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## Proceedings of the Annual Meeting of the Cognitive Science Society

### Title

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### Permalink

<https://escholarship.org/uc/item/1m26r4hg>

### Journal

Proceedings of the Annual Meeting of the Cognitive Science Society, 48(0)

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### Publication Date

2026

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Peer reviewed

# Modeling individual differences in learning and memory using large language models

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## Abstract

Computational cognitive models allow formalizing theories of cognition and arbitration between competing accounts. Traditional cognitive modeling (1) entails carefully handcrafting cognitive models, which require substantial domain knowledge and programming expertise; (2) favors single shared model class for all participants, limiting its ability to capture the full range of individual differences. Recent advances in large language models (LLMs) offer a new route for addressing these challenges. In this work, we develop a pipeline for scalable generation of computational cognitive models that prompts an LLM to propose a bespoke cognitive model for each participant given task instructions, their full behavioral trajectory, and a code template, and iteratively refines its proposal based on its predictive performance. In the domains of planning and working memory, we find the resulting models contain participant-specific mechanisms that accurately capture individual differences in learning and memorization, leading to substantial improvement in predictive performance while passing all posterior-predictive checks.