## Evaluating Large Language Model Quality in Resource-Constrained Environments: An Educational Stakeholders' Survey on Accuracy, Completeness, and Readability in Brazil

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Artificial Intelligence in Education (AIED) has shown potential to personalize learning, automate feedback, and support teachers in classrooms [Chen et al. 2022, Crompton et al. 2024]. However, countries in the Global South still face barriers related to infrastructure, devices, and connectivity, deepening inequalities in educational opportunities [Gasevic et al. 2018, Barron Rodríguez and Cobo 2022]. Lightweight Large Language Models (LLMs), designed to run locally on smartphones, emerge as a viable alternative. This study, derived from a dissertation project [Barros 2024], examines their feasibility and quality in Brazilian education.

The analysis focused on three lightweight LLMs: Gemma, Qwen and Red-Pajama. Answers to 18 educational questions, based on Bloom's taxonomy and translated into Portuguese, were evaluated by 196 participants from diverse regions. Using Likert-scale questionnaires, evaluators judged accuracy, completeness and readability, following approaches from previous educational technology research [Graesser et al. 2004, Vanlehn 2006]. Unlike studies focused on latency or performance [Monteiro Santos et al. 2024], our emphasis was on perceived response quality.

Results indicate that Qwen and Gemma scored higher in accuracy and completeness, while RedPajama, despite readability, struggled with Portuguese, lowering overall quality. These findings show that LLMs can operate offline in constrained devices with pedagogical relevance, but also highlight the need for language adaptation to ensure equitable impact in contexts such as Brazilian public schools [Isotani et al. 2023].

Implications are significant: lightweight LLMs may democratize access to AIED, enabling teachers and students to benefit from adaptive feedback without internet access [Veloso et al. 2023, Rodrigues et al. 2024]. For researchers, they open opportunities to study trade-offs between efficiency and linguistic quality [Yan et al. 2024]. For policy-makers, they stress the urgency of investing in localized models to mitigate the digital divide. Limitations include reliance on self-reported evaluations, restriction to Earth Sciences questions, and exclusion of Portuguese-specific models. Future work should

extend analyses to other domains, integrate expert validation, and explore optimization techniques to strengthen applicability.

Through the systematic evaluation of lightweight LLMs in offline contexts, this study contributes to the advancement of the AIED unplugged, stressing the urgency of designing educational technologies that are not only effective in fostering learning but also inclusive and accessible to students in underserved regions.

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