect of tasks

Group	Task	Accuracy (%)			d-prime		
		M	SD	Wilcoxon W	M	SD	t-statistics
Adults (N = 19)	Recognition	87.5	13.6	190, $p < .001$	2.76	1.27	9.47, $p < .001$
	Categorisation	83.8	16.6	170, $p < .001$	2.48	1.34	8.04, $p < .001$
Children (N = 25)	Recognition	72.0	17.5	314, $p < .001$	1.47	1.33	5.52, $p < .001$
	Categorisation	69.0	18.2	295, $p < .001$	1.29	1.34	4.79, $p < .001$

Table 1: Behavioral performance of the two groups in recognition and categorization tasks.

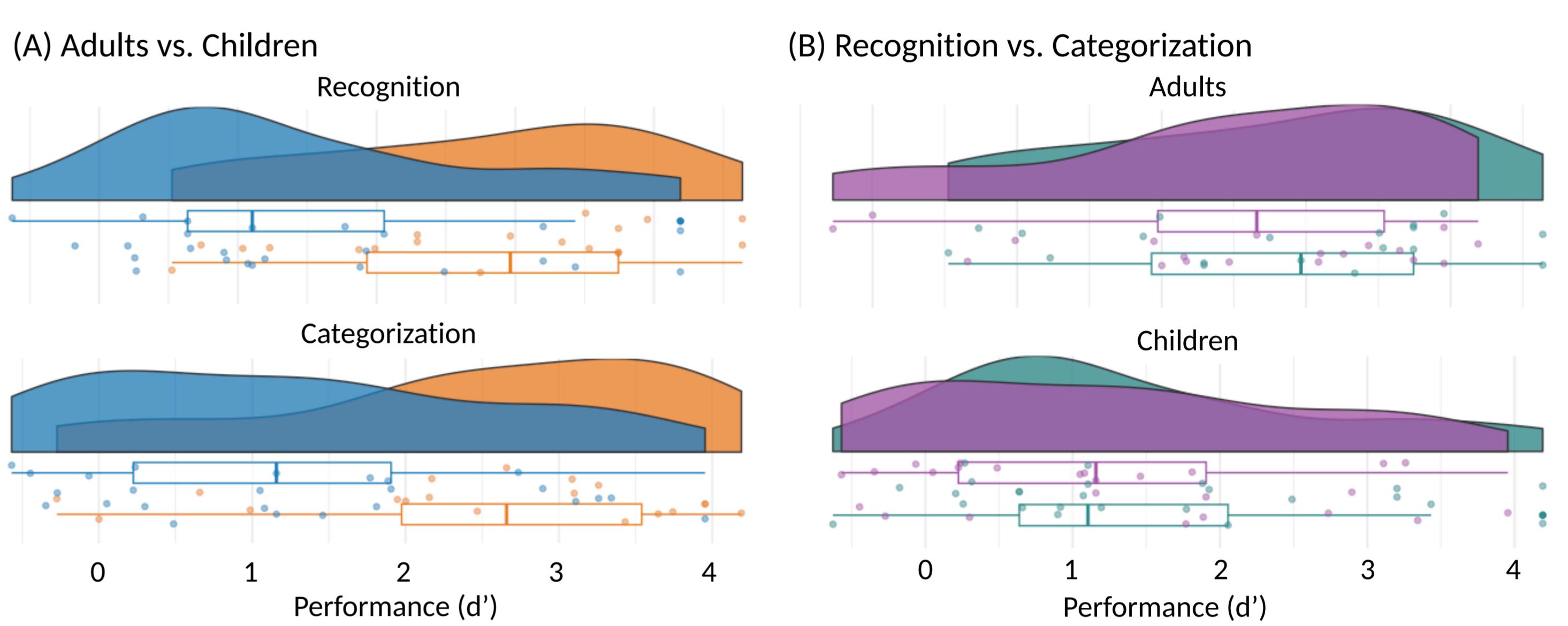


Fig. 1: Comparisons of performance scores expressed as sensitivity scores (d'). (A) Yellow = Adults, Blue = Children; (B) Green = Recognition task, Purple = Categorization task

ERP RESULTS

- Adults: N400 in recognition and categorization
- Children: N400 in recognition; in categorization, after additional analysis
- For both groups, LPC in recognition but not in categorization

Group	Task	Cluster type	Latency	p	Cohen's d
Adults (N=19)	Recognition	Negative	285-497	.001	-1.77
		Positive	625-997	.001	1.62
	Categorization	Negative	513-673	.005	-1.44
Children (N=24)	Recognition	Negative	156-464	.001	-1.39
		Positive	572-896	.002	1.23
	Categorization†	Negative	760-876	.043	-1.44

Table 2: Cluster-based permutation test results with effect sizes of the ERPs of the two groups in recognition and categorization tasks. † Results of the additional analysis on ERPs from correctly identified trials only.

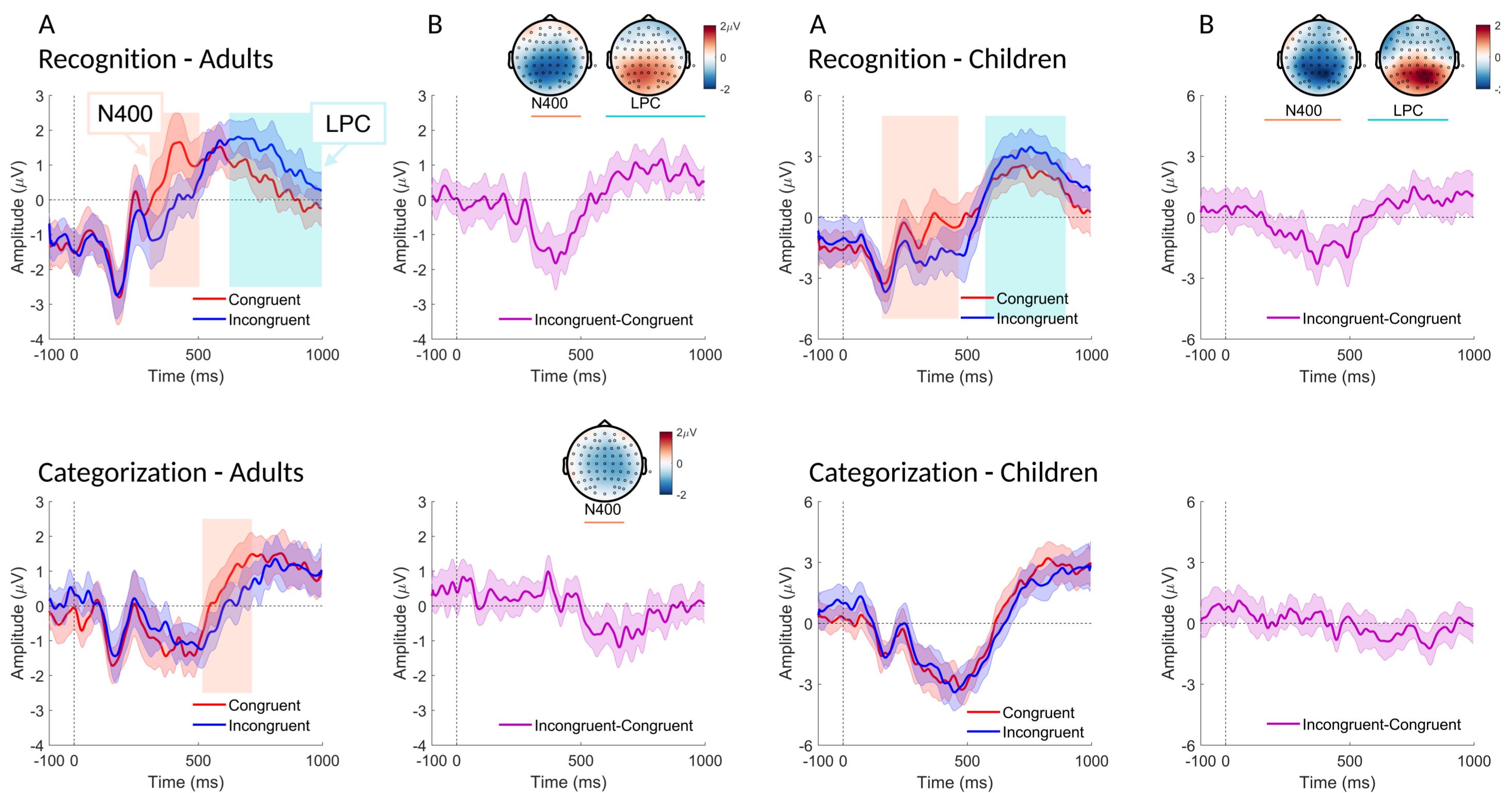


Fig. 2: Grand averaged ERPs of the adult group divided by task (recognition task, upper row; categorization task, bottom row). (A) The N400 and LPC time windows are highlighted in red and blue, respectively (B) The shading of the difference waveform encompasses 95% confidence intervals.

SUMMARY

Spoken words can be rapidly mapped onto novel pseudosigns via cross-situational learning. Pseudosigns can be rapidly associated with meaning and, in case of semantic violation, elicit brain responses (N400) similar to spoken words.

These findings suggest that spoken words + pseudosigns can constitute an **ecologic language input**. Future research should further explore how younger populations respond to this form of language input, to check for developmental differences in this process.

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