Learnersourcing: Student-generated Content @ Scale

3rd Annual Workshop

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ABSTRACT

We propose the third annual workshop on Learnersourcing: Student-generated Content @ Scale, reimagined for an era where AI is rapidly transforming the educational landscape. This full-day workshop is designed to explore the vast potential of learnersourcing, which combines the efforts of humans, AI, and other data sources to create and assess educational materials. invites instructors, researchers, learning engineers, and professionals from diverse backgrounds to explore the innovative intersection of human insight and AI-sourced content in learnersourcing. By drawing on principles from education, crowdsourcing, learning analytics, data mining, and natural language processing, we aim to foster an environment where all participants can learn from and contribute to the conversation. As AI tools become increasingly integral to content creation, our program will delve into pedagogically sound learnersourcing practices that effectively integrate these technologies. Participants will examine the broader implications of AI on learnersourcing, from ensuring content quality to fostering authentic student engagement, and will be provided with practical guidelines for incorporating AI in ways that enrich learning. Through bringing together these different perspectives, attendees will leave equipped with both a conceptual framework and practical strategies to engage with learnersourcing.

CCS CONCEPTS

• Human-centered computing ~ Collaborative and social computing ~ Collaborative and social computing theory, concepts and paradigms ~ Computer supported cooperative work

• Applied computing ~ Education ~ Interactive learning environments

KEYWORDS

Learnersourcing; Student-generated content; Question creation; Educational content creation

1 Background

Learnersourcing involves students in generating and evaluating educational content such as questions, hints, and examples [7]. Previous research has demonstrated that this enhances student learning, by actively engaging them with the material and

invoking critical thinking [15]. This approach utilizes such student-generated content in online courses for the benefit of future learners, offering a scalable method to create high-quality assessments while deepening students' understanding of the material. The unique insights from students, who may spot nuances overlooked by experts, are invaluable due to their direct engagement with the course content. Systems such as PeerWise [2], RiPPLE [5], Upgrade [16], CodeTree [4], and ReadingQuizMaker [9] have utilized learnersourcing by enabling students to create, assess, and use peer-generated educational materials. This approach has expanded beyond just crafting new questions; it now includes the creation of various assessments, like programming exercises and instructional videos [4, 8, 13].

The interest in learnersourcing at the Learning @ Scale conference has primarily grown over the past nine years, showcasing the development of new techniques to enhance the quality of student-generated content. The advent of large language models (LLMs) offers fresh opportunities for this field and simplifies previous challenges in learnersourcing. For instance, recent work has explored involving students in improving LLM-generated educational content [17] and using LLMs to support students by lowering the complexity of learnersourcing tasks [3]. This workshop aims to introduce the concept of learnersourcing to a wider audience, demonstrating how to utilize or develop activities that empower students to generate and evaluate educational content. We will share tools, activities, and datasets from thousands of students to help instructors and researchers leverage learnersourcing interventions to improve student learning and enhance courseware. Such initiatives provide rich data for creating question banks, understanding student learning, and supporting natural language processing research. Our goal is for participants to discover the potential of learnersourcing, gain practical experience with related tools and activities, and explore how advances in AI creates ample opportunities in this space. We aim to attract those interested in expanding the creation of educational content and using online learning platforms.

2 Opportunities and Challenges

Learnersourcing presents a wealth of opportunities to enhance student learning and advance the field through innovative approaches for the creation, evaluation, utilization, and oversight of educational content [6]. These four pillars of the learnersourcing process can be seen in Figure 1. One such opportunity lies in facilitation of high-quality student-generated content, utilizing the recent advances of LLMs like Gemini 2.0 and o3 to assist students in the development of educational materials [3]. These advances help overcome previous challenges, such as leveraging LLMs to address the cold start problem in learnersourcing. Students can also collaborate with LLMs to develop and refine assessment questions and learning content explanations, engaging in critical thinking as they assess the model's suggestions. Such collaborations, when designed intentionally by prioritizing student learning, engage learners in scaffolded interactions with AI [18], and prepare them for a

future workforce where effective collaboration with AI will be a critical skill. Previous work has explored using NLP, trust-based networks, and deep learning to assist students in the evaluation of educational content, highlighting the critical role of human input alongside AI [1]. Balancing the innovative use of LLMs with careful human oversight promises to improve learnersourcing techniques and outcomes, making educational content more accessible, diverse, and effective for learners.
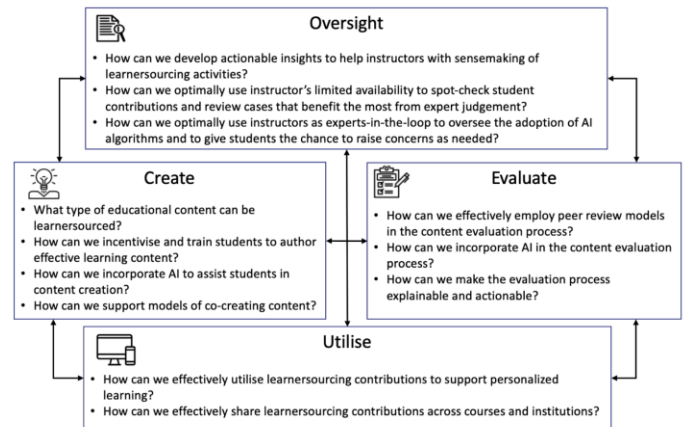

Figure 1: Challenges and opportunities relating to the four key aspects of learnersourcing that involve students, instructors, and AI.

Learnersourcing faces several challenges in the four facets of creating, evaluating, utilizing, and overseeing student-generated content. One significant hurdle is low participation rates in learnersourcing activities, especially when they are optional, with a tendency for a small portion of students to contribute most of the content [10]. Additionally, the process of having students and AI review and revise peer-generated questions introduces another layer of complexity. This AI integration into the learnersourcing process introduces additional biases inherent in LLMs [17]. The risk of perpetuating these biases in educational content underscores the importance of a human-in-the-loop approach, where both students and instructors play a crucial role in content evaluation and moderation. Recent findings suggest that student-authored MCQs can match the quality of those created by experts, highlighting the potential of learnersourcing [14]. Yet, the challenge of ensuring high-quality contributions from all students across any domain persists.

3 Pre-Workshop Plans

Prior to the workshop, we will grant participants early access to various learnersourcing systems created by our organizing committee and provide a curated list of significant learnersourcing research papers on the workshop's website. This pre-workshop access is designed to help participants understand the functionalities and evaluation methods of each system and engage with the learnersourcing process. While optional, we wanted to make these resources available, particularly for

participants that might not be familiar with this field. Additionally, we will distribute a short survey to gather details about participants' backgrounds and interests. The feedback from this survey will enable us to tailor the workshop content and activities to better match the preferences and needs of our attendees, aiming to create an engaging and relevant experience for all, regardless of prior experience in learnersourcing.

4 Workshop Structure

The inaugural version of our workshop took place at the 2022 Learning @ Scale conference as a half-day event, drawing over 30 participants and featuring 6 accepted paper presentations [11]. Our second iteration of the workshop in 2024 was just as successful, also drawing over 30 participants and having 6 accepted papers. Encouraged by this success and having since expanded our organization team and community engagement, we anticipate closer to 40 participants in attendance for this year's edition. This year's workshop will include a mix of brief presentations, interactive activities, and round-table discussions focused on learnersourcing. While we have outlined a provisional schedule below, please note that certain sessions, such as invited speaker talks and paper presentations, will be interspersed throughout the day to maintain a dynamic and engaging agenda:

- **Opening remarks**: Introductions of workshop organizers and participants, and a background to the focus of the workshop.
- **Invited Guest Speakers**: Two leading researchers in the field, not on the organizing committee, will give a presentation throughout the day regarding the use of AI in the learnersourcing process.
- **Short Presentations**: Authors of accepted submissions present their work which would be followed by Q&A.
- **Round-table discussion**: Participants will rotate among discussions on topics related to learnersourcing, focusing on creation, evaluation, use, and management of content.
- **Hands-on activity**: Participants will collaborate to brainstorm and develop new learnersourcing activities and contribute to a learnersourcing literature review.
- **Open discussion**: A facilitator will guide a collective discussion among all participants, summarizing insights from the round-table discussions, building consensus through the collaborative creation of shared notes.
- **Concluding remarks**: Closing remarks on the workshop will be made with future steps and opportunities for collaboration between participants, to maintain community engagement.

The primary aim of this workshop is to investigate the collaborative potential of students, instructors, and AI in generating, evaluating, and utilizing educational content through learnersourcing. We welcome participants from diverse fields, not limited to the usual learning sciences, machine learning, natural language processing, and learning analytics, believing that a variety of backgrounds enriches our exploration. With learnersourcing being a multifaceted domain involving students, instructors, and researchers, diverse perspectives are crucial for

identifying valuable future content, improving content quality assessment, and fostering collaboration among attendees.

5 Post-Workshop Plans

After the workshop, we will publish the accepted submissions via a CEUR workshop proceedings, which was done from the previous workshop iterations [12]. We hope the workshop encourages participants to adopt learnersourcing tools, use the discussed datasets, or start their own learnersourcing projects. Participants will be invited to join a dedicated Slack channel and mailing list, which has also been maintained since the first workshop iteration, providing continuous access to learnersourcing advancements and datasets. These resources are expected to support participants' future research and potentially inspire future Learning @ Scale workshops or competitions. We want to continue to cultivate a community of active collaboration and engagement in learnersourcing.

6 Call for Participation

Join our full-day workshop on learnersourcing! Featuring invited speakers, interactive activities, paper presentations, and discussions. Attendees will engage in hands-on development of learnersourcing activities suited to their own use cases and gain access to learnersourcing systems and datasets for exploration.

This workshop aims to foster discussions on new types of learnersourcing activities, strategies for evaluating the quality of student-generated content, the integration of LLMs with the field, and approaches to scaling learnersourcing to produce valuable instructional and assessment materials. We welcome participants from all backgrounds, as learnersourcing benefits from the diverse perspectives of students, instructors, researchers, and practitioners.

While no submission is required to participate in this workshop, we encourage attendees to share their insights and experiences with learnersourcing through 4-6 page submissions, including research papers, work-in-progress reports, and position papers. Submissions might cover topics such as:

- Strategies for engaging and motivating student participation in learnersourcing activities
- Exploration of innovative learnersourcing content formats
- Methods for evaluating the quality of student- and AI-generated content
- Incentivizing high-quality student contributions
- Techniques for providing actionable feedback during the learnersourcing process
- Enabling collaboration and content sharing across institutions
- Training students to develop high-quality resources and critically evaluate AI-generated content
- Exploring models of co-creating content in human-human and human-AI collaborations
- How LLMs can assist in the cold start problem for student-content creation

- Leveraging LLMs to assist in the different stages (creation, evaluation, aggregation, etc.) of the learnersourcing process
- Guidelines to maintaining pedagogical value in learnersourcing when AI assistance is involved
- Considerations for integrating student and AI contributions in learnersourcing, balancing the dual objectives of improving learning outcomes and enhancing the efficiency of educational content generation

This workshop is an opportunity to shape the field, share insights, and collaborate with peers dedicated to enhancing education through learnersourcing.

7 Organizers

Steven Moore is a postdoctoral researcher in Human-Computer Interaction at Carnegie Mellon University and is advised by Dr. Ken Koedinger. His research is focused on engaging students in the learnersourcing process and finding ways to assess the quality of their contributions.

Anjali Singh is a PhD candidate at the University of Michigan and is advised by Dr. Christopher Brooks and Dr. Xu Wang. Her research focuses on the use of human and machine intelligence to improve Data Science education, in both formal and informal learning settings like MOOCs.

Xinyi Lu is a PhD student at the University of Michigan advised by Dr. Xu Wang. Her research focuses on learning about how instructors perceive and interact with AI tools and building effective Human-AI tools to elaborate AI and CS techniques into teaching practice.

Hyoungwook Jin is an MS candidate in the School of Computing at KAIST, advised by Dr. Juho Kim. His research supports personalized learning environments at scale, leveraging AI agents as tutors and tutees to build personalized and adaptive interactive learning systems.

Hassan Khosravi is an Associate Professor at the University of Queensland's Institute for Teaching and Learning Innovation and an Affiliate Academic in the School of Information Technology and Electrical Engineering. He leads learnersourcing research and the development of the RiPPLE system.

Paul Denny is an Associate Professor in Computer Science at the University of Auckland, New Zealand. He leads the PeerWise project, which hosts more than six million practice questions, with associated solutions and explanations, created by students from 90 countries.

Christopher Brooks is an Associate Professor at the University of Michigan, is an applied computer scientist focused on educational technologies in higher education and informal learning. He led learnersourcing initiatives on Coursera, exploring student choice in generating questions.

Xu Wang is an Assistant Professor in Computer Science and Engineering and the School of Information at the University of Michigan. She develops human-AI collaborative techniques to support education, such as empowering educators to create effective learning experiences.

Juho Kim is an Associate Professor in the School of Computing at KAIST. His research focuses on building interactive and intelligent systems that support interaction at scale, aiming to improve the ways people learn, collaborate, discuss, make decisions, and take action online.

John Stamper is an Associate Professor at the Human-Computer Interaction Institute at Carnegie Mellon University. His work involves leveraging educational data mining techniques and the creation of data tools that can be used with learnersourcing data.

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