Towards AI-Assisted Multiple Choice Question Generation and Quality Evaluation at Scale: Aligning with Bloom's Taxonomy

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Introduction

- 1) Multiple Choice Questions (MCQs) are used in education for fast grading and feedback.
- 2) No current way to automatically generate MCQs based on Bloom's Taxonomy [1] (a taxonomy for the cognitive levels required to answer questions)
- 3) We automatically generate MCQs based on Bloom's Taxonomy using GPT-3.5.
- 4) We evaluate generated MCQs using an automated rules-based approach and a domain expert.
- 5) We assess the alignment of the taxonomy specified in the prompt (GPT-taxonomy) with the CNN-classified taxonomy (ML-taxonomy) and human classified taxonomy (human taxonomy), and the quality of questions based on their prompt-specified taxonomy.

Methods

Generation:

- 1) Zero-shot prompted (provided no training data) GPT-3.5 to generate MCQs based on textbook excerpts from OpenStax Chemistry 2e and Biology 2e [4]. Prompt template available in QR code.
- 2) Generated 5 MCQs per every Bloom level, for every excerpt. 120 questions were generated using excerpts from two randomly selected chapters from each textbook.

Evaluation:

- 1) Used a convolutional neural network (CNN) to verify the Bloom's taxonomy of generated MCQs. Model architecture based on [2].
- 2) Used the model in [3] to assess the quality of generated questions using IWFs (mistakes in educational MCQs). > 1 IWF = Low Quality, as per [5].
- 3) Domain expert with 28+ years of STEM educational experience was asked whether they would use a question in a classroom setting and the Bloom's taxonomy the question for

Fig 1 Subplot A (agrmt. between prompt taxonomy and CNN-classified taxonomy) suggests that GPT-3.5 has the capability for generating MCQs based on Bloom's taxonomy: strongest for Evaluation and weakest for Synthesis and Comprehension.

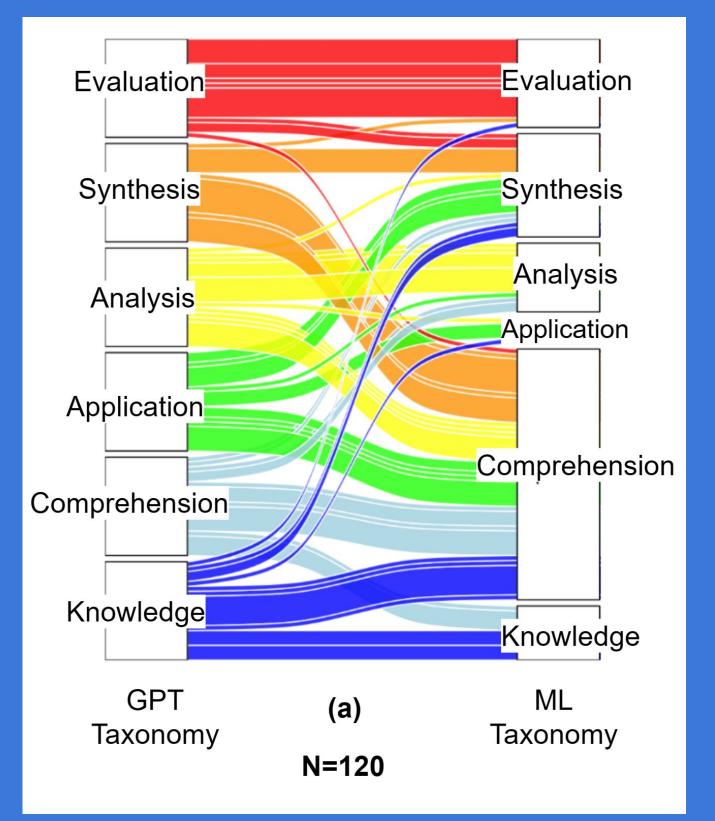
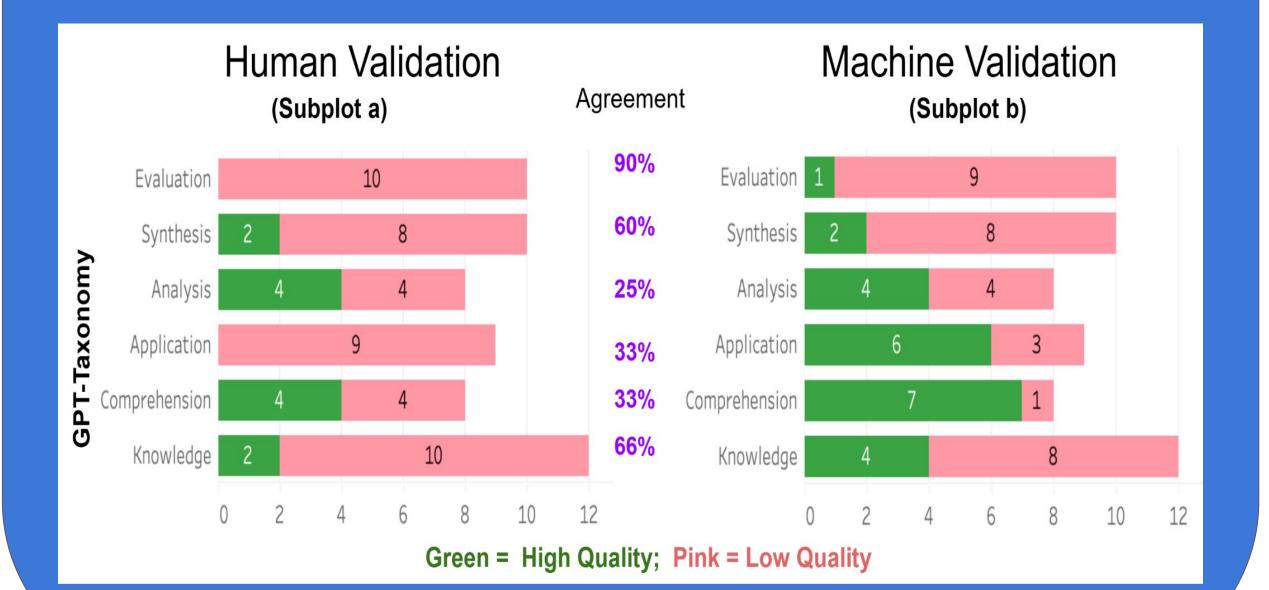
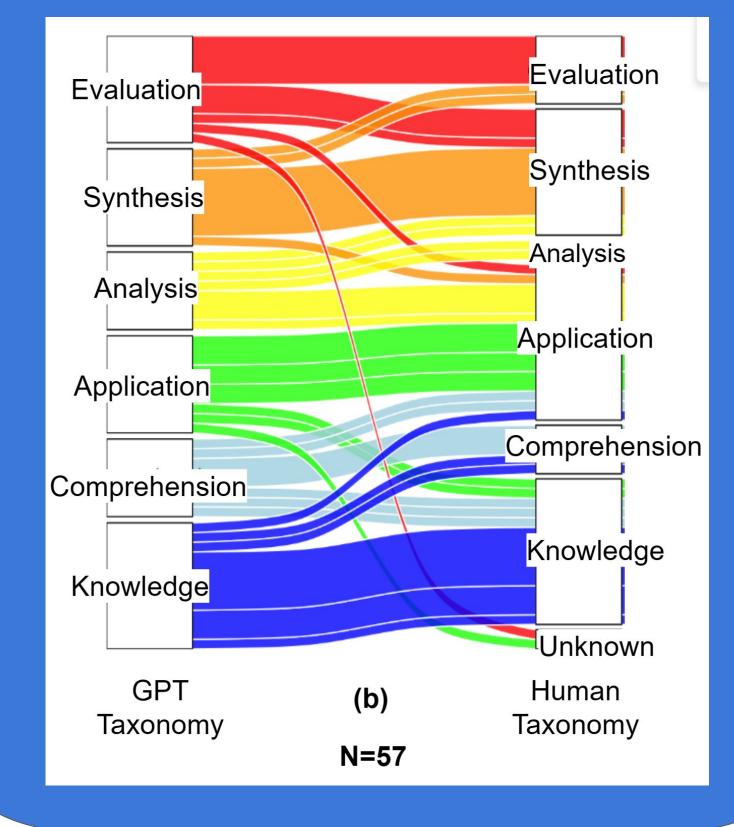


Fig 1 Subplot B (agrmt. between prompt taxonomy and human-classified taxonomy) suggests that there is room for improvement in generating questions for higher levels of Bloom's Taxonomy (Analysis, Synthesis, and Evaluation) each question in a subset of 57 of the MCQs.

Figure 2 shows the # of questions marked as high/low quality for each Bloom level and the percent agreement for each Bloom level.

- 1) In general, cognitive complexity and generated question quality show an inverse relationship.
- 2) In general, agreement between humans and machines decreases as cognitive complexity goes down.
- 3) Domain expert marked 21% of questions as high quality and model marked 42% of questions as high quality, showing difference of perspective.









Take Home Messages

- GPT-3.5 has promising capabilities in generating MCQs aligning with Bloom's Taxonomy, but aligning better requires further exploration.
 Particularly at higher levels of Bloom's taxonomy, GPT-3.5 struggles with generating questions that are usable in the classroom, more work needs to be done in refining the quality of questions.
- 3) Automated question evaluation using IWF and human evaluation differs significantly, requiring exploration on how to best model human validation using automated measures.

References

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