



From Prompting to Epistemic Proactivity: Temporal Trajectories of Student–AI Interaction in Mathematics Learning

Rania Abdelghani¹, Peter Kaiser^{1,2}, Kou Murayama¹

¹ The Hector Institute of Educational Sciences and Psychology, University of Tübingen, Germany

² Department of Mathematics, Faculty of Sciences, University of Tübingen, Germany

Background

- Students are increasingly using (Generative AI) GenAI independently to do learning tasks. However, little is still known about which interaction patterns / strategies can help rather than compromise their active learning.

- For this use to be pedagogically useful, we suggest the need for **epistemic proactivity during AI use**:

- **Regulation of the AI towards one's epistemic needs,**
- **Help-Seeking strategies that seek to support rather than replacement of cognitive work,**
- **Domain-specific proactive engagement.**

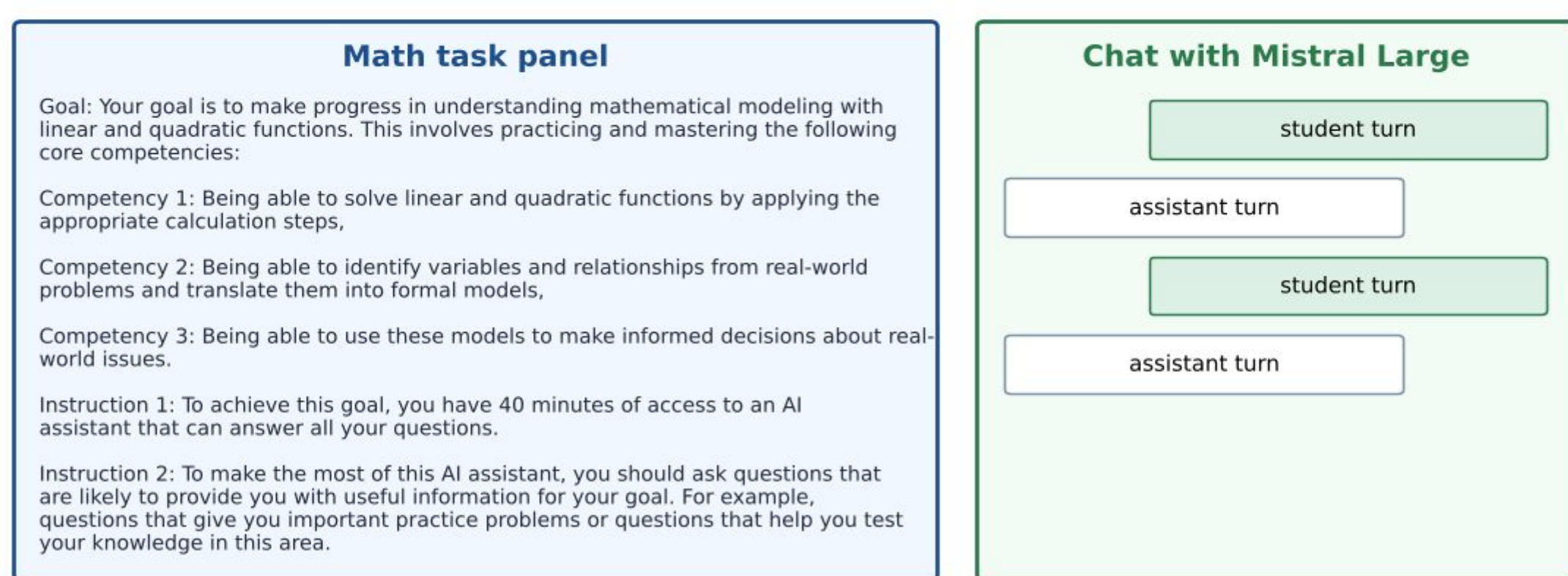
Study Design

- German schoolchildren (age = 14-15) interact with an AI tutor (Mistral Large) to practice a mathematical skill by asking questions.
- Several measures are also collected: domain knowledge, cognitive load, metacognitive skills, AI knowledge.



Step 1 — Pre-chat intention screen
8 multi-select intent options (e.g., "step-by-step", "only hints", "just give me the answer") · 5-pt AI-confidence rating

Step 2 — 40-minute open-ended chat



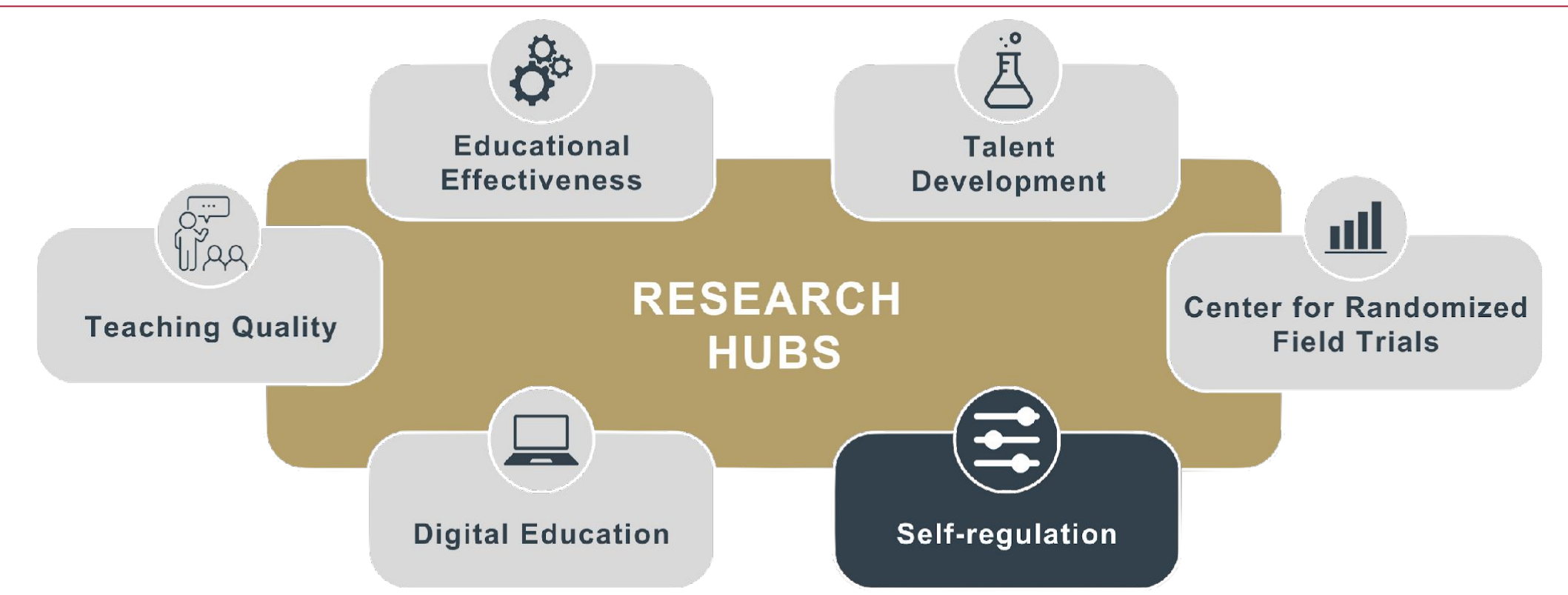
Discussion

Students who were able to keep a **balance** between epistemically-proactive and reactive behaviors, across the whole session, or who were successful in developing such behaviors over time, were the ones with better performance scores.

→ Temporal developments in help-seeking and domain-specific strategies should be monitored during AI use, to better understand students' AI-supported learning.

References

- Stadler, M., Bannert, M., Sailer, M.: Cognitive ease at a cost: LLMs reduce mental effort but compromise depth in student scientific inquiry. Computers in Human Behavior 160 (2024). <https://doi.org/10.1016/j.chb.2024.108386>
- Wang, J., Fan, W.: The effect of ChatGPT on students' learning performance, learning perception, and higher-order thinking: insights from a meta-analysis. Springer (2025). <https://doi.org/10.1057/s41599-025-04787-y>
- Yan, L., Greiff, S., Lodge, J.M., Gašević, D.: Distinguishing performance gains from learning when using generative AI. Nature Publishing Group (2025). <https://doi.org/10.1038/s44159-025-00467-5>



Research Questions and Analysis Method

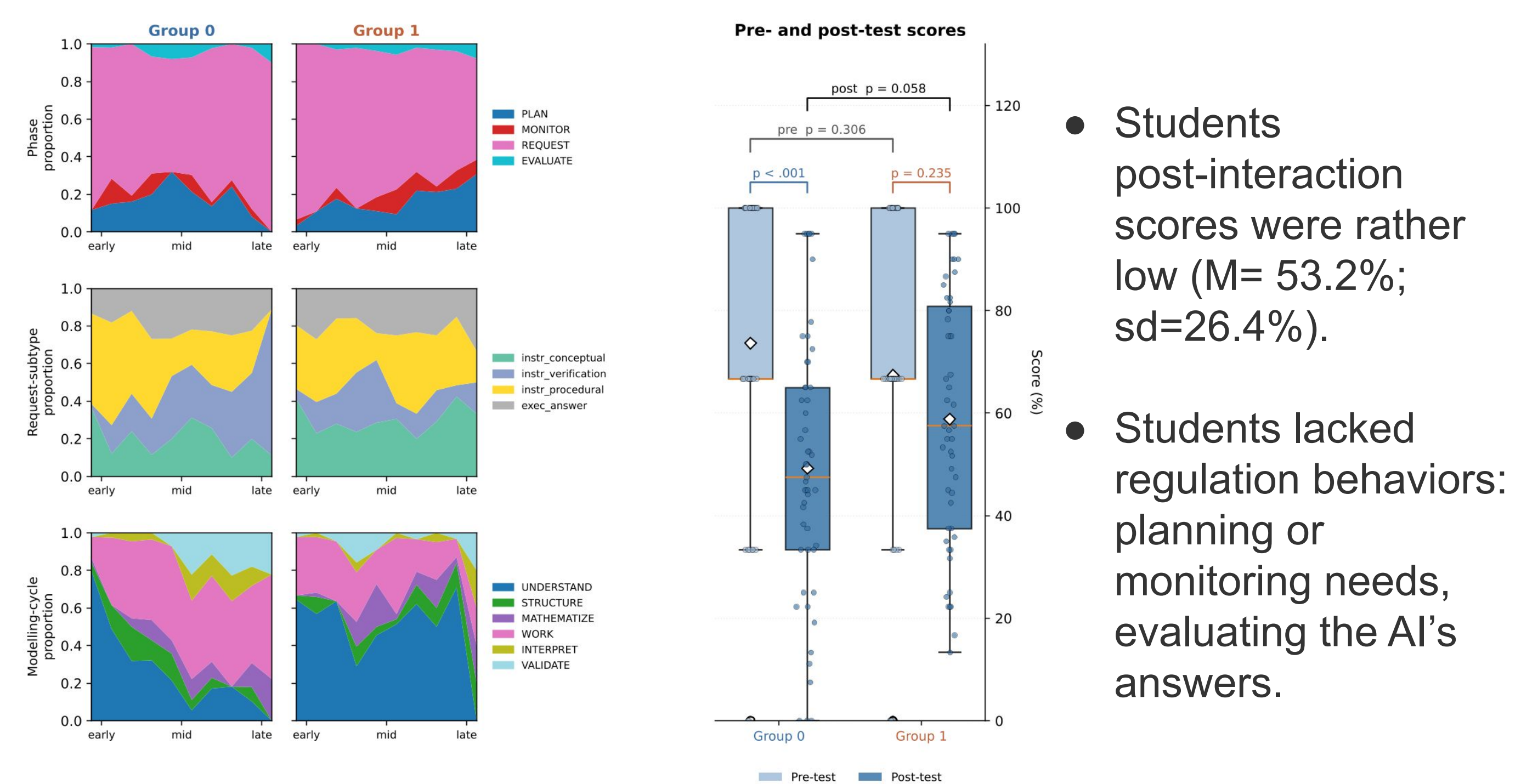
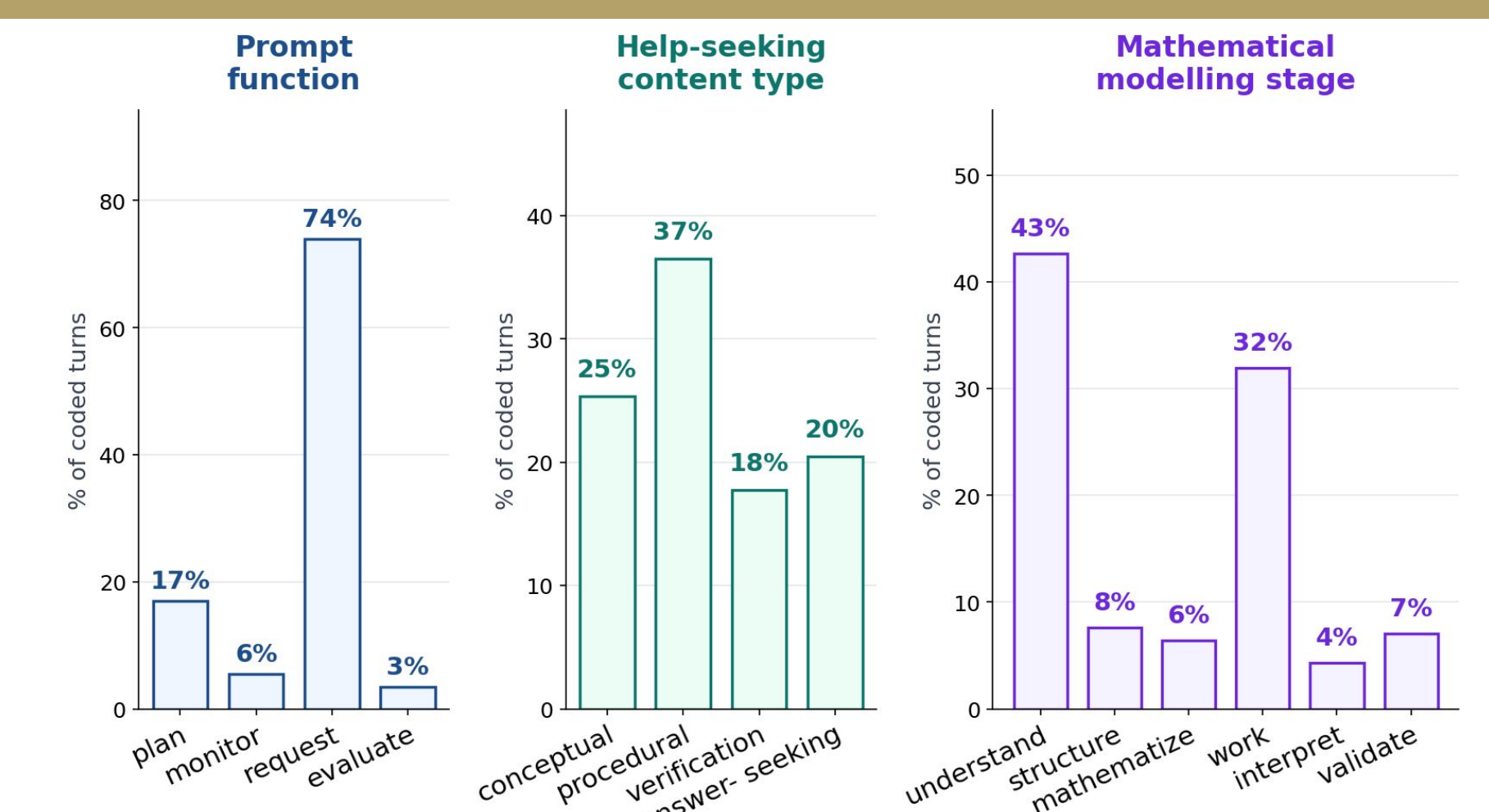
RQ1: How do students **seek information** and **regulate** their use of an LLM tutor while learning a mathematical-modeling skill?

RQ2: What features of these interactions impact their **independent tasks outcomes**?

→ We use LLMs to code (cross-validated with human-expert codes):

- **Presence of Self-regulation functions,**
- **Help-Seeking strategies content,**
- **Mathematical modelling activity**

Main Results (N= 112)



- Students post-interaction scores were rather low (M= 53.2%; sd=26.4%).
- Students lacked regulation behaviors: planning or monitoring needs, evaluating the AI's answers.

- They tended to start the interaction with **high-quality help-seeking** and **mathematical modelling** strategies, but shifted toward more **mathematical validation** and **executive answer-seeking** towards the end (i.e less epistemic proactivity).
- These two **temporal shifts** were associated with their learning performance post-interaction (OLS model: F= 8.98, adj-R² .3 = , p <.0001).