

Behavioural and electrophysiological markers of integration in novel word learning

Maria Korochkina^{1,2,3}, Lyndsey Nickels¹, Audrey Bürki²

¹ School of Psychological Sciences, Macquarie University, Australia

² Department of Linguistics, University of Potsdam, Germany

³ International Doctorate for Experimental Approaches to Language and Brain (IDEALAB): Universities of Groningen (The Netherlands), Newcastle (United Kingdom), Potsdam (Germany) & Macquarie University (Australia)

13th Annual Meeting of the Society for the Neurobiology of Language

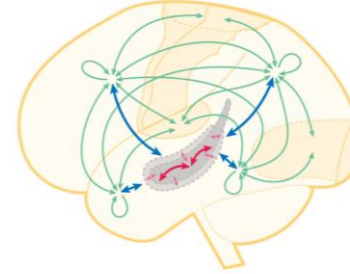
October 5, 2021



Markers of integration

Complementary Learning Systems [1-4]

- formation of new **episodic memory** representations
- integration of new information into **semantic memory**



→ Only integrated newly learned words compete with familiar words during lexical selection

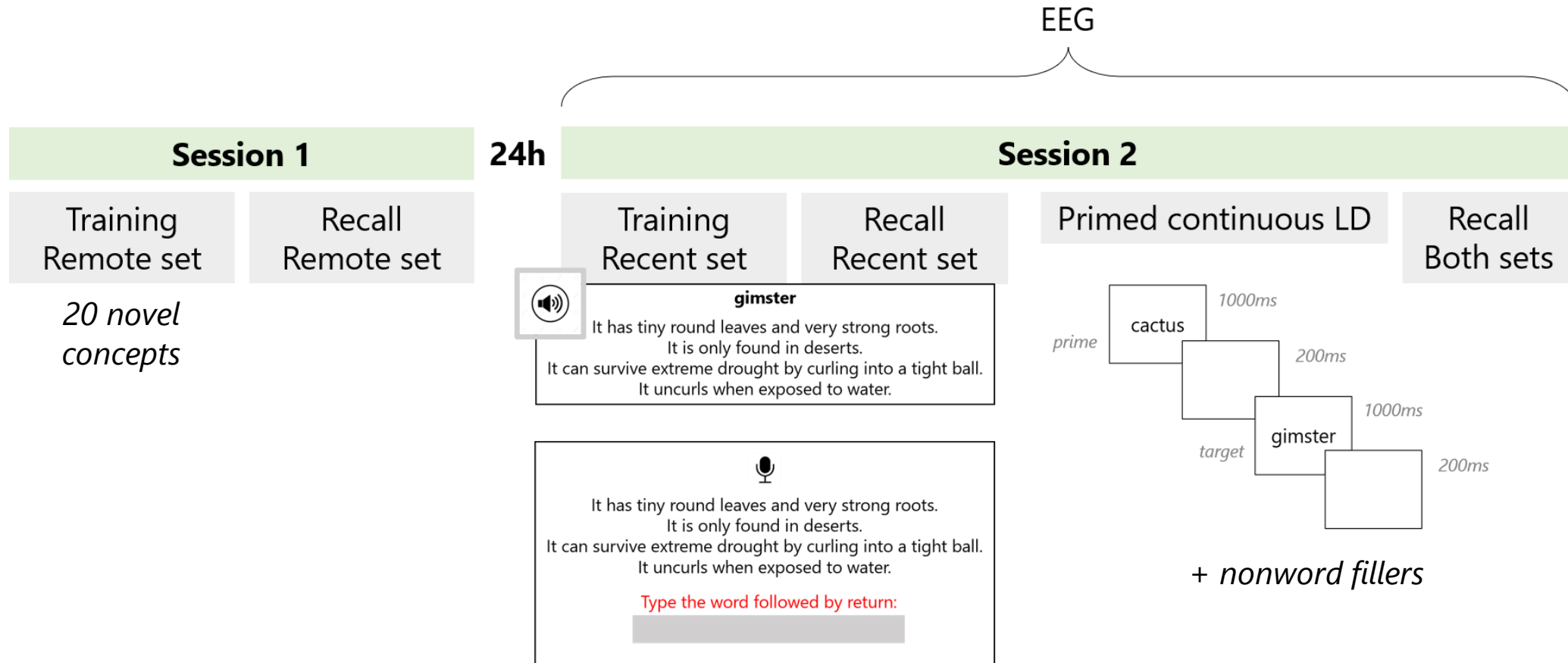
- **Semantic priming effect** as a marker of integration [e.g., 5-7]
 - longer RTs to (familiar) targets preceded by (trained novel) unrelated vs related primes

But... **how does the behavioural priming effect map onto electrophysiology?**

- **Automatic** or **controlled** lexical-semantic retrieval? [e.g., 8-9]
- Two spatiotemporal windows, the **N400** and the **LPC** [e.g., 10-11]



Methods



Participants

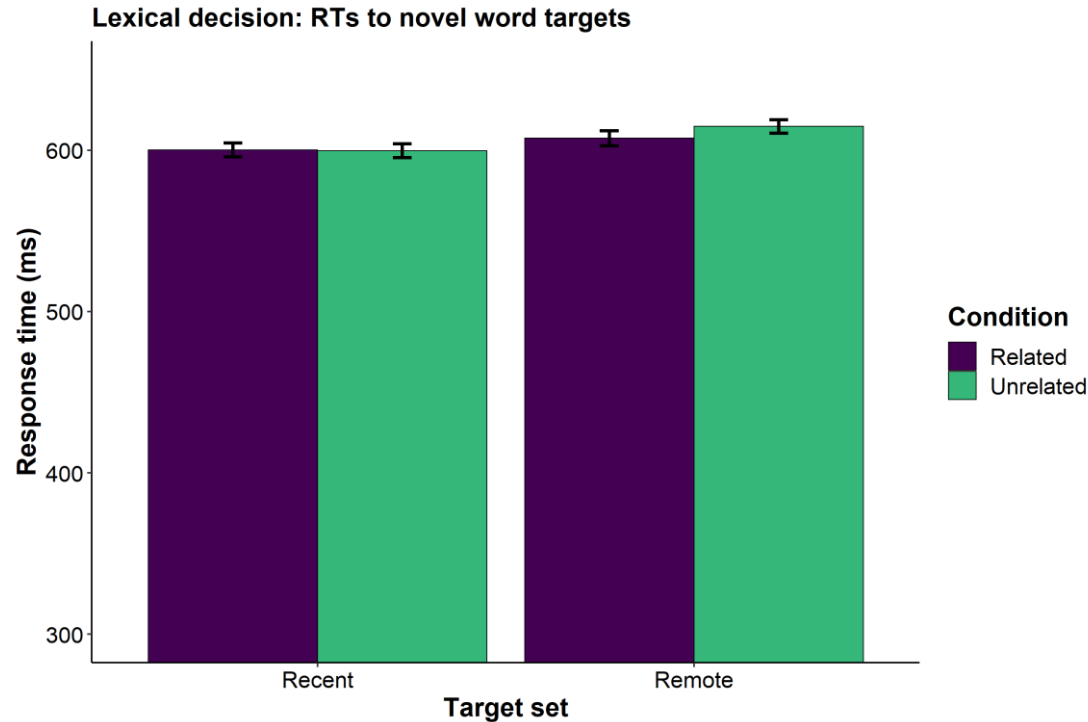
- 72 monolingual speakers of Aus English
- 28 males, *mean* age 20.94, *SD* 3.86
- Analysed: N = 71, 27 males, *mean* age 20.94, *SD* 3.87

Pre-registered at <https://osf.io/su7d3> on March 17, 2021



Behavioural results

Bayesian LMMs with correlated varying intercepts and slopes

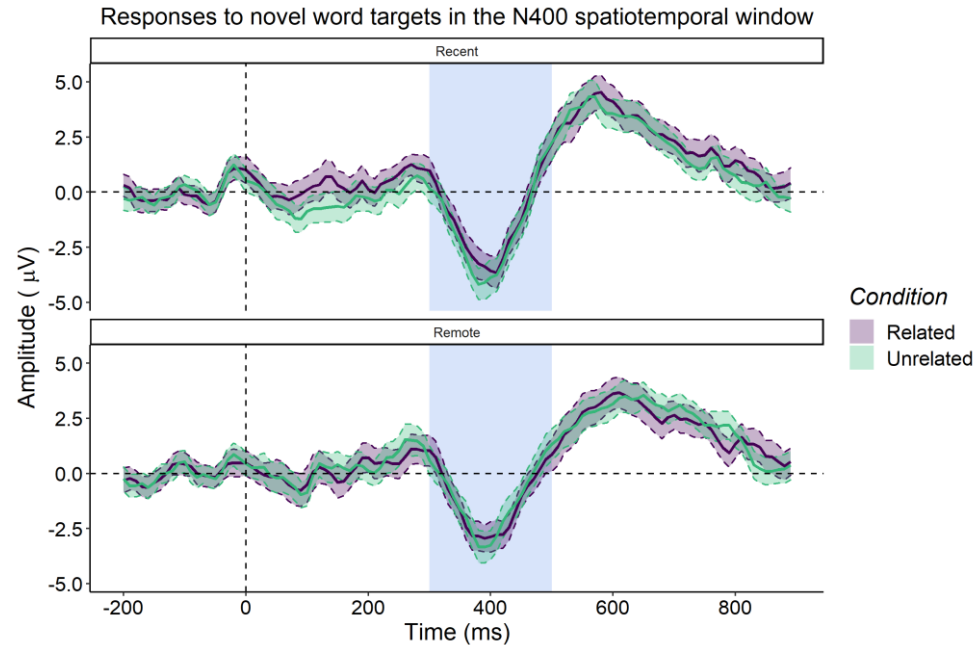


- Anecdotal evidence **for** ME set, $\text{diff} = 11\text{ms}$, $\beta = -0.01$, $95\%CrI = [-0.018, 0.002]$, $BF_{10} = 3.95$
- Moderate evidence **against** ME condition, $\beta = 0.003$, $95\%CrI = [-0.005, 0.012]$, $BF_{10} = 0.12$
- Moderate evidence **against** the interaction, $\beta = -0.004$, $95\%CrI = [-0.011, 0.004]$, $BF_{10} = 0.13$

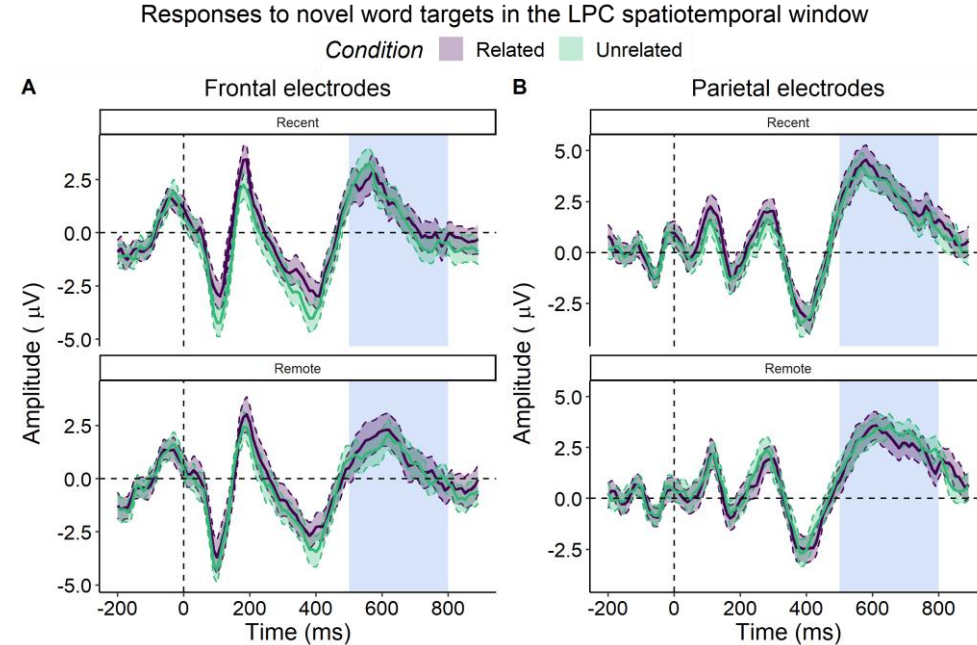


Semantic priming: EEG analysis 1

Bayesian distributional regression models with correlated varying intercepts and slopes for μ and varying intercepts for δ [12]



Strong evidence **against** ME condition
 ($\beta = -0.06$, $95\%CrI = [-0.26, 0.14]$, $BF_{10} = 0.017$), ME
 set ($\beta = -0.14$, $95\%CrI = [-0.33, 0.06]$, $BF_{10} = 0.04$)
 & interaction ($\beta = -0.08$, $95\%CrI = [-0.03, 0.11]$,
 $BF_{10} = 0.03$)

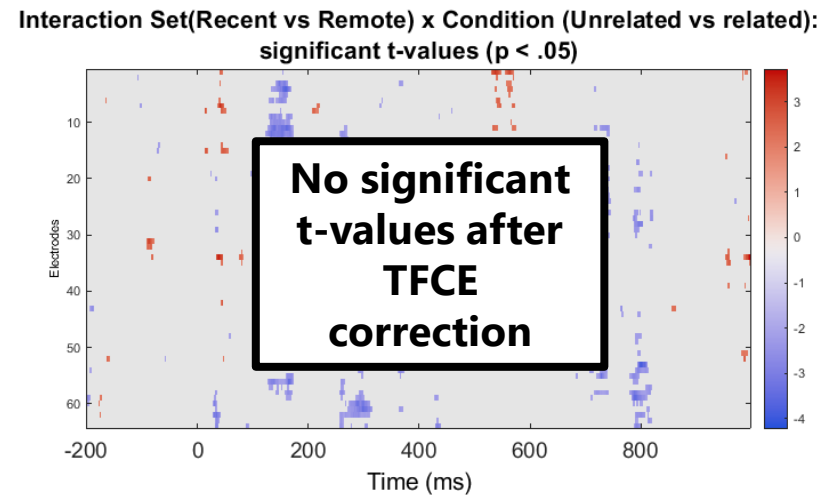
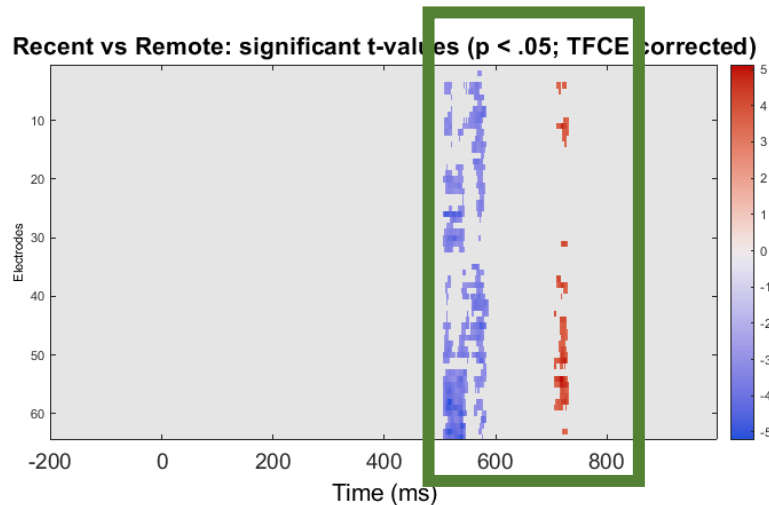
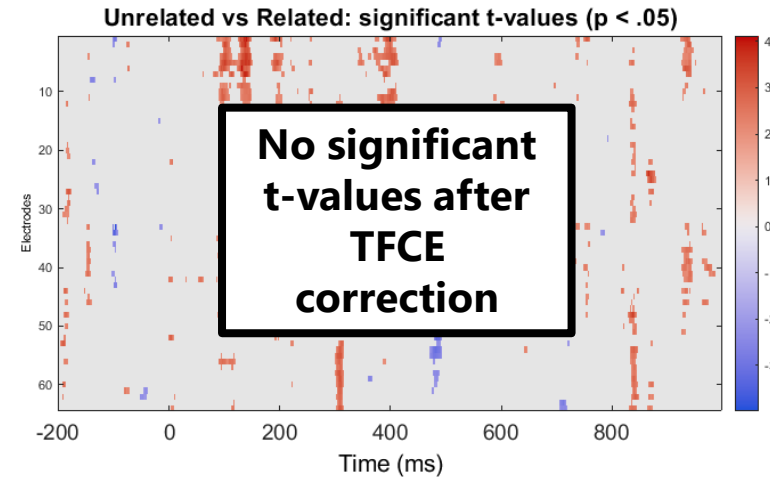
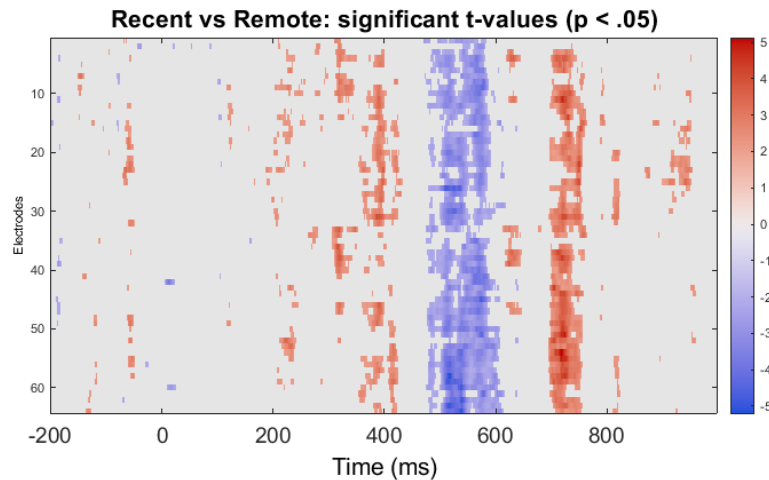


Strong evidence **against** all MEs and
 interactions both at frontal and
 parietal electrodes



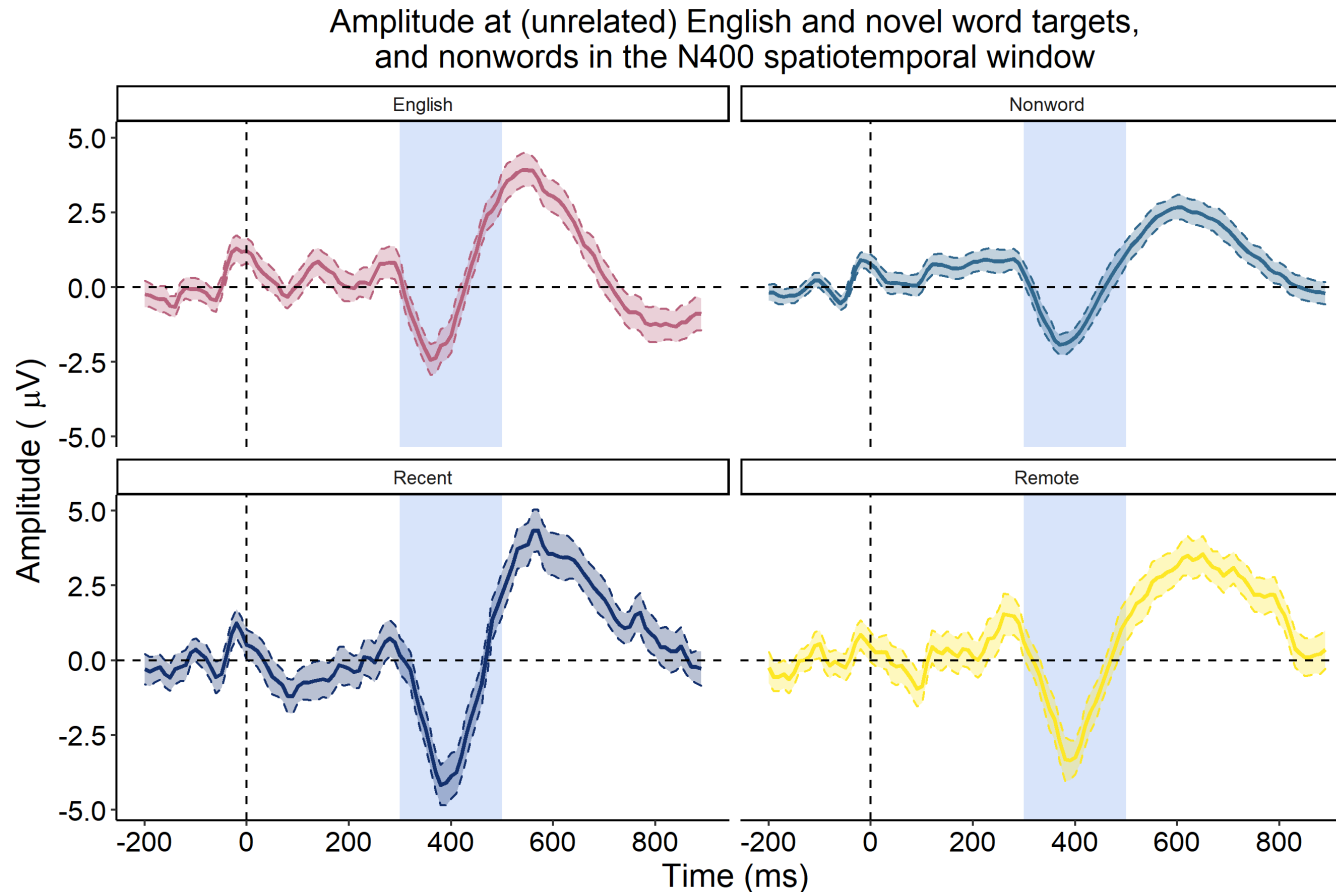
Semantic priming: EEG analysis 2

- Mass univariate analysis [e.g., 13-14]
- Correction for multiple comparisons with Threshold Free Cluster Enhancement (TFCE) [15-16]



Lexicality: EEG analysis 1

Bayesian distributional regression models with correlated varying intercepts and slopes for μ and varying intercepts for δ



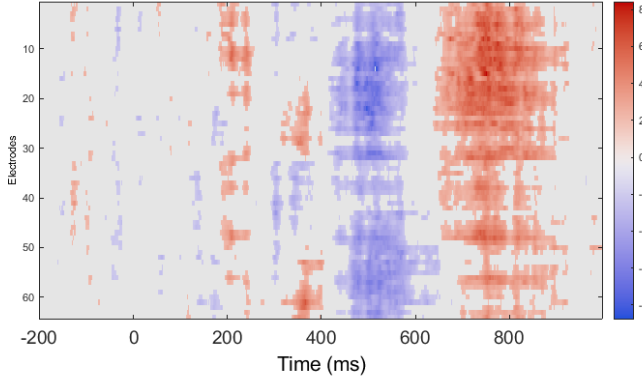
- Moderate evidence **for** the difference between nonwords and English words ($\beta = -0.4$, $95\%CrI = [-0.84, 0.03]$, $BF_{10} = 6.87$)
- Extreme evidence **for** the difference between recent words and nonwords ($\beta = -0.93$, $95\%CrI = [-1.42, -0.45]$, $BF_{10} = 846.21$)
- Strong evidence **against** the difference between remote and recent words ($\beta = 0.39$, $95\%CrI = [-0.12, 0.9]$, $BF_{10} = 0.09$)



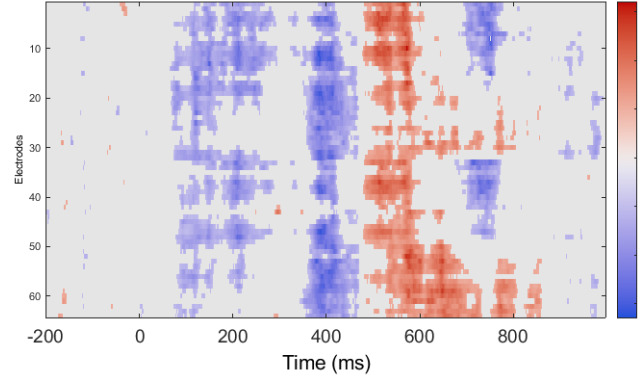
Lexicality: EEG analysis 2

- Mass univariate analysis
- Correction for multiple comparisons with Threshold Free Cluster Enhancement (TFCE)

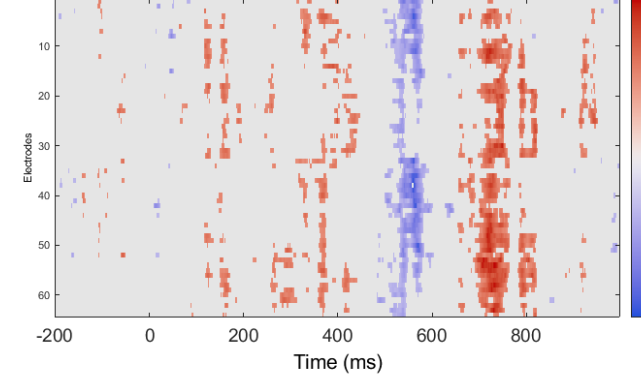
Nonwords vs English words: significant t-values ($p < .05$)



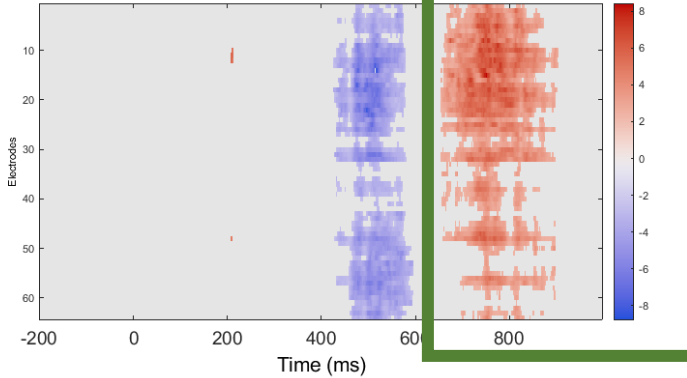
Recent words vs nonwords: significant t-values ($p < .05$)



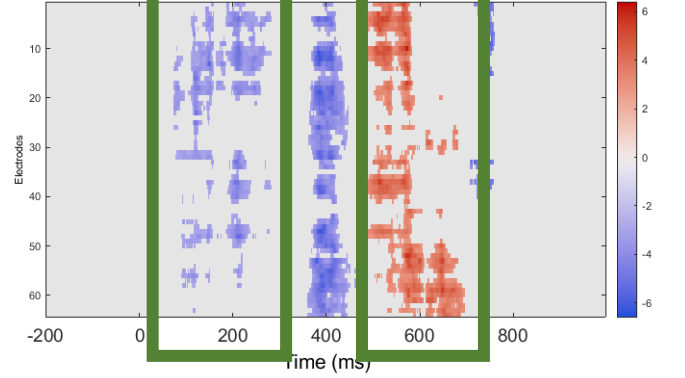
Remote vs recent words: significant t-values ($p < .05$)



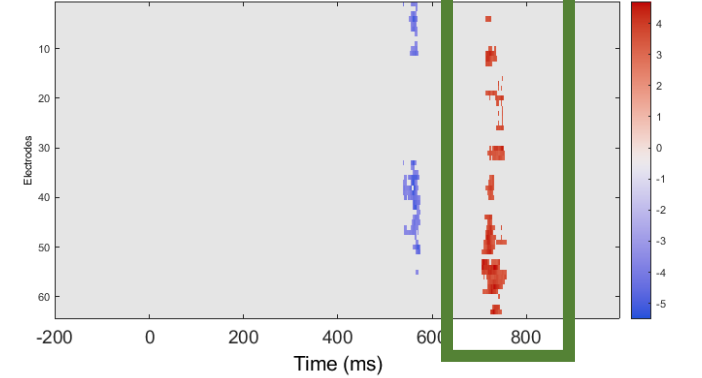
Nonwords vs English words: significant t-values ($p < .05$; TFCE corrected)



Recent words vs nonwords: significant t-values ($p < .05$; TFCE corrected)



Remote vs recent words: significant t-values ($p < .05$; TFC corrected)



Conclusions

- Compared to *nonwords*, (*unrelated*) *recent words* elicited
 - greater negativity 100-200ms post onset
 - greater negativity in the N400 spatiotemporal window
 - greater positivity 500-700ms post onset
 - No difference between (*unrelated*) *recent and remote words* in the N400 spatiotemporal window, but *remote words* elicited greater positivity 700-750ms post onset
 - Evidence *against semantic priming* effects in both sets
 - Greater positivity to (*related & unrelated*) *recent vs remote words* 500-600ms post onset
- Immediately after exposure, novel words are processed differently compared to previously unseen nonwords in **both episodic and semantic memory**
- 24h after exposure to novel words (remote), the system still **relies on episodic memory** to distinguish between those words and the words acquired on the following day (recent)



References

- [1] Davis, M., & Gaskell, M. (2009). A Complementary Systems Account of word learning: neural and behavioural evidence. *Philos. Trans. R. Soc. B*, 364, 3773–3800. <https://doi.org/10.1098/rstb.2009.0111>.
- [2] Kumaran, D., Hassabis, D., & McClelland, J. L. (2016). What learning systems do intelligent agents need? Complementary Learning Systems Theory updated. *Trends Cogn. Sci.*, 20(7), 512–534. <https://doi.org/10.1016/j.tics.2016.05.004>.
- [3] McClelland, J. L., McNaughton, B. L., & O'Reilly, R. C. (1995). Why there are complementary learning systems in the hippocampus and neocortex: Insights from the successes and failures of connectionist models of learning and memory. *Psychol. Rev.*, 102(3), 419–457. <https://doi.org/10.1037/0033-295X.102.3.419>.
- [4] McClelland, J.L., McNaughton, B. L., & Lampinen, A.K. (2020). Integration of new information in memory: new insights from a complementary learning systems perspective. *Philos. Trans. R. Soc. B*, 375: 20190637. <http://dx.doi.org/10.1098/rstb.2019.0637>.
- [5] Dumay, N., Gaskell, M. G., & Feng, X. (2004). A day in the life of a spoken word. In K. Forbus, D. Gentner, & T. Regier (Eds.), *Proceedings of the twenty-sixth annual conference of the Cognitive Science Society* (pp. 339 – 344). Erlbaum.
- [6] Tamminen, J., & Gaskell, M.G. (2013). Novel word integration in the mental lexicon: Evidence from unmasked and masked semantic priming. *Q. J. Exp. Psychol.*, 66 (5), 1001-1025. <https://doi.org/10.1080/17470218.2012.724694>.
- [7] van der Ven, F., Takashima, A., Segers, E., & Verhoeven, L. (2015). Learning word meanings: Overnight integration and study modality effects. *PLoS ONE*, 10, e0124926. <https://doi.org/10.1371/journal.pone.0124926>.
- [8] Bakker, I., Takashima, A., van Hell, J., Janzen, G., & McQueen, J. (2015). Tracking lexical consolidation with ERPs: Lexical and semantic-priming effects on N400 and LPC responses to newly-learned words. *Neuropsychologia*, 79, 33-41. <https://doi.org/10.1016/j.neuropsychologia.2015.10.020>.
- [9] Liu, Y., & van Hell, J. (2020). Learning novel word meanings: An ERP study on lexical consolidation in monolingual, inexperienced foreign language learners. *Language Learning*, 70(S2), 45-74. <https://doi.org/10.1111/lang.12403>.
- [10] Kutas, M., & Federmeier, K.D. (2011). Thirty years and counting: finding meaning in the N400 component of the event-related brain potential (ERP). *Annu. Rev. Psychol.*, 62, 621-647. <https://doi.org/10.1146/annurev.psych.093008.131123>.
- [11] Rugg, M.D., & Curran, T. (2007). Event-related potentials and recognition memory. *Trends Cogn. Sci.*, 11(6), 251-257. <https://doi.org/10.1016/j.tics.2007.04.004>.
- [12] Bürkner, P.C. (2017). Bayesian Distributional Non-Linear Multilevel Modelling with the R package *brms*. https://mran.microsoft.com/snapshot/2017-04-05/web/packages/brms/vignettes/brms_multilevel.pdf
- [13] Groppe, D. M., Urbach, T. P., & Kutas, M. (2011a). Mass univariate analysis of event-related brain potentials/fields I: A critical tutorial review. *Psychophysiology*, 48(12), 1711–1725. doi: <https://doi.org/10.1111/j.1469-8986.2011.01273.x>.
- [14] Groppe, D. M., Urbach, T. P., & Kutas, M. (2011b). Mass univariate analysis of event-related brain potentials/fields II: Simulation studies. *Psychophysiology*, 48(12), 1726–1737. doi: <https://doi.org/10.1111/j.1469-8986.2011.01272.x>.
- [15] Smith, S.M., & Nichols, T.E. (2009). Threshold-free cluster enhancement: addressing problems of smoothing, threshold dependence and localisation in cluster inference. *Neuroimage*, 1, 44(1): 83-98. <https://doi.org/10.1016/j.neuroimage.2008.03.061>.
- [16] Pernet, C.R., Latinus, M., Nichols, T.E., & Rousselet, G.A. (2015). Cluster-based computational methods for mass univariate analyses of event-related brain potentials/fields: A simulation study. *J Neurosci Methods*, 30 (250): 85-93. <https://doi.org/10.1016/j.jneumeth.2014.08.003>.