



RESEARCH ARTICLE

Evaluating the Use of Large Language Models vs. Traditional Textbooks in Physiology Education: Insights into Generation Z Medical Students' Preferences

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ABSTRACT

Background: Medical education is rapidly evolving with the integration of artificial intelligence [AI] and digital technology. Large language models [LLMs] such as ChatGPT, Bing, Bard, Gemini, Claude, and Perplexity are increasingly used by medical students primarily for their accessibility and ease of use. However, the abundance of information generated by these tools may be overwhelming and can complicate the independent assimilation of knowledge.

Objectives: This experimental study aimed to evaluate the effectiveness and acceptability of LLM based AI tools compared with traditional textbooks for learning Physiology among Generation Z medical students.

Methods: First-year medical students were randomized in to 2 groups: 1- standard Physiology textbooks and 2- Large language models based AI tools. After one hour of self directed study, students completed a 15 item multiple choice questions [topic specific] via Google form. The Medical Artificial Intelligence Readiness Scale for Medical Students was used to evaluate AI readiness and perceptions towards AI-assisted learning.

Results: Analysis revealed no statistically significant difference in assessment scores between students who used traditional textbooks and those who used large language model based artificial intelligence tools across the selected Physiology topics. Qualitative feedback indicated that many students favored combining both resources rather than relying exclusively on a single method.

Conclusion: Although large language models are gaining prominence in medical education, this study did not demonstrate a performance advantage over traditional textbooks in learning Physiology among Generation Z medical students highlighting a shift in learning preferences among Generation Z medical students. Future research is warranted to explore the role of large language models based artificial intelligence tools roles in medical education, with particular attention to global content validation, learning retention and ethical consideration.

Keywords: AI readiness, Medical Education, Generation Z, learning preferences.

Introduction

Medical education is already undergoing a generational transformation, particularly with increasing integration of artificial intelligence and digital technologies. Among these innovations, large language models, such as Chat GPT, Bing, Bard, Gemini, Claude, Perplexity, etc, have emerged as supplementary educational tools, functioning as virtual tutors for medical students education as tutors¹. These tools may serve as adjuncts to conventional learning methods, standard curriculum, and educational practices^{2,3}. Undergraduate medical education represents a critical phase in the development of future healthcare professionals. Although teacher-led instructions remain a cornerstone of medical curriculum, student centered led learning approached are gaining prominence due to their flexibility and higher engagement⁴. Medical students are increasingly inclined to use AI based tools to acquire knowledge and explore the educational contents rather than relying solely on traditional textbooks. This preference may be attributed to widespread availability, accessibility, rapid generation of responses, ready-to-use answers, related explanations, and existing resources on the internet with less effort offered by LLMs. AI driven tools & algorithms can help students save time, provide access to extensive information in the online world, and enable independent learning without the direct involvement of instructors⁵. However, these platforms often provides abundant data of information from the available resources, which can be overwhelming and challenging for students to critically evaluate, synthesize and retain. Consequently, the overall impact of AI assisted tools on student comprehension, learning outcomes and long term attention remains insufficiently understood and the entire learning and retention process using these tools is less known^{6,7}. Artificial intelligence tools have established their important role across multiple academic disciplines⁸. The application of artificial intelligence in medical education and training offers numerous potential, particularly in facilitation personalized learning experiences⁹. Research indicates that when used thoughtfully and appropriate prompting, AI tools can support students in studying and preparing for examinations. Additionally, AI can provide personalized academic support while adhering to educational standards¹⁰. AI is also extensively utilized in many clinical research and Physiology systems, such as an ECG analysis and evaluation from a large number of data sets to anticipate and predict disease conditions¹¹.

The scientific research studies directly comparing the freely available large language models with reference standard textbooks in Physiology learning remain limited, particularly in context of undergraduate Physiology education. Although LLMs shows promise in reinforcing clinical knowledge reinforcement, their effectiveness may be enhanced by rethinking AI development to ensure easier, safer, and more equitable to use. Nevertheless, medicine remain a particularly intricate domain for the application of LLMs¹². Several studies have demonstrated that chat GPT can successfully pass medical examinations; a recent study by Subramani M et al.,¹³ reported that ChatGPT

scored more than 75% marks in an University examination that consists of MCQs from Physiology. While AI has the potential to reshape and improve medical systems, there remains a risk of AI generated content and automation bias if not adequately validated by clinicians and academicians.

Traditional textbook based learning has been the cornerstone of medical education for decades. Textbooks offer a comprehensive and structured approach to knowledge acquisition and promote deep learning, they often lack interactivity and real-time updates. Medical students continue to rely on textbooks for foundational knowledge, but may perceive them monotonous and static compared to dynamic and interactive nature of LLM based AI tools. Studies have shown that although medical students possess relatively strong technical skills and their practical usability related to AI are significantly lower¹⁴. Despite the availability of educational AI resources, most users report having little or no formal AI training and rely primarily on self-directed learning. Incorporating LLM based AI tools awareness programs and certificate courses, including training on Large language models, in to medical education may help to bridge this gap¹⁵. Recent advancements in AI driven educational technologies and tools encompassing various technological innovations to improve Physiology education have been utilized in medical education for current-generation students. However, the adoption of LLMs requires valid and essential background scientific evidence to recognize and address the concerns related to accuracy, bias, ethical implications and educations effectiveness of AI based LLMs¹⁶. To fully realize the potential of LLM based AI tools, it is necessary to ensure and understand the students and educators perceptions of their effectiveness. In addition, issues related to accessibility, effectiveness usability, and cultural acceptance must be examined to support widespread recommendation and adoption. Therefore, this study was designed to evaluate and understand the medical students and doctors perceptions and attitudes towards integration of LLM based AI tools in medicine is crucial and need of an hour¹⁷. With this background, the present study aimed to explore the effectiveness of large language models in Physiology education compared with traditional textbook learning among 1st MBBS students.

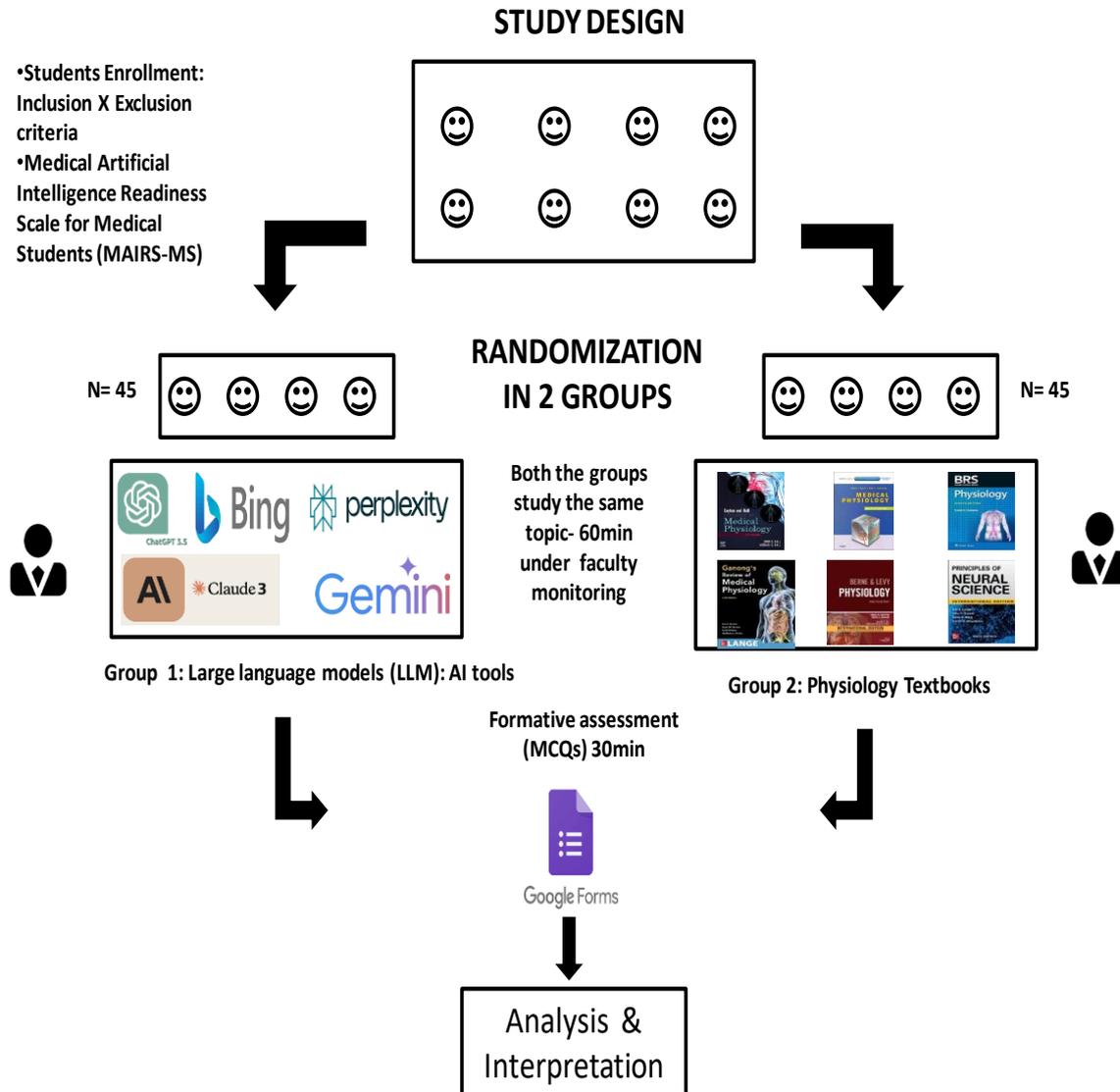
Materials and methods:

The outline of the study design is given in the study design flowchart [Figure:1]. This study was conducted in the Department of Physiology after obtaining institutional ethical approval. [M5/F142/2024] Medical students were enrolled in the study after detailed informed consent and scrutiny based on the inclusion and exclusion criteria. Students who joined 1st MBBS - Bachelor of Medicine, Bachelor of Surgery after their higher secondary schools were included in the study. Students who already had a graduate degree in Physiology or completed any course containing Physiology before joining MBBS were excluded. Students who did not participate in the questionnaire were also excluded. Upon enrollment, students meeting

the inclusion criteria were allocated to two groups through randomization using their roll numbers. The sample size was calculated using G power computer software. A total sample size of 45 in each group was

calculated to detect a significant effect [$d=0.45$]. The power of the test was fixed at 90% and alpha at 0.05¹.

Figure 1: Flow chart- Study Design:



Group 1- This group of students used standard textbooks to read Physiology [Authors by: Guyton, Ganong, Boron, Costanzo, Berne & Levy, Principles of neural Science by Eric Kandel], Group 2- This group of students used LLM [ChatGPT, Google Gemini, Microsoft Bing, Claude AI, Perplexity] to study different topics of Physiology. Two main aspects, cognitive and affective, were assessed in this study. The cognitive part included studying the topic, followed by the MCQ test. The practical part included an assessment questionnaire and interview questions in google forms regarding the study and its methodology Table 3 & 4. Important topics related to each physiological system were chosen for the study. The topic was disclosed to the participants only on the day of the study. Moreover, only topics that were new or not already covered during routine lectures were selected for this study. The study was conducted in a practical lab with seating arrangements. The Textbook group and the LLM group were placed in separate rooms, and both were supervised by faculty and staff. The students were given 60 minutes to read about the topic, and immediate recall was assessed

using 15 MCQs using google forms, and the time given was 30 minutes. During the entire session, the departmental faculty and residents monitored the students. The experts in Physiology internally and externally validated the MCQs used for the assessment. The marks obtained in the MCQs were announced to the students, and a review session was conducted before the next topic. The marks were not considered for their internal assessment.

The medical artificial intelligence readiness scale questionnaire for medical students [MAIRS-MS] was administered to evaluate and monitor perceived readiness levels among medical students on AI technologies and applications. The MAIRS-MS consists of four main aspects- cognition [1-8 questions], vision [17-19 questions], ability [9 -16 questions], and ethics [20-22 questions]¹⁸. Following the MCQ test, the student participants were asked five questions, with responses rated on a 5-point Likert-type scale: 1—strongly disagree, 2—disagree, 3—neutral, 4—agree, and 5—strongly agree.

QUESTIONS:

1. How do you rate the effectiveness of the above study on your learning outcome?
2. How do you rate textbooks as a source of study for acquiring knowledge?
3. How do you rate AI tools as a source of study for acquiring knowledge?
4. Do you think this AI system supports learner–instructor interaction?
5. Which of the two learning tools, in your opinion, is likely to enhance your critical thinking skills? AI or Textbooks?

DATA ANALYSIS:

The responses were exported to Excel from the Google form and questionnaires, and a systematic analysis using SPSS was done. Statistical significance was determined with a threshold [alpha] level set at 0.05. The quantitative data obtained from surveys and

questionnaires were analyzed using statistical tools such as descriptive statistics and independent *t*-tests. The mean average scores obtained from all the variables were calculated and compared statistically.

Results:

A total of 90 first-year medical students participated in the study and were equally allocated to the large language model group and the textbook group. Comparison of post-learning assessment scores revealed no statistically significant difference between the two groups across the evaluated Physiology topics, including respiratory, cardiovascular, gastrointestinal, renal, special senses, and general Physiology systems ($p > 0.05$). The corresponding results are presented in Tables 1 to 4. Student feedback regarding the use of large language model–based artificial intelligence tools and textbooks is illustrated in Figures 2 and 3.

Figure 2: Students feedback about LLM: AI tools [AI] and Test book [T] usage.

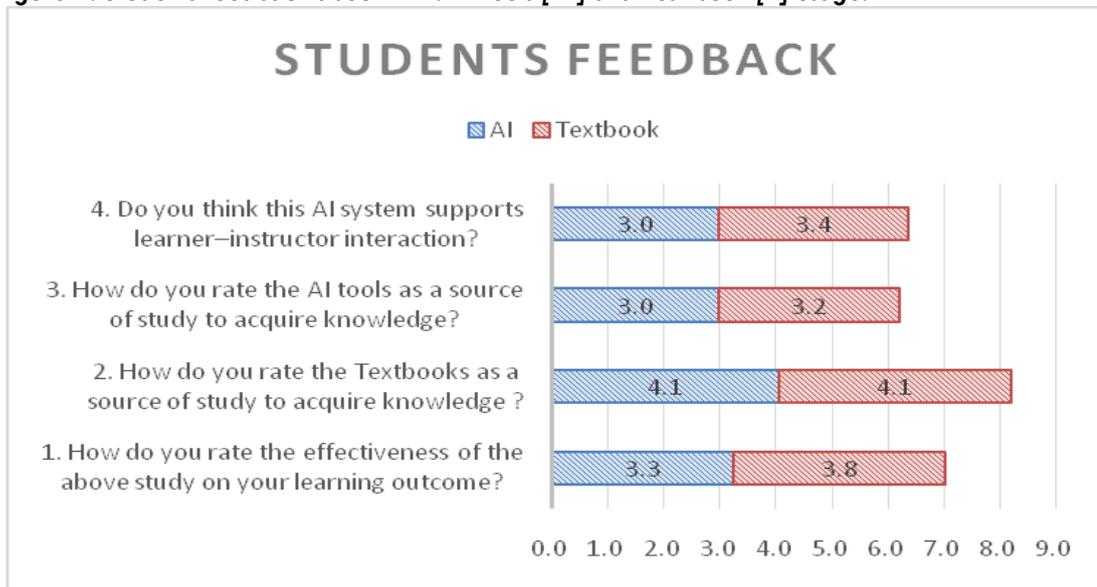
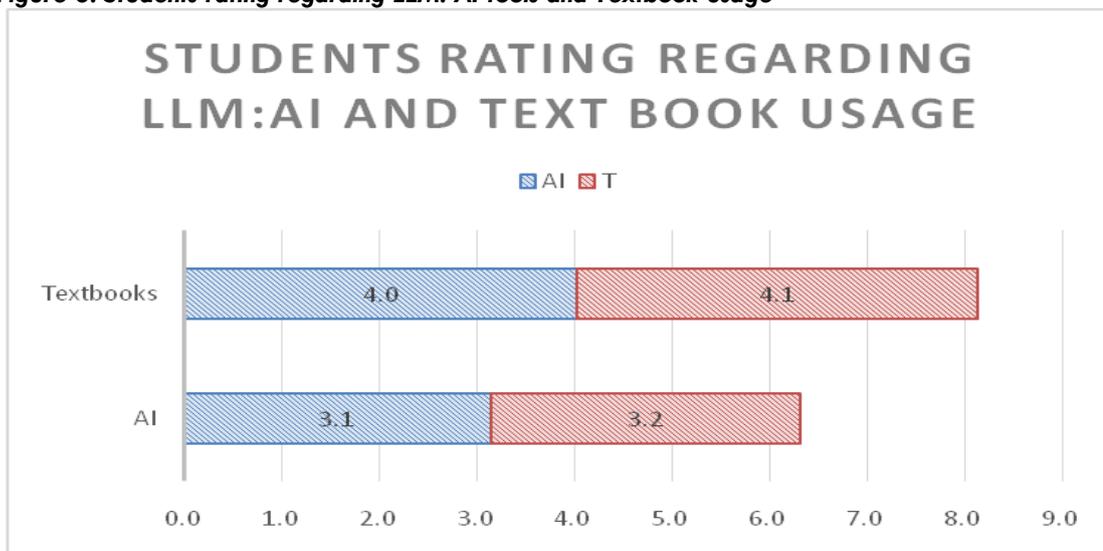


Figure 3: Students rating regarding LLM: AI tools and Textbook usage



Figures 2 and 3 represent the feedback and ratings regarding using LLM tools [AI] versus traditional Textbook [T] methods for learning Physiology compilation among Generation Z students.

Table 1: Comparison of scores between the AI group and text group on different topics

S. No	Topic/System	Group	Score [Mean \pm SD]	p-value
1	Respiratory System	LLM: AI tools	3.1 \pm 1.3	0.06
		Textbook	3.1 \pm 1.2	
2	Special senses	LLM: AI tools	2.01 \pm 2.52	0.58
		Textbook	1.53 \pm 1.99	
3	Cardiovascular System	LLM: AI tools	2.21 \pm 2.74	0.86
		Textbook	1.34 \pm 1.82	
4	General Physiology	LLM: AI tools	1.42 \pm 1.67	0.29
		Textbook	1.65 \pm 2	
5	Gastrointestinal System	LLM: AI tools	1.66 \pm 2.04	0.10
		Textbook	1.33 \pm 1.92	
6	Renal System	LLM: AI tools	1.98 \pm 2.38	0.12
		Textbook	1.83 \pm 2.2	

Table 1. Comparison of mean \pm SD scores between the large language model group and the textbook group across selected Physiology topics. Data were analyzed using an independent t-test, with $p < 0.05$ considered statistically significant. No statistically significant differences were observed between groups.

Table 2: Comparison of overall mean scores between the AI group and Textbook group

S. No	Group	Score [Mean \pm SD]	p-value
1	LLM: AI tools	6.3 \pm 2.9	0.81
	Textbook	6.7 \pm 2.6	

Table 2 provides the overall mean scores between the LLM tools and the textbook group, which showed no significant distinctions.

Table 3: Distribution of responses to the Medical Artificial Intelligence Readiness Scale (MAIRS) across four domains (ability, cognition, vision, and ethics)

Tables 3 and 4. Medical Artificial Intelligence Readiness Scale for Medical Students scores across four domains. MAIRS is a validated instrument with a Cronbach's alpha of 0.87 used to assess readiness for artificial intelligence among medical students. Most responses fell within the neutral and agree categories, with higher scores observed for ability and cognition compared with vision and ethics.

S. No	Parameters	Strongly Disagree n [%]	Disagree n [%]	Neutral n [%]	Agree n [%]	Strongly Agree n [%]
1. 1	I can define the basic concepts of data science	4[4.7]	15 [17.4]	35[40.7]	30.2[26]	8 [7]
2. 2	I can define the basic concepts of statistics	5 [5.8]	11 [12.8]	34 [39.5]	28 [32.6]	8 [9.3]
3. 3	I can explain how AI systems are trained	11 [12.8]	31 [36]	25 [29.1]	15 [17.4]	4 [4.7]
4. 4	I can define the basic concepts and terminology of AI	5 [5.8]	16 [18.6]	44 [51.2]	19 [22.1]	2 [2.3]
5. 5	I can properly analyze the data obtained by AI in healthcare.	3 [3.5]	18 [20.9]	33 [38.4]	28 [32.6]	4 [4.7]
6. 6	I can differentiate the functions and features of AI-related tools and applications.	4 [4.7]	15 [17.4]	39 [45.3]	23 [26.7]	5 [5.8]
7. 7	I can organize workflows compatible with AI.	3 [3.5]	20 [23.3]	28 [32.6]	29 [33.7]	6 [7]
8. 8	I can express the importance of data collection, analysis, evaluation, and safety in developing AI in healthcare.	3 [3.5]	7 [8.1]	35 [40.7]	28 [32.6]	13 [15.1]
9. 9	I can harness AI-based information combined with my professional knowledge.	2 [2.3]	6 [7]	33 [38.4]	36 [41.9]	9 [10.5]
0. 10	I can use AI technologies effectively and efficiently in healthcare delivery.	1 [1.2]	7 [8.1]	38 [44.2]	28 [32.6]	12 [14]

S. No	Parameters	Strongly Disagree n [%]	Disagree n [%]	Neutral n [%]	Agree n [%]	Strongly Agree n [%]
1. 11	I can use artificial intelligence applications for their purpose.	1 [1.2]	2 [2.3]	34 [39.5]	40 [46.5]	9 [10.5]
2. 12	I can access, evaluate, use, share and create new knowledge using information and communication technologies.	2 [2.3]	8 [9.3]	33 [38.4]	35 [40.7]	8 [9.3]
3. 13	I can explain how AI applications offer a solution to which problem in healthcare.	2 [1.2]	12 [14]	37 [43]	29 [33.7]	7 [8.1]
4. 14	It is valuable to use AI for education, service and research purposes.	1 [1.2]	2 [2.3]	26 [30.2]	43 [50]	14 [16.3]
5. 15	I can explain the AI applications used in healthcare services to the patient.	1 [1.2]	8 [9.3]	35 [40.7]	33 [38.4]	9 [10.5]
6. 16	I can choose proper AI applications for the problems encountered in healthcare.	4 [4.7]	9 [10.5]	41 [47.7]	25 [29.1]	7 [8.1]
7. 17	I can explain the limitations of AI technology.	1 [1.2]	7 [8.1]	34 [39.5]	34 [39.5]	10 [11.6]
8. 18	I can explain the strengths and weaknesses of AI technology.	2 [2.3]	5 [5.8]	38 [44.2]	32 [37.2]	9 [10.5]
9. 19	I can foresee the opportunities and threats that AI technology can create.	2 [2.3]	7 [8.1]	34 [39.5]	32 [37.2]	11 [12.8]
10. 20	I can use health data by legal and ethical norms.	2 [2.3]	4 [4.7]	44 [51.2]	29 [33.7]	7 [8.1]
11. 21	I can conduct my work under ethical principles while using AI technologies.	2 [2.3]	7 [8.1]	37 [43]	36 [41.9]	4 [4.7]
12. 22	I can follow legal regulations regarding the use of AI technologies in healthcare.	2 [2.3]	6 [7]	38 [44.2]	27 [31.4]	13 [15.1]

Table 4 represents the overall mean score and the proportion of the study population who scored more than 50 %.

S. No	Parameter	Score mean \pm SD	n [%]
1.	Cognition score	24.9 \pm 5.5	73 [85]
2.	Ability factor	27.9 \pm 5.1	84 [98]
3.	Vision factor	10.5 \pm 2.2	80 [93]
4.	Ethics factor	10.2 \pm 2.1	80 [93]
5.	MAIRS points	73.6 \pm 12.9	83 [97]

Following are the few comments from the students after participation in this study:

- To enhance our understanding, we must combine the tools and textbooks.
- Textbook is better
- AI is a great way to facilitate the process of learning and memorizing,
- Both textbooks and AI should be used together to gain a better understanding of the subject
- A teacher's explanation is always better than reading books or AI.
- AI can't be a primary source for acquiring knowledge
- Integrated use of AI and books is a good way to learn.
- Combining AI and textbooks is a better option than AI or textbooks alone.
- Some more time should have been provided for Studying.
- Textbooks give us limited and essential information that is vital
- Textbooks have more authentic and relevant content than AI
- AI cannot be the primary source of knowledge acquisition. However, it can be a supplementary resource along with the Textbook...
- Both are important for gaining the required knowledge and information. Anyone should replace none.
- Textbooks have complicated language and are very difficult to understand, but AI provides the easiest way to learn
- AI tends to mislead information and not necessarily understand the concept, but standard books generally lack significant errors.
- Both are effective for students.
- AI tools are not for beginners, but the Textbook provides information from the basics.
- Books are better for acquiring knowledge and cannot depend on AI
- I felt There was a shortage of resources and study materials while using the AI mode of studying
- AI is better
- AI can complement the traditional learning method. However, AI cannot be the sole means of learning, especially if the student is involved in a professional

course. Indeed, under the proper guidance of the instructors, AI will benefit the learning process.

22. Textbooks are effective but require more time to cover such vast topics
23. I do not know how to use AI, so it may not be fruitful to use AI for me
24. Textbooks are better than AI tools in matters of study purpose
25. Textbooks are better than AI

Discussion:

This study explored the effectiveness of conventional learning. Using textbook versus large language models: AI tools among generation Z first year medical students. Previous data on this topic is limited, especially the original articles involving medical students are scarce. Most of the studies have explored the role of large language models, mainly ChatGPT. These tools were mainly used to generate and solve the MCQs-based exams/tests, and the performance was compared with actual students^{19–22}. A recent study states that ChatGPT could potentially revolutionize medical education settings by assisting medical students and faculty, offering innovation within the medical sciences and education framework. This interesting opinion could be due to the intriguing nature of ChatGPT or any large language models that create rapid responses and conversations, such as interactions within seconds. In another study, it was reported that Claude-2 showed better proficiency than ChatGPT-3.5 in conceptual MCQ explanation^{23,24}. The novelty of the present study is its unique study design, which did not assess the language model's utility to prepare MCQs or question papers; instead, this study was used to evaluate the efficacy of LLM: AI tools in transferring knowledge to medical students. Two groups were assigned to study Physiology topics, one of the significant subjects in 1st year MBBS, using either textbooks or LLMs and their performance was assessed. The findings shed light on several critical aspects of this comparative educational approach in Physiology learning and the concerns regarding using LLM as an educational tool²⁵. The study also assessed the students attitudes towards these tools.

Despite the predominant learning methodology among Generation Z students centered around technology and videos²⁶, Gen Z participants being optimistic about LLM: AI's benefits for enhanced productivity and personalized learning, the results show no significant difference among the scores between LLM and textbook group. This indicates that there was no added benefit while learning using either tool. This is even though language models provide topic-specific searching options, explanations and suggestions, allowing students to acquire the required information in less time and effort²⁷ compared to conventional textbooks. It is also worth noting that the feedback and ratings compiled from Generation Z students regarding using LLM tools versus traditional textbook methods for learning Physiology credited both approaches equally [Table 1 & 2].

A recent study says that ChatGPT and Bard are extremely useful because they can imitate human conversation, engage students in interactive discussions,

and offer helpful information^{28,29}. The other possible reason for the students who preferred LLM could be the complex languages and new words used in medical textbooks that demand a high level of comprehensive understanding and critical thinking³⁰.

It has been found that the perceived effectiveness of tools employed by textbooks remains high among Generation Z students in medical education³¹. This could be the reason for the finding in most of the student's feedback stating that textbook content is validated, reliable and better for beginners when compared with LLM: AI tools. This perception may vary based on demographic factors and settings, with graphics often proving particularly efficacious. In medical education, the preference for traditional Textbook learning over newer technologies like large language models such as Chat GPT is often influenced by established academic norms, familiarity with traditional resources, and the perceived reliability of textbook information³². Various factors have been postulated to potentially impact medical students textbook preferences, including text clarity, language complexity, directness to the intended concept, flow of ideas, clinical/multidisciplinary integration, and effectiveness of aiding tools utilized within textbooks^{30,33}.

Regarding preference for various learning language models, Chat GPT received the highest rating, followed by Claude two and perplexity and Gemini/Bard. Any students did not use Microsoft Bing. Chat GPT has also been reported to have a higher score in solving hematology cases than Google Bard and Microsoft Bing³⁴. The LLMs, namely, Chat GPT, Google Bard, Microsoft Bing, and Perplexity, all showed >75% concordance in suggested statistical tests for research case vignettes, with all having acceptance of >95%. Another study regarding using AI language models for medical decision-making on complex scenarios found that Claude AI performs better than ChatGPT, Google Bard, and Perplexity AI³⁵. This suggests a wide variation in responses regarding preference and efficacy among different AI language models with no consistent finding. Interestingly, based on the feedback comments following the current study, most students reported that combining LLM: AI and textbook approach can enhance learning and understanding of any topic more effectively.

Gen X students and Gen Y teachers acknowledge these benefits but, at the same time, voice concerns about overreliance, ethical issues, and the need for guidelines to ensure the responsible use of AI in education³⁶. Excessive LLM dependency on answers and explanations may put the student's capacity for exploration and analysis at risk, which may reduce their innate nature²⁵. Additionally, medical students utilize such language tools to generate answers, and for the assignment, seminars may lead to misconduct³⁷ and require immediate attention. Further, students use it for written assignments and examinations, leading to concerns about AI-assisted cheating³⁸. Moreover, such electronic resources like LLM: AI tools may attenuate the student's exposure to critical thinking, which is crucial for a deep understanding of concepts in medicine.

In terms of AI acceptance and readiness among medical students, overall readiness scored above 50% across all domains of the MAIRS assessment. The highest mean scores were recorded in the areas of ability [27.9±5.1] and cognition [24.9±5.5], compared to lower scores in vision [10.5±2.2] and ethics [10.2±2.1]. These scores align with the majority of students selecting "agree" or "neutral" in their MAIRS responses [Table 3 & 4]. The ability domain evaluates a student's preparedness to choose, implement, and articulate the use of AI applications within the medical field. Vision assesses their capacity to analyze and communicate AI technologies limitations, advantages, and prospects. Ethics determines their commitment to adhering to legal and ethical guidelines in the application of AI ³⁹. AI readiness and acceptance showed varied effects across domains among medical students from various countries, according to the existing literature on MAIRS scores. However, the MAIRS scale's cut-off point has not yet been determined, and further research in diverse countries is needed for more effective cross-cultural comparisons ⁴⁰.

Large language models could have yet another application in the sense that these can be incorporated with machine learning tools that sync and automatically perceive or extract the student's stress data from wearables ⁴¹. This can assist in the prediction of anxiety and depression during exams ⁴². Language models can also be tuned to analyze individual student performance and generate personalized learning materials that address their specific areas of weakness ⁴³. Strict regulations, guidelines, and supervision must be formulated to prevent misinformation on LLM: AI tools in medical education and ensure optimum use that will benefit students, teachers, and society. ²

Limitations

This study involves data exclusively from first-year students at a single central institution, with efforts to ensure uniformity in student motivation and involvement across all groups. It is important to note that textbooks may not always provide the most up-to-date information on specific topics. The time allotted for learning was the same for students with different levels of prior knowledge. Future research should involve collaboration with multiple medical institutions and compare study methods across developed and developing countries to fully understand the role of large language models [LLMs] in medical education. The

study did not validate the content or address data privacy concerns related to using LLMs for Physiology, which included information from textbooks, articles, news, blogs, and other verified online sources. It is also crucial to recognize that while LLMs often provide highly accurate responses, they are unreliable. Recent advancements in LLMs and generative AI tools may produce different outputs for the same prompt, and information from ChatGPT may not be fully up-to-date beyond 2021, with several associated limitations ¹⁷.

Conclusion:

The study compared the effectiveness of traditional textbook learning vs large language models [LLMs: AI tools] in teaching Physiology to first-year Generation Z medical students. Most existing research focuses on LLM tools for generating multiple-choice questions [MCQs] and exam preparation rather than their broader use in medical education. Among various LLMs, ChatGPT shows the potential to revolutionize medical education with its fast, interactive responses, surpassing other models like Claude, Perplexity, Gemini, and Bing. However, the study found no significant preference for LLM over textbooks among students studying and learning physiology. Most students favoured a blended approach, combining both LLM and textbooks. Future research must focus on collaborative studies, both nationally and globally, to thoroughly explore the role of LLM tools in medical education. These studies should ensure robust content validation and address the ethical and educational challenges of LLM usage among medical students. Continuous research and validation of LLM tools will be essential for understanding their long-term impact and acceptance among medical students, faculty, and clinicians. It will provide valuable insights for effective resource allocation in education shortly.

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Conflicts of interest: Nil

Data availability: The data that support the findings of this study are not openly available due to reasons of sensitivity and are available from the corresponding author upon reasonable request.

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