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Systematic Review

# Immersive Extended Reality (I-XR) in Medical and Nursing Education: A Systematic Review and Pedagogical Directives

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**Abstract:** Simulation has evolved from basic practice to Immersive Extended Reality (I-XR). This systematic review examined 56 published studies on the impact of I-XR, including virtual reality (VR), augmented reality (AR), and mixed reality (MR), on the education of medical and nursing students, specifically their skill competency and knowledge acquisition. The results demonstrate the significant potential of I-XR in healthcare education, with 42.5% of VR studies, 42.9% of AR studies, and the single MR study also demonstrating greater improvements in clinical skills and knowledge acquisition compared to non-I-XR training conditions. In contrast, only 2.5% of VR studies and 7.14% of AR studies favored non-I-XR methods. It is important, however, to acknowledge the 26.8% of studies that showed mixed results (some evidence for the I-XR methods on some outcomes, but also some evidence for the non-immersive (non-I-XR) methods, on other outcomes). Notably, the review also identified a critical gap in the theoretical foundations of I-XR learning, highlighting an urgent need for research to inform the effective pedagogical implementation of these powerful tools.

**Keywords:** medical/healthcare education; conventional simulation-based learning; immersive extended reality(I-XR); learning theory

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## 1. Introduction

Simulation-based training is a pedagogical approach that employs simulated scenarios or environments to improve clinical skills, mainly among health field workers. The objective is to replicate real-world situations to provide a realistic and immersive experience without exposing patients to real risks (Becker & Hermosura, 2019). Simulation-based training complements conventional teaching methods, focusing on improving skill acquisition, decision-making, and teamwork (Ajemba et al., 2024). Such clinical experiences can help to create an opportunity for students to apply didactic content to clinical practice which allows students to bridge knowledge, skills, and competency (Hultquist & Bradshaw, 2016).

Halsted's "see one, do one, teach one" model, while groundbreaking in 1890, is now criticized for prioritizing rapid experience over deliberate skill development and patient safety (Ayub, 2022; Baskaran et al., 2023). This approach raises concerns about inadequate preparation, potentially leading to medical errors and a lack of emphasis on critical thinking and self-reflection (Rohrich, 2006). Modern medical education is shifting towards comprehensive training models that prioritize patient safety and incorporate simulation, deliberate practice, and feedback (Ayub, 2022; Baskaran et al., 2023).

Simulation-based medical education (SBME) (Saleem & Khan, 2023) offers a viable solution to bridge this gap. SBME enables deliberate practice and standardized training in a safe learning

environment. It incorporates a wide array of techniques, ranging from basic mannequin-based simulations and standardized actor-patients to complex multimedia scenarios (Shah et al., 2022; Ziv, 2003). Through SBME, students can build confidence and competence, and ultimately enhance patient safety and the overall quality of care.

SBME has three primary objectives: (1) the execution of clinical skills, (2) supervised practice of that skill, and ultimately, (3) the independent and confident performance of clinical skills. This educational approach extends beyond technical proficiency to encompass the development of exteroceptive awareness in aspiring professionals. Exteroceptive awareness, the ability to perceive and interpret external stimuli, is crucial for stress management, confidence building, and overall well-being (Baskaran et al., 2023). Specifically, cultivating exteroceptive awareness helps regulate the sympathetic nervous system and manage stress responses in high-pressure situations, such as surgical procedures, thereby enabling healthcare professionals to maintain composure and perform optimally.

Specifically, significant positive effects for SBME (vs. non-simulated methods) have been shown for theory knowledge, analytic skills, learning interest and understanding, satisfaction, cooperative ability, problem-solving ability, teaching success, and situation awareness (Su & Zeng, 2023). Additionally, SBME proves effective for assessing teamwork and communication among healthcare providers (Dodson et al., 2023; Sezgin & Bektas, 2023). These benefits translate into improved patient outcomes and reduced healthcare costs (Le Cook et al., 2013; Issenberg et al., 2005), establishing it as a fundamental pillar of healthcare clinical training (Saleem & Khan, 2023) that provides a safe and effective environment for learners to develop crucial skills and decision-making abilities (Pottle, 2023).

### 1.1 Integration of I-XR in Healthcare Education

Recent technological advancements have further propelled SBME to include methods like virtual reality (VR), augmented reality (AR), and mixed reality (MR) (Tang et al., 2016). These methods are thought to help better equip healthcare students with practical experience and readiness for actual scenarios and procedures (Horowitz et al., 2022; Wu & Norvell, 2022). Such technologies are increasingly seen as having promise for teaching in the healthcare professions, even though the largest market is still in the entertainment and the gaming industry (Bankar et al., 2023; Wohlgemant et al., 2022).

Extended reality (XR) is an umbrella term used to include VR, AR, and MR (Aebersold et al., 2020), but typically refers to computer-generated images in the wearer's field of vision and includes the range of the user's view of the world that can exist from fully visible to fully occluded (Zhang et al., 2023), and thus blurs the line between the digital world and the physical world (Chengoden et al., 2023). Finally, Immersive Extended Reality (I-XR) is characterized as a screen-based simulation that highlights the 3D nature of the simulated patients, graphics, sound, and navigation through the environment (McGrath et al., 2018). The use of I-XR for healthcare education is becoming increasingly more recognized as it allows an almost unlimited number of clinical scenarios to be simulated, with the ability to allow real-time feedback on student progress and patient status (Lu & Bowman, 2021), and can improve users' motivation, engagement and enjoyment in educational learning across different domains by providing powerful experiential learning (Fokides et al., 2021; Tonteri et al., 2023). As more healthcare educators pioneer innovative methods, I-XR is emerging at the forefront of teaching and learning and is preparing healthcare students with practical skills, hands-on experience, and preparedness for real-world scenarios and procedures (Horowitz et al., 2022). Embracing these changes represents innovative steps towards cultivating a more knowledgeable, proficient, and self-assured healthcare workforce, benefiting patients, and enhancing healthcare education.

### 1.2. Review of I-XR Studies for Training Effectiveness in Healthcare Education

Whether I-XR methodologies specifically improve student learning outcomes compared to non-immersive (non-I-XR) methods is still up for debate. Historically, the majority of meta-analytic studies focused on medical students' surgical skills (Zhang et al., 2023). Within the last five years, there has been a surge of systematic reviews and statistical meta-analyses seeking to explore whether I-XR (specifically VR, AR, MR) methodologies improve students' learning outcomes across domain-specific skills, procedural outcomes, and even non-technical skills (e.g., empathy, self-efficacy, teamwork). Despite the number of reviews, the results are still inconclusive and often dependent on how "knowledge acquisition" and "skill performance" are operationalized. For instance, a meta-analysis of 11 studies on VR endoscopy training for medical students found that while VR improved procedure completion and overall performance ratings compared to traditional methods, it resulted in fewer independent completions and showed no difference in other behavioral outcomes (Khan et al., 2018). Another VR systematic review of nine studies revealed that only two showed improved *knowledge acquisition* in healthcare professionals compared to control groups (Abbas et al., 2023), suggesting that while VR can enhance knowledge, it may not be consistently superior to other methods. However, the same review found significant increases in *skill performance* in 19 out of 21 studies comparing VR to control methodologies (Abbas et al., 2023), indicating a stronger impact on practical skills. Other recent reviews (Dicheva et al., 2023; Huai et al., 2024; Jallad & Işık, 2022; Uslu-Sahan et al., 2023) show positive effects of I-XR on nursing students' knowledge, skill performance, skill acquisition, and clinical reasoning. These varied results suggest that the effectiveness of I-XR may depend on factors such as the specific technology used, the type of training, and the outcome being measured. Further research is needed to fully understand the conditions under which I-XR can optimize learning outcomes in healthcare education.

The results are also mixed as to whether one kind of immersive technology (AR, VR, MR) is more effective than another. For example, a systematic review of VR head-mounted displays in medical education (Kovoov et al., 2021) showed significant increases in learning surgical *procedures* in seven of the 11 studies reviewed compared to other immersive methodologies, but no differences when directly compared with AR. Yet, when assessing anatomy *knowledge*, only one of the six studies reviewed showed increased anatomy knowledge using VR. The authors concluded that VR technologies outperformed conventional methods for learning surgical skills, but they were not superior for learning anatomy knowledge. Yet, in another meta-analysis, VR methodologies proved to be more effective in improving nursing students' performance skills compared to other mixed reality methodologies (Chen et al., 2023).

Despite the mixed evidence for I-XR technologies compared to non-immersive technologies in learning procedural skills and knowledge acquisition, the specific teaching and learning processes remain under-explored. Therefore, how these methods use or adhere to learning theories to understand the delivery of skills and knowledge could be another overlooked possibility for the mixed evidence.

For example, educational learning theories offer valuable frameworks for understanding how students acquire knowledge and develop skills. Researchers have identified thirteen key theoretical perspectives relevant to medical education, including cognitivism, constructivism, experiential learning, and reflective learning, among others (Kaufman & Mann, 2011; Mann, 2011). These theories guide a range of pedagogical approaches, from optimizing information presentation (Issa et al., 2013) to fostering reflective practice (Van Merriënboer & Sweller, 2010).

However, despite the potential benefits of integrating learning theories into I-XR research, several studies have identified a concerning trend. Many studies on I-XR in healthcare education either poorly described or completely omitted their pedagogical approaches (Kivuti-Bitoket al., 2022). In one study, less than 3% of papers on VR simulations incorporated a conceptual framework or theory in their design, and self-initiated training was the most common mode (Jiang & Fryer, 2024).

This lack of theoretical grounding raises concerns about the rigor and generalizability of I-XR research in healthcare education. Without a clear theoretical foundation, it becomes difficult to understand why certain interventions are effective and how they can be optimized for different

learning contexts. This highlights a critical need for researchers to explicitly integrate learning theories into the design, implementation, and evaluation of I-XR training programs.

### 1.3. Summary

In sum, I-XR is transforming healthcare education by providing immersive and engaging learning experiences that enhance skill acquisition, knowledge retention, and preparedness for clinical practice. This technology aligns with the evolving needs of healthcare training, offering adaptable and flexible solutions for integrating cutting-edge techniques. As technology continues to advance, I-XR is poised to play an increasingly vital role in shaping the future of healthcare education. However, realizing the full potential of I-XR depends on equipping healthcare educators with the necessary skills and resources to integrate these technologies into high-quality patient care training effectively.

### 1.4. Purpose

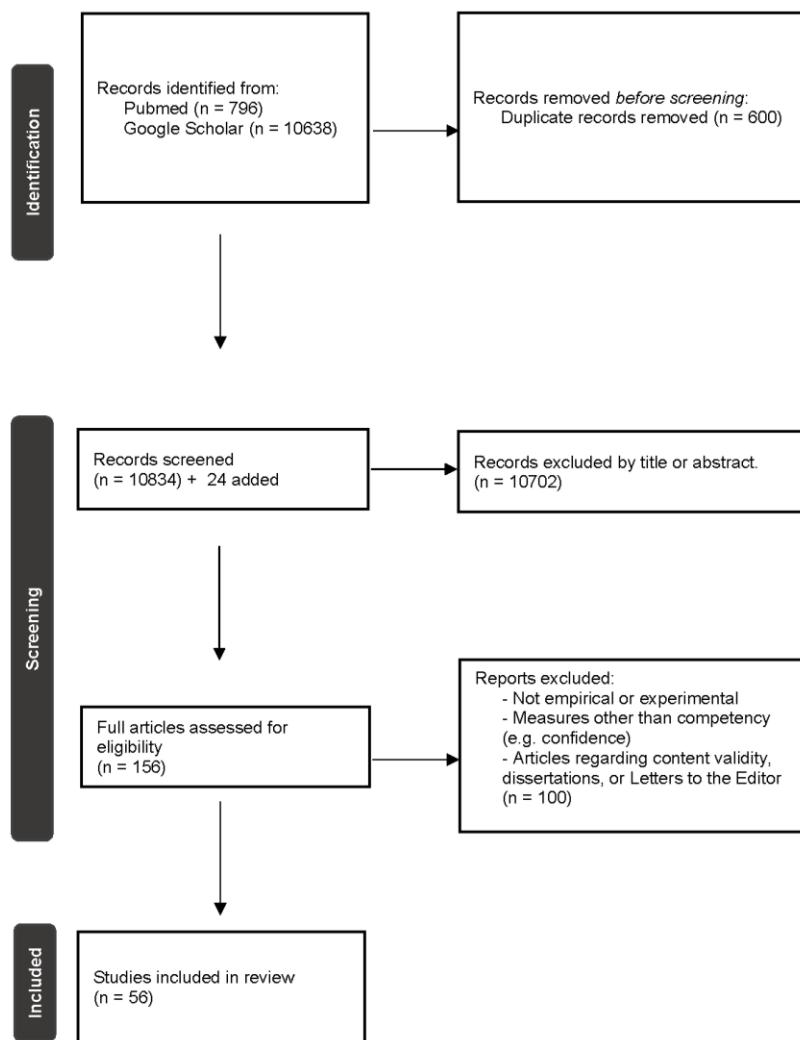
This systematic review aimed to examine the effectiveness of I-XR training for healthcare students (e.g., medical and nursing). The review focused on comparing I-XR training methods to traditional, non-immersive approaches. Specifically, we sought to determine whether I-XR training leads to greater improvements in both skill competency and knowledge acquisition. Skill competency and knowledge acquisition were assessed across multiple domains, including performance skills, objective performance measures, clinical reasoning, internship grades, problem-solving skills, and skills knowledge.

This review offers several unique contributions to the field. First, it examines six distinct outcomes related to skill competency and knowledge acquisition. Second, it focuses on both graduate medical and nursing education across a wide range of techniques, providing a comprehensive overview of I-XR's potential in healthcare training. Third, recognizing that the effectiveness of I-XR may depend on the specific alternative being compared, the review analyzes results according to the type of non-immersive (comparison) group used (e.g., didactic instruction, manikin-based training, simulated patients). Fourth, unlike many reviews that focus on a single technology, this review encompasses the full breadth of I-XR techniques, including AR, VR, and MR. Finally, it assesses each study for its integration of pedagogical approaches and learning theory, providing insights into best practices for implementing I-XR in healthcare education.

## 2. Methods

### 2.1. Search Strategy

Consistent with Best Evidence in Medical Education (BEME) recommendations, the search was performed on the inclusive databases of PubMed and Google Scholar in January 2023 and again on April 1, 2024, and followed the Joanna Briggs Institute for scoping reviews. Search terms included “simulation in medical (or nursing) education” and “augmented reality (AR)” [or virtual reality (VR) or “mixed reality (MR)’] and “skill outcome” (or “skill performance”). We selected articles based on the PRISMA method (see Figure 1). We have adhered to the PRISMA guidelines. This review was not pre-registered.



**Figure 1.** PRISMA Flowchart.

## 2.2. Inclusion and Exclusion Criteria

Only English articles were included. Only peer-reviewed empirical articles between the dates of January 1, 2016, and April 1, 2024, were included. XR papers surged in 2017, so this window captures the increase from 2016 to the present (Velev & Zlateva, 2017). No dissertations, letters to the editor, commentaries, or opinion pieces were included. Only articles with a control group were included, which could include a within-subject pre-posttest design, or a true, external comparison group. We also only included articles that focused on objective performance measures of skill competency and knowledge acquisition (i.e., performance skills, internship grades, performance measures, clinical reasoning or judgment, and problem-solving skills). Lastly, we included only articles that included graduate medical or nursing students. Studies were excluded if they were outside the indicated dates, did not include a control group or pre-post design, did not include medical or nursing graduate students (e.g., undergraduates), were not available in English, or were not in a peer-reviewed journal. Studies that solely included measures of confidence, self-esteem, or other non-technical skills were excluded from the process. For cases in which the studies included these measures but also included skill competency and/or knowledge acquisition measures, we reported only on the latter.

The initial search with parameters yielded 11,434 articles. Duplications were removed and titles and abstracts were evaluated for topic appropriateness. We also reviewed additional papers using the ancestral method (i.e., reviewing the references of included papers) that did not come up in our original search. After screening, 156 papers were left to review (see Figure 1). After reading these

papers fully, 56 were included in the final data set ( $n = 100$  removed for reasons indicated above) [Appendix A].

### 2.3. Data Extraction

We extracted the year, student sample size, field (e.g., medical, nursing), type of I-XR (AR/VR/MR), comparison group types, outcome measures, and any mention of learning theory. We also included a description of each study's methods, results, and descriptive statistics for comparisons of interest (Table 1). At least three researchers reviewed each paper for the accuracy of these variables. Two coders coded the year, student sample, field, learning theory, and type of I-XR method. The agreement was  $> 95\%$  on the first pass. The first and second authors coded the type of comparison groups and outcome measures, as well as made the final assessment for each. Initial agreement exceeded 85%. If codes were discrepant, the first author made the final decision.

### 2.4. Data Coding

Outcome measures (i.e. skill competency and knowledge acquisition) were initially operationalized with six categories based on past literature: “performance skills”, including “direct observation of procedure” (DOPS); “performance measures” (Objective Structured Clinical Exams (OSCE), Global Operative Laparoscopic Assessment (GOALS), Global Rating Scale (GRS), Academic Achievement Test (AAT); “clinical judgment or clinical reasoning” (Laster Clinical Judgment Rubric (LCJR); “skills knowledge” (e.g., Mini-Clinical Evaluation Exercise (CEX), Mini-CEX, Neurological Physical Exam (NPE), multi-choice questions of skills (MCQ); “grade in internship”; and “problem-solving skills”. We then felt it necessary to subdivide “performance skills” into specific modifiers, regardless of the actual behavior (e.g., “performance skills errors,” “performance skills time,” “performance skills injury,” “performance skills dexterity”). Finally, we added an additional category: “performance skills other”, for cases in which the performance skill was not one of those noted above. If studies had more than one outcome measure, we evaluated each outcome separately. Therefore, the total number of measures exceeded the number of studies.

Comparison groups were grouped into seven categories, also based on previous literature. These groups included learning with: “print materials” (e.g., study guides, books, technical manuals), “teacher-led” (e.g., didactic/instructor-led), “electronic materials” (e.g., video tutorial, e-learning/computer materials), “practice” (e.g., BOX trainer, dissection, mannikin, simulated patient, case-based learning (CBL), or hands-on with instructor that was not immersive-XR). If a study had multiple comparison methods (e.g., electronic materials and practice), we coded it as “combined”. However, if a study included some students learning through print materials (comparison 1), and some learning via practice, the study was evaluated twice, in this case – once for each comparison group. If *all* students had additional training of some sort (including the I-XR group), we did not include that training as a comparison group. In those cases, the comparison group was coded as “did nothing additional”. Finally, some studies were pre-post studies with a comparison group. In those cases, the results were assessed concerning differences between the I-XR group and the comparison group(s) *after* the training, rather than any pre-post change.

We gave each study one of five final assessments (Table 1): (1) “Positive” (total support for I-XR methods on all outcomes for all comparisons), (2) “Negative” (total support for the control methods on all outcomes for all comparisons), (3) “No difference” (I-XR and control methods produced no differences on all variable and comparisons), (4) “Mixed Evidence Positive” or (5) “Mixed Evidence Negative”. “Mixed Evidence Positive” was used for cases in which the I-XR methods produced enhanced effects on more outcomes compared to the non-immersive (comparison) methods. Similarly, “Mixed Evidence Negative” was used for cases in which the comparison methods produced enhanced effects on more outcomes compared to the I-XR methods. Thus, “Mixed Evidence Positive” was given to any study that favored the I-XR methods, even if on some of the variables there were no differences with the comparison, or there were some differences that favored the comparison methods (as long as that number was fewer than those favoring the I-

XR methods). A similar logic was used for Mixed Evidence Negative, except that the evidence favored the comparison methods. There was only one instance where “Mixed” without a qualifier was used. This was because there were two outcomes and one favored each.

**Table 1.** Empirical Articles (56) Included in the Systematic Review.

Authors	Date	Type of I-XR	Population studied/ Setting	Number of participants	Control Group	Basic Experimental design/ Description	Results	DVs	Outcome Evidence for I-XR	Study Evidence for I-XR
Aebersold et al	2018	AR	nursing - tube placement	69 nursing students	1 (Combined)	Nursing students were tested on their ability to place a nasogastric tube. They were randomly assigned to either usual training (which included both video and didactic content) or an iPad anatomy-augmented virtual simulation training module.	The AR group was able to more accurately and successfully place the NGT, $p = 0.011$ . CONTROL: n = 34, M = 15.39 (SD = 1.01). EXPERIMENTAL: n = 35, M = 15.96 (SD = 0.75).	1 (performance skills - specific)	+ Performance skills	Positive
Andersen et al.	2021	VR + Electronic*	Medical - catheter placement	19 medical students	1 (Did nothing additional)	Students were split into two different training groups: immersive virtual reality versus the control group. Both groups viewed videos showing ultrasound-guided peripheral venous cannulation placement. The control group was given no further training.	The immersive VR group was significantly more successful at peripheral venous cannulation placement in comparison to the control group, $p \leq 0.001$ . CONTROL: n = 9, M = 22.2% placement [0.11, 0.41]. EXPERIMENTAL: n = 10, M = 73.3% placement [0.56, 0.86].	1 (performance skills - specific)	+ Performance skills	Positive
Andersen et al.	2022	VR	medical - Ultrasound skills	104 medical students	1 (Teacher)	Medical students were divided into two groups to learn Point-of-care ultrasound (POCUS) skills: a self-directed immersive virtual reality (IVR) group versus an instructor-led learning group. US skills were then assessed according to an OSAUS test.	There were no significant differences between the self-directed IVR and instructor-led groups in terms of OSAUS scoring or any other subgroup objectives. Overall effect, $p = 0.36$ . EXPERIMENTAL: n = 51, M = 10.3 [9.0, 11.5]. CONTROL: n = 53, M = 11.0 [9.8, 12.2].	1 (performance measures - OSCE)	X No difference performance measures	No Difference
Arents et al	2021	VR	medical - Obstetrics training	89 medical students	1 (Print)	Two weeks prior to medical students' OB/Gyn internship,	No significant difference in internship grade	2 (Grade internship (Skills knowledge based on MCQs)	X No difference Grade internship	No Difference

						students underwent teaching on gentle Caesarean Sections (gash) and general obstetric knowledge. Students were divided into either a control group that underwent conventional study, or an experimental group who watched 360-degree videos using VR. No significant difference	between groups, $p = .66$ (adjusted 0.68). EXPERIMENTAL: n = 53, M = 7.75	X No difference Skills knowledge
Azimi et al.	2018	AR	nursing - IV placement & Chest compression	20 nursing students	1 (Practice)	After the internship, the on multiple-choice testing authors analyzed the grade received for the internship, as well as administered both open-ended and multiple-choice question tests.	EXPERIMENTAL: n = 53, M = 6.63. CONTROL: n = 48, M = 6.67	Mean difference CI [-0.61, 0.55].
Banaszek et al.	2017	VR	Medical - surgery	40 medical students	2 (Practice) (Did nothing)	Results are assessed with respect to control. Students underwent either standard training or training with AR via a head-mounted display for learning needle chest decompression and IV-line placement skills. The students' skills were measured with a post-assessment both immediately after training and 3 weeks later.	EXPERIMENTAL: n = 10 CONTROL: n = 10	+ Performance skills 2 (performance skills - specific chest) (performance skills - other IV) X No difference performance skills Mixed Positive
						Medical students underwent five weeks of independent training sessions in one of three groups: a high-fidelity VR simulator and bench-top arthroscopic	Results are not reported for cross-over group post training. Both the high-fidelity VR simulator and bench-top arthroscopic	+ Performance measures 2 (performance measures GRS) (performance skills - time) + Performance skills Positive

Bayram & Caliskan 2019	VR	nursing -tracheostomy care	172 nursing students	1 (Combined)	<p>virtual reality arthroscopic simulator, a bench-top arthroscopic simulator, or an untrained group (control). To measure post-test skill acquisition, students performed a diagnostic arthroscopy on both simulators and were tested in a simulated intraoperative environment using a cadaveric knee. A more difficult surprise skills transfer test was also administered. Students were evaluated using the Global Rating Scale (GRS) and a timer to determine efficiency.</p> <p>The VR simulation group showed the greatest improvement in performance in the diagnostic arthroscopy crossover tests using the GRS, <math>p &lt; 0.001</math>.  <b>CONTROL:</b> <math>n =</math> not reported, <math>M_D = 0.75</math> (<math>SD</math> only reflected in error bars).  <b>EXPERIMENTAL VR:</b> <math>n =</math> not reported, <math>M_D = 12.6</math> (<math>SD</math> only reflected in error bars).</p> <p>VR group showed the fastest improvement simulated cadaveric setup with timer, <math>p &lt; .001</math>.  <b>CONTROL:</b> <math>n =</math> not reported, <math>M_D = 9.1</math> (<math>SD</math> only reflected in error bars).  <b>EXPERIMENTAL VR:</b> <math>n =</math> not reported, <math>M_D = 17.3</math> (<math>SD</math> only reflected in error bars).</p>	<p>Nursing students were divided into control and VR groups for tracheostomy care and skill knowledge. Both groups completed a theoretical class, labs, and small group study. The experimental group training, but did not differ was provided a game-based virtual reality</p> <p>Results for the less immersive VR are not reported. Only the immersive experimental group is compared to the 2 (performance skills performance skills control group. Both groups increased their skills performance after from one another, <math>p = 0.443</math>.</p>	<p>X No difference  <b>No Difference</b></p> <p>- specific suctioning (skills knowledge) X No difference skills knowledge</p>

Author	Year	Technology	Field	Number of students	Intervention	Control	Outcomes		Conclusion
							Performance skills	Performance measures	
Blumstein et al	2020	VR	medical - surgery	20 medical students	1 (Print)		Medical students were randomized into either standard guide (SG) or virtual reality (VR) learning groups to learn intramedullary nailing (IMN) of the tibia. Students then performed a simulated tibia IMN procedure immediately following their training and were evaluated by an attending surgeon using a procedure-specific checklist and 5-point global assessment scale. Students returned 2 weeks later for repeat training and testing.	The VR groups showed significantly higher global assessment scores, $p < 0.001$ . CONTROL: n = 10, M = 7.5, SD = <i>not reported</i> . EXPERIMENTAL: n = 10, M = 17.5, SD = <i>not reported</i> . The VR also completed a higher percentage of steps correctly according to the procedure-specific checklist, $p < 0.002$ . CONTROL: n = 10, M = 25, SD = <i>not reported</i> . EXPERIMENTAL: n = 10, M = 63, SD = <i>not reported</i> .	+ Performance skills + Performance measures
Bogomolova et al	2020	AR	BIO medical students - anatomy	58 (bio)medical students	2 (Print) (Practice)		Students were divided into three groups: (1) stereoscopic 3D Augmented-Reality (AR) group, (2) monoscopic 3D desktop model group, or (3) 2D anatomical atlas group. Students were told what	All groups performed equally well on the knowledge test, $p = 1.00$ . <i>Results are between the AR and atlas control.</i>	X No difference skills knowledge No Difference

Bork et al.	2019	AR	medical - anatomy	749 medical students	2 (Print) (Practice)	<p>the learning goals consisted of and were given instructions for the session. Visual-spatial abilities were measured before the learning session began. Post-session learning was measured using a 30-question knowledge test that tested factual, functional, and spatial organization of anatomical structures.</p> <p>Medical students were divided into one of three groups: (1) the control group using radiology atlases, (2) a virtual dissection table, or (3) AR Magic Mirror. A pre and post-test was taken about anatomy questions.</p>	<p>EXPERIMENTAL: n = 20, M = 47.8 (SD = 9.8).</p> <p>Pre-post not evaluated for final assessment. Both the AR Magic Mirror group and the Theory (control) group showed significantly increased post-test scores but did not differ from one another. <i>No p value for comparison between change in improvement given.</i></p> <p><i>Results are from the post scores between the AR and the theory from print control group.</i></p> <p>CONTROL: n = 24, M = 50.60, (SD = 12.53).</p> <p>EXPERIMENTAL: n = 24, M = 48.00 (SD = 13.07).</p>	X No difference	No skills knowledge
Brinkmann et al.	2017	VR	med medical - surgery	36 medical students	1 (Practice)	<p>Medical students underwent a 5-day laparoscopic basic skills training course using either a box-trainer or virtual reality (VR) training curriculum. Skills were measured by students' performance of</p>	<p>5 (performance skills - performance skills other peg)</p> <p>(performance skills - other cutting) X No difference</p> <p>(performance skills - other loop) X No difference</p> <p>(performance skills - performance skills other knot)</p>	Mixed Negative	

Author	Year	Modality	Task	Sample size	Intervention	Outcomes	Conclusion	
							Performance measures	No difference
Bube et al.	2020	VR	medical- cystoscopy	32 medical students	1 (Teacher)	an ex-situ laparoscopic cholecystectomy on a pig liver using RT and errors. The performance was evaluated by the Global Operative Assessment of Laparoscopic Skills (GOALS) score.	CONTROL: n = 18, M = 53 (SD = 21.3). EXPERIMENTAL: n = 18 , M = 44.4 (SD = 14.9).	(performance measures- GOALS) performance skills - Performance measures
						The two groups also did not differ on their pattern cutting, $p = 0.088$ . CONTROL: n = 18, M = 31.6 (SD = 17.3). EXPERIMENTAL: n = 18 , M = 42.6 (SD = 16.9).		
						The two groups did not differ on loop placement, $p = 0.174$ . CONTROL: n = 18, M = 46.3 (SD = 54). EXPERIMENTAL: n = 18 , M = 53.1 (SD = 32.5).		
						The two groups did not differ on their knot tying, $p = 0.174$ . CONTROL: n = 18, M = 37.2 (SD = 11.9). EXPERIMENTAL: n = 18, M = 42.6 (SD = 16.4).		
						The GOALS scores on four of the five items were significantly higher in the box-trained group compared to the VR-trained group ( <i>individual comparisons in Table 5 of original publication</i> ).		

Butt et al.	2018	VR	nursing - catheter	20 nursing students	1 (Combined)	training whereas the experimental group used VR and other self-directed simulation training methods. Three weeks after the training, participants performed cystoscopies on two patients, and performance was measured using a global rating scale (GRS).	CONTROL: n = 12, M = 14.3. EXPERIMENTAL: n = 13, M = 13.6. <i>CI of the difference only reported: [-2.4, 3.9].</i>	Nursing students were assigned to either a control group (traditional learning with a task trainer) or an experimental group (VR software/game) to learn catheter insertion skills. Skills were assessed approximately two weeks after completion of the training session.	VR group completed more procedures than traditional group, $p < 0.001$ . CONTROL: n = 10, M = 1.8 (SD = 0.42). EXPERIMENTAL: n = 10, M = 3.0 (SD = 1.3). Pass rates at two weeks were identical, no $p$ value given.	2 (performance skills – other number of procedures completed) (performance skills – other specific pass rates)	+ Performance skills	Mixed Positive
Cevallos et al.	2022	VR	medical - surgery	20 medical students and orthopedic residents	1 (Combined)	Medical students and orthopedic residents were randomized into either standard guide (SG) or virtual reality (VR) learning groups to learn pinning of a slipped capital femoral epiphysis (SCFE), a pediatric orthopedic surgery procedure. All participants watched a technique video, and the VR group completed additional training on the Osso VR surgical trainer. Participants then were asked to achieve "ideal placement," and performed a SCFE	The VR group showed superiority across multiple domains but where not statistically different from the control in the following: time to final pine placement, $p = 0.26$ . CONTROL: n = 10, M = 706 (SD shown in figure). EXPERIMENTAL: n = 10, M = 573 (SD shown in figure). VR performed better compared to control for pin in and out, $p = 0.28$ . CONTROL: n = 10, M = 1.7 (SD shown in figure).	5 (performance skills – other specific pin in and outs) (performance skills – errors) (performance skills – other specific pin tip to bone) (performance skills – other specific angle)	X No difference performance skills X No difference performance skills X No difference performance skills + Performance skills	X No difference performance skills X No difference performance skills X No difference performance skills + Performance skills	Mixed Positive	

Chao et al	2021	VR	nursing - tube placement	45 nursing students	1 (Electronic)	<p>guidewire placement on EXPERIMENTAL: n = 10, Sawbones model 1161. Evaluation was based on time, number of pins "in-and-outs", articular surface penetration, angle between the pin and physis, distance from pin tip to subchondral bone, and distance from center-center point of the epiphysis.</p> <p>VR group performed fewer surface penetrations, <math>p = 0.36</math>. CONTROL: n = 10, M = 0.4 (SD shown in figure). EXPERIMENTAL: n = 10, M = 0.2 (SD shown in figure).</p> <p>VR group had smaller distance pin to tip to subchondral bone, <math>p = 0.42</math>. CONTROL: n = 10, M = 5.8 (SD = 3.36). EXPERIMENTAL: n = 10, M = 7.2 (SD = 6.5).</p> <p>VR group had lower angle deviation between the pin and physis, <math>p &lt; 0.05</math>. CONTROL: n = 10, M = 4.9 (SD = 3.0). EXPERIMENTAL: n = 10, M = 2.5 (SD = 1.42).</p>	<p>1 (skills knowledge)</p> <p>Nursing students were randomly assigned into two groups to learn nasogastric (NG) tube feeding: (1) immersive 3D interactive video program group or (2) regular demonstration video. Students completed a pre- and post-intervention questionnaire, which</p> <p>Knowledge scores on NG tube feeding improved significantly in both groups; however, there was no significant difference in the knowledge scores after treatment, <math>p = 0.77</math>. CONTROL: n = 23, M = 11.7 (SD = 1.86).</p>	X	No difference skill knowledge	No Difference

Author	Year	Intervention	Topic	Sample	Intervention Type	Description			Conclusion	
						Design	Outcomes	Conclusion		
Chao et al	2022	VR	Medical - intake skills	64 medical students	1 (Electronic)	Students were randomized into two groups and received either a 10-minute immersive 360-degree virtual reality or a 2D virtual reality instructional video on history taking and physical examination skills. Within 60 minutes of watching the video, students performed a focused history and physical on a patient. The Direct Observation of Procedural Skills (DOPS) was used to measure physical exam skills, and the Mini-CEX was used to measure general history and physical exam skills.	The average DOPS-total score was significantly higher in the VR video group compared to the 2D video group, $p = .01$ . CONTROL: n = 32, M = 85.8 (SD = 3.2). EXPERIMENTAL: n = 32, M = 88.4 (SD = 4.0). No significant differences in the average Mini-CEX scores were found between the groups, $p = 0.75$ . CONTROL: n = 32, M = 39.8 (SD = 5.2). EXPERIMENTAL: n = 32, M = 40.1 (SD = 4.1).	2 (performance skills - other/DOPS) (skills knowledge - mini-CEX)	+ Performance skills	Mixed Positive
Chen & Liou	2023	AR	Nursing students – first aid	95 nursing students	1) Practice	Nursing students were divided into two groups for pediatric first-aid training. The control group performed simulation using a traditional Resusci Annie whereas the experimental group used an interactive Resusci Anne that was overlaid AR. Pre and post tests were given to evaluate participant	The AR intervention group showed significantly higher post-test knowledge, $p < 0.001$ . CONTROL: n = 49, M = 18.08, (SD = 1.6). EXPERIMENTAL: n = 46, M = 18.78, (SD = 1.1). The AR group also showed improved skill in first aid level scoring compared to the control group post-test, $p < 0.001$ .	2 (skills knowledge) + Skills knowledge (performance measures – other specific first aid)	+ Performance measures	Positive

						knowledge and skills. Knowledge was assessed using a 20-question test. Skill level was assessed using a graded evaluation checklist.	CONTROL: n = 49, M = 29.71, (SD = 1.5). EXPERIMENTAL: n = 46, M = 32.52, (SD = 1.3) .		
Ekstrand et al	2018	VR + Print*	medical - neuroanatomy	64 medical students	1 (did nothing additional)	Medical students were assigned into either a control group who underwent paper-based neuroanatomy learning, or an experimental group who underwent neuroanatomy learning with VR. Pre and post-intervention tests were given, including a post-test immediately after the study completion and one 5-9 days later.	Both groups showed significant improvement between pre- and post-test scores, but no significant differences on the neuroanatomy test between the groups on either of the post-test results, $p = 0.5$ . Means and SDs are not reported: T-statistic reported for control ( $n = 33$ ) vs. VR ( $n = 31$ ) post-training, $t(62) = -0.38$ .	1 (skills knowledge)	X No difference skills knowledge No Difference
Fu et al	2020	VR	medical - suturing	14 medical students	1 (Practice)	Students were assigned to one of two training groups: (1) the VBLaST-SS (virtual simulator) training group or (2) the FLS training group. Students then watched a video that taught the intracorporeal suturing task they were going to be practicing. Students then performed the task on both systems to measure baseline performance. Students then practiced once a day, five days a week, for three weeks. Performance scoring was based on the original FLS scoring system.	Both training modalities showed significant performance improvement, but there were no significant differences in the group x time interaction, $p = 0.20$ . Learning curves for both learning modalities were also similar. Means and SD only shown in figure.	1 (performance skills – specific FLS)	X No difference performance skills No Difference

Haerling et al.	2018	VR	Nursing - case evaluation for COPD	81 nursing students	1 (Practice)	<p>This study placed students in two groups, those using mannequin-based simulations and those using VR simulations. Participants completed a standardized patient encounter of a complex case involving a patient with COPD. Pre and post-intervention knowledge assessments were also performed using the LCJR and the C-SEI.</p> <p>Students in both groups showed significant improvement in post-test knowledge assessment. Scores between the groups were not significantly different in the post-test knowledge assessment, <math>p = 0.48</math>.</p> <p>CONTROL: <math>n = 14</math>, <math>M = 79.82</math> (<math>SD = 17.63</math>).  EXPERIMENTAL: <math>n = 14</math>, <math>M = 82.16</math> (<math>SD = 11.76</math>).</p>	3 (skills knowledge)	X No difference skills knowledge	X No difference No clinical reasoning Difference
Han et al.	2021	VR + SP	medical - neurological	95 medical students	1 (did nothing additional)	<p>Medical students were divided into two groups: a standardized patient (SP) group that was provided neurological findings using conventional methods (verbal description, pictures, videos) versus a SP with Virtual Reality-based</p> <p>There was also no statistical difference post-intervention on the C-SEI between groups.</p> <p>CONTROL: <math>n = 14</math>, <math>M = 84.62</math> (<math>SD = 14.91</math>).  EXPERIMENTAL: <math>n = 14</math>, <math>M = 81.93</math> (<math>SD = 16.41</math>).</p> <p>The SP + VR group had significantly higher NPE scores compared to the SP group, <math>p = 0.043</math>.</p> <p>CONTROL: <math>n = 39</math>, <math>M = 3.40</math> (<math>SD = 1.01</math>).  EXPERIMENTAL <math>n = 59</math>, <math>M = 3.81</math> (<math>SD = 0.92</math>).</p>	1 (skills knowledge - NPE score) + Skills knowledge Positive	X No difference clinical reasoning	

Henssen et al.	2019	AR	medical & Biomedical-neuroanatomy	31 medical and biomedical students	1 (Print)	<p>Neurological Examination Teaching Tool (VRNET) group. A researcher measured student performance using the Neurologic Physical Exam (NPE) score.</p> <p>Results are assessed with respect to control. The control group showed to one of two groups for improved post-test scoring learning neuroanatomy. compared to the AR, <math>p = 0.035</math>.</p> <p><i>Results for adapted test scores after training are reported next.</i></p> <p>CONTROL: <math>n = 16</math>, <math>M = 60.6</math> (<math>SD = 12.4</math>). EXPERIMENTAL: <math>n = 15</math>, <math>M = 50.0</math> (<math>SD = 10.2</math>).</p>	<p>1 (skills knowledge) - Skills knowledge Negative</p>	
Hu et al.	2020	VR + workshop in ultrasound*	medical - Ultrasound skills	101 medical students	1 (Print)	<p>Medical students took place in an ultrasonography (US) training program. They were divided into either the virtual reality (VR) intervention group, or the control group. Both groups participated in an ultrasound workshop; however, the intervention group used a self-directed VR-enhanced anatomy review and used VR to complete additional review sessions during the US hands-on practice. After the US workshop was completed, participant competency was measured using a</p> <p>Participants in the intervention group showed significantly higher scores on US task performance overall, <math>p &lt; 0.01</math>. Results below are for mean rank. No variability given.</p> <p>CONTROL: <math>n = 54</math>, <math>M_R = 38.52</math>. EXPERIMENTAL: <math>n = 47</math>, <math>M_R = 65.34</math>.</p> <p><math>2</math> (skills knowledge) + Skills knowledge (performance skills - other practical US) + Performance skills</p> <p>The VR group also showed significantly better scores on the knowledge test, <math>p &lt; .05</math>. CONTROL: <math>n = 54</math>, median = 2 (IQR = 3). EXPERIMENTAL: <math>n = 47</math>, Median = 3 (IQR = 3).</p>	<p>Positive</p>	

						standardized practical US test, which focused on the identification of various anatomical structures, and a 10-Q MCQ on anatomy.				
Issleib et al.	2021	VR	medical - CPR	160 medical students	1 (Practice)	Medical students were randomized into an intervention or control group. The intervention group completed a the BLS course in virtual reality, whereas the control group underwent standard BLS training. At the end of training, all students performed a 3-minute practical test using the Leardal Mannequin to record no flow time on the task.	The control group had significantly shorter no flow time compared to the VR, $p < 0.0001$ . CONTROL: n = 104, M = 82.03 (SD = not reported). EXPERIMENTAL: n = 56, M = 92.96 (SD = not reported).	1 (performance skills - specific no flow time)	- Performance skills	Negative
Jaskiewicz et al.	2020	VR + Practice*	medical - CPR	91 medical students	1 (Teacher)	Both the control and experimental groups completed a 3-hour BLS course including background training and practice on a CPR mannequin. Students then participated in either a tadeonal teaching or VR scenario where hands-only CPR was completed. The quality of the chest compressions (rate and depth) was then tested and analyzed.	There were no significant differences in chest rate compression performance between the control and virtual reality groups, $p = 0.48$ . CONTROL: n = 45, Median = 114 (IQR 108-122). EXPERIMENTAL: n = 45, Median = 115 (IQR 108-122). There was also no significant difference on chest rate depth between groups, $p > 0.05$ . CONTROL: n = 45, Median = 48 (IQR = 44-55). EXPERIMENTAL: n = 45, Median = 49 (IQR = 43-53).	3 (performance skills - specific rate) (performance skills - specific depth) (performance skills - specific relaxation)	X No difference performance skills X No difference performance skills -Performance skills	Mixed Negative

						Finally, there was a significant increase in the percentage of chest compression relaxation for the control group compared to the VR group, $p < 0.01$ . CONTROL: $n = 45$ , Median = 97 (IQR = 85-100). EXPERIMENTAL: $n = 45$ , Median = 69 (IQR = 26-98).
Jung & Park	2022	VR	Nursing students – chempot insertion surgery	60 nursing students	1 (Print)	<p>Nursing students were divided into two groups to learn chempot insertion surgery. The control group's learning consisted of instruction by an operating nursing instructor, learning via a handout, and time for self-study. The experimental group used VR. Pre and post-test knowledge was assessed using a 10-point questionnaire about key knowledge of insertion.</p> <p>The VR group showed significantly higher post-test knowledge scores compared to the control group after training, <math>p = 0.001</math>. CONTROL: <math>n = 30</math>, <math>M = 4.80</math> (<math>SD = 1.65</math>). EXPERIMENTAL: <math>n = 30</math>, <math>M = 6.97</math> (<math>SD = 1.35</math>).</p>
Kane et al.	2022	VR	med students- obstetrics	69 medical students	1 (Electronic)	<p>Medical students were placed into one of two groups to help them learn and conceptualize fetal lie and presentation. The interventional group was immersed in a virtual reality learning environment (VRLE) to explore fetal lie, and the control group used traditional 2D images. After their sessions,</p> <p>No significant differences were found between the two groups in terms of knowledge assessment, although the authors note that there was a noticeable trend of higher success rates in the intervention group in the VR group compared to the control group) for combined lie and presentation scores, <math>p</math> = not reported.</p> <p>2 (performance skills - time)      X No Difference performance skills (performance skills - specific success)      + Positive performance skills</p>

Kowalewski et al 2019	VR	medical - surgery	100 medical students	1 (Did nothing)	clinical exam skills were tested using an obstetric abdominal model. Knowledge was assessed by students' ability to determine fetal lie and presentation on this model. Time taken to complete the test was also measured.	CONTROL: n = 34, M = 70.0 (SD = <i>not reported</i> ). EXPERIMENTAL : n = 33, M = 56% (SD = <i>not reported</i> ). However, time to complete the task was significantly less in the intervention group compared to the control group, $p = 0.012$ . CONTROL: n = 34, M = 38 (SD = 10.83). EXPERIMENTAL: n = 33, M = 45 (SD = 12.95).	X No difference performance measures	X No difference performance measures	Mixed Positive
					Medical students were divided into three groups to complete laparoscopic training: (1) the control group, which received no training, (2) the "alone" group, and (3) the dyad group. Intervention groups completed box and VR training, after which performance was measured with a cadaveric porcine laparoscopic cholecystectomy (LC), and the objective structured assessment of technical skills (OSATS) was used. Global operative assessment of laparoscopic skills (GOALS), time to completed LC, and VR performances were also measured.	Results are reported for improvement between the VR and control group only.  The VR group and the control group did not differ on the OSATS, $p = 0.548$ . CONTROL: n = 20, M = 37.1 (SD = 7.4). EXPERIMENTAL: n = 40, M = 40.2 (SD = 9.8).  The two groups did not differ on the GOALS either, $p = 0.998$ . CONTROL: n = 20, M = 10.1 (SD = 3.0). EXPERIMENTAL: n = 40, M = 10.6 (SD = 3.0).  The VR groups were faster than the control group in completion time, $p < 0.001$ . CONTROL: n = 20, M = 13.5 [11.8, 17.5].	5 (performance measures -OSATS) (performance measures - GOALS) (performance skills - time) (performance skills- other specific path) (performance skills- other specific length)	+ Performance skills + Performance skills + Performance skills	

						EXPERIMENTAL: n = 40, M = 10.2 [7.9, 11.3].		
						The VR group also had fewer movements, $p = 0.002$ . CONTROL: n = 20, M = 871 [637, 1105]. EXPERIMENTAL: n = 40, M = 683 [468, 898].		
						The VR group also had a shorter path length, $p = 0.004$ . CONTROL: n = 20, M = 1640 [1174, 2106]. EXPERIMENTAL: n = 40, M = 1316 [948, 1684].		
Küçük et al	2016	AR	medical -anatomy	70 medical students	1 (Print)	Medical students were placed into a control group, which used traditional teaching methods (textbook) or an experimental group, which used mobile augmented reality (MAR) technology (MagicBook) to learn neuroanatomy. Post-intervention knowledge was measured using an Academic Achievement Test (AAT), a 30-question multiple-choice test.	The experimental MAR group showed significantly superior performance on the AAT test, $p < 0.05$ . CONTROL: n = 34, M = 68.34 (SD = 12.83). EXPERIMENTAL: n = 36, M = 78.14 (SD = 16.19).	1 (performance measures - AAT) + Performance measures
Lau et al	2023	VR	Nursing students	34 nursing students	1) Nothing (pre-post)	Nursing students participated in pre-test and post-test knowledge questionnaires regarding both subcutaneous insulin injection and intravenous therapy.	There was a significant improvement in knowledge test scores (IV and subcutaneous injection) after training, $p = 0.075$ . Only z-scores were reported with ranks.	1 (skills knowledge) + Skills knowledge

							After completing the pre-test, participants underwent learning using immersive VR before completing the post-test questionnaire.	<i>No M or SDs are reported pre vs. post.</i>	
Lemke et al	2020	AR	medical - suturing	44 medical students	2 (Teacher) (Combined)		<p>Results are reported among three groups. No significant differences among the intervention groups in the number of full-length sutures used, <math>p = .376</math>. <i>Means and SD are reported for Teacher control vs. AR</i></p> <p>Students were randomized into one of three intervention groups to learn suturing skills: (1) faculty-led, (2) peer tutor-led, or (3) holography-augmented intervention arms (Suture Tutor).</p> <p>Outcomes measured include: the number of simple interrupted sutures that were placed to achieve proficiency, the total number of full-length sutures used, and time to achieve proficiency.</p>	<p><b>CONTROL:</b> N = 16, M = 80.0 (SD = 47.2).</p> <p><b>EXPERIMENTAL:</b> n = 14, M = 107.4 (SD = 61.3).</p> <p>No significant differences among the intervention groups in the number of simple interrupted sutures placed, <math>p = 0.735</math>.</p> <p><b>CONTROL:</b> n = 16, M = 9.3 (SD = 5.3).</p> <p><b>EXPERIMENTAL:</b> n = 14, M = 11.0 (SD = 6.0).</p> <p>No significant differences among the intervention groups in the time to achieve proficiency, <math>p = 0.390</math>.</p> <p><b>CONTROL:</b> n = 16, M = 158.1 (SD = 89.2).</p> <p><b>EXPERIMENTAL:</b> n = 14, M = 205.0 (SD = 113.2).</p>	<p>X No difference performance skills</p>
Logishetty et al.	2019	AR	Medical - surgery	24 medical students	1 (Practice)	This study simulated total hip arthroplasty	AR intervention group showed smaller average 2 (performance skills - errors)	+ Performance skills Mixed Positive	

Maresky et al. 2018	VR immersive + VR simulation*	medical - anatomy	42 medical students	1 (Print)	(THA) and placed students in one of two groups to determine what training is more effective at improving the accuracy of acetabular component positioning: (1) augmented reality (AR) training (with live holographic orientation feedback) or (2) hands- on training with a hip arthroplasty surgeon. Students participated in one baseline assessment, training session, and reassessment a week for four weeks and were recorded on the target angle (inclination – anteversion).	errors than the control practice with surgeon group after training, $p <$ 0.0001. CONTROL: n = 12, M = 6 (SD = 4). EXPERIMENTAL: n = 12, M = 1 (SD = 1).	(performance skills – other specific target angle) X No difference Performance skills	
					Before learning cardiac anatomy, medical students underwent a VR simulation of the subject. Students then were separated into either a control group that continued independent anatomy study, or an experimental group that underwent an immersive VR experience. Pre and post-test scores were obtained, which measured both conventional and visual- spatial (VS) cardiac anatomy questions.	In the final session, both groups showed improvement on the target angle, but there was no significant difference in performance between groups, $p = 0.281$ . Means and SD are reported as differences from pre. CONTROL: n = 12, $M_D = -$ 8.4 [-7.0, -9.8]. EXPERIMENTAL: n = 12, $M_D = -7.8$ [-5.5, -10.2].		

Moll- Khosrawi et al.	2022	VR	medical - BLS	88 medical students	1 (Electronic)	<p>Medical students were placed into one of two groups to compare a control group that received web-based basic life support (BLS) training to an intervention group that underwent additional individual virtual reality (VR) training. The quality of BLS skills was assessed after training with a no-flow-time indicator. Overall BLS performance was also assessed using an adapted observational checklist, graded by experts.</p> <p>The VR intervention group showed significantly lower no-flow-time compared to the control, <math>p = 0.009</math>.  <b>CONTROL:</b> <math>n = 42</math>, <math>M = 11.05</math> (<math>SD = 10.765</math>).  <b>EXPERIMENTAL:</b> <math>n = 46</math>, <math>M = 6.46</math> (<math>SD = 3.49</math>).</p> <p>The VR group also showed significantly superior overall (lower penalty point) for their BLS performance in comparison to the control group, <math>p &lt; 0.001</math>.  <b>CONTROL:</b> <math>n = 42</math>, <math>M = 29.19</math> (<math>SD = 16.31</math>).  <b>EXPERIMENTAL:</b> <math>n = 46</math>, <math>M = 13.75</math> (<math>SD = 9.66</math>).</p>	<p>2 (performance skills – other specific flow time) (performance measures - checklist)</p> <p>+ Performance skills</p> <p>+ Performance measures</p>	Positive
Moro et al	2017	AR /VR	medical- anatomy	59 medical and health science students	1 (Electronic)	<p>Medical students completed a lesson on skull anatomy using one of three learning modalities: (1) virtual reality (VR), (2) augmented reality (AR), or (3) tablet-based (TB). After their 10-minute anatomy lesson using their respective learning modality, students completed a 20-question multiple-choice anatomy test to measure their knowledge.</p> <p><i>Means are for VR group and control group only. Standard deviations are depicted only in figure.</i></p> <p><b>CONTROL:</b> <math>n = 22</math>, <math>M = 61</math>.  <b>EXPERIMENTAL:</b> <math>n = 20</math>, <math>M = 59</math>.</p>	<p>1 (skills knowledge)</p> <p>X No difference skills knowledge</p>	No Difference
Nagayo et al	2022	AR	medical - surgery	38 medical students	1 (Electronic)	<p>Medical students were randomized into one of two groups for self-training suturing learning: (1) the</p> <p>Both groups showed significant improvement between pretest and posttest scores in both global rating and task-</p>	<p>1 (performance skills – other specific suturing performance)</p> <p>+ Performance skills</p>	Positive

						augmented reality (AR) training group or (2) the instructional video group. Both groups watched an instructional video on subcuticular interrupted suturing and took a pretest. They then practiced the suture 10 times using their assigned learning modality, before completing a post-test. Pre- and post-tests were performed on a skin pad and were graded using global rating and task-specific subscales.	specific subscales on suturing performance; however, no significant difference in performance was found between the AR and instruction video training groups using the global rating, $p = 0.38$ . CONTROL: n = 19, M = 15.11 (SD = 2.84). EXPERIMENTAL: n = 19, M 15.03 (SD = 1.94).			
Neumann et al	2019	VR	medical - surgery	51 medical students	1 (Electronic)	Medical students were randomized into two groups for cystoscopy (UC) and transurethral bladder tumor resection (TURBT) training. The control group watched video tutorials by an expert. After completion of the training, students performed a VR-UC and VR-TURBT task and 12 measures of performance were recorded.	Both groups improved on three variables after training, including significantly lower average procedure length, lower resectoscope movement, and accidental bladder injury, but there was only one significant difference in the improved performances for the VR VR-TURBT compared to the control task and 12 measures of (procedure time, $p = 0.04$ ). All Means and SD listed in Table 2.	12 (performance skills - other specific skills - injury) $\times 9$ (performance skills - injury) (performance skills - time $\times 2$ )	X No difference performance skills (for 11 of the 12 variables) + Performance skills	No Difference
Nielsen et al.	2021	VR	medical - Ultrasound skills	20 medical students	1 (Electronic)	Medical students were randomized into either a virtual reality (VR) or e-learning group for ultrasound education and training. Performance was scored using the OSAUS.	The VR group showed significantly higher scoring on the OSAUS compared to the e-learning group after intervention, $p \leq 0.001$ . CONTROL: n = 9, M = 125.7 (SD = 16.2). EXPERIMENTAL: n = 11, M = 142.6 (SD = 11.8).	1 (performance measures - OSAUS)	+ Performance measures	Positive

Reference	Year	Intervention	Setting	Sample	Design	Outcomes			Conclusion
						Skills	Knowledge	Performance	
Noll et al	2017	AR	medical - dermatology	44 medical students	1 (Print)	1 (skills knowledge)	+ (marginal) Skills knowledge	Positive	
Orland et al.	2020	VR+	Medical - surgery	25 medical students	1 (Print)	1 (performance skills - errors)	+ Performance skills	Positive	

							EXPERIMENTAL: n = 9, M = 18 (SD = 8).		
Plotzky et al.	2023	VR 'high'	Nursing students- endotracheal suctioning skills	131 nursing students	2 (low VR) + Electronic	<p>Nursing students were split into one of three groups for learning of endotracheal suctioning skills. The control group's intervention was a video tutorial. The intervention groups consisted of a VR low group and a VR high group. The VR 'low' group's intervention consisted of basic VR technology, including head tracking and controller-based controls. The VR 'high' group's intervention consisted of more advanced VR technology, including head and hand tracking, allowing users to interact with their actual hands, and as well as supplementing with real-world video clips.</p> <p>Participants were assessed using a knowledge test and a skill demonstration test on a manikin using an objective structured clinical examination (OSCE). The knowledge test was given immediately after intervention and 3 weeks later.</p>	<p>Each of three groups showed a significant increase in knowledge acquisition, however, there was no significant difference among them for skills knowledge, <math>p = 0.730</math>.</p> <p><i>Means and SDs are for the VR+ vs. control group after intervention.</i></p> <p>CONTROL: n = 43, M = 7.16 (SD = 2.29).</p> <p>EXPERIMENTAL: n = 47, M = 7.06 (SD = 1.42).</p> <p>There was a significant difference among groups on the OSCE after intervention, <math>p &lt; 0.001</math>.</p> <p><i>Means and SDs are for the VR+ vs. control group after intervention.</i></p> <p>CONTROL: n = 43, M = 11.95 (SD = 1.65).</p> <p>EXPERIMENTAL: n = 47, M = 9.41 (SD = 2.70).</p>	<p>X No difference</p> <p>2 (skills knowledge (performance measures -OSCE)</p> <p>- Performance measures</p>	<p>skills knowledge</p> <p>Mixed</p> <p>Negative</p>

Price et al.	2018	VR	nursing - triage skills	67 nursing students	1 (Practice)	There were no significant differences between the		+ Performance skills	Positive	
						This study compared the VR and clinical simulation use of VR to clinical simulation to determine victims that were correctly triaged, $p = 0.612$	- specific triage variables)			
Ros et al.	2020	VR + Print	medical -surgery	176 medical students	1 (Did nothing additional)	All students were given a technical note detailing an external ventricular drainage (neurosurgical) technique. Students were randomized into two groups, one of which received no additional training, and one which used immersive virtual reality (VR) as supplemental teaching. Knowledge was assessed with a multiple-choice test immediately after training and six months after.	VR training showed significantly superior knowledge gain, after both initial assessment and at the six-month mark, $p = 0.01$ . <i>Means and SD are reported after the initial training.</i> CONTROL: n = 88, M = 4.59, (SD = 1.4). EXPERIMENTAL: n = 85, M = 5.17 (SD = 1.29).	1 (skills knowledge) + Skills knowledge	Positive	
Ros et al.	2021	VR	Medical- lumbar puncture	89 medical students	1 (Teacher)	Medical students were randomized into one of two groups to complete training in how to perform a lumbar puncture. The control group participated in traditional lecture learning whereas Group 2 participated in immersive VR 3D video filmed from first-person point of view (IVRA-FPV). After training,	The group that participated in the traditional lecture showed significantly superior scoring n oral examination, $p < 0.001$ . CONTROL: n = 45, M = 4.06, SE = 0.12. EXPERIMENTAL: n = 44, M = 4.97, SE = 0.10.	- Skills knowledge 3 (skills knowledge) (performance skills - time) (performance skills - errors)	+ Performance skills	Mixed Positive
							The VR group took more time to perform the	+ Performance skills		



Smith et al.	2018	VR immersive or VR computer	nursing - decontamination skills	172 nursing students	1 (Print)	<p>teaching group for teaching skull base tumors and skull anatomy. The traditional teaching group used literature-based learning, problem-based teaching, and case-based teaching, whereas the VR groups used real case images and Hololens (VR) glasses. After completion of their intervention.</p> <p>Nursing students were divided into 3 groups for decontamination skills training. The control group used a traditional written instructions learning method, whereas the experimental groups underwent immersive VR or computer-based VR training. Post-learning competency was measured using a Decontamination checklist in which students performed skills on a mannequin. Cognitive test scores, performance scores, and time to complete skills were measured immediately post-training and 6 months later.</p> <p>Results are reported for the immediate follow-up post intervention. There were no significant differences among groups for the cognitive scores, <math>p = 0.568</math>. <i>Means and SD are shown both VR groups combined and the traditional control group only.</i></p> <p>CONTROL: n = 43, M = 19 [8, 23].</p> <p>EXPERIMENTAL: n = 43, M = 19 [13, 23].</p> <p>There was no difference among the groups for time to completion on the OSCE, <math>p = 0.723</math>. <i>Means and SD are shown both VR groups combined and the traditional control group only.</i></p> <p>CONTROL: n = 43, M = 260 [180, 360]</p> <p>EXPERIMENTAL: n = 43, M = 260 [180, 360].</p>	<p>traditional control group, <math>p &lt; 0.001</math>. CONTROL: n = 15, M = 63.6 (SD = 3.81). EXPERIMENTAL: n = 15, M = 77.07 (SD = 4.00).</p> <p>Results are reported for the immediate follow-up post intervention. There were no significant differences among groups for the cognitive scores, <math>p = 0.568</math>. <i>Means and SD are shown both VR groups combined and the traditional control group only.</i></p> <p>CONTROL: n = 43, M = 19 [8, 23].</p> <p>EXPERIMENTAL: n = 43, M = 19 [13, 23].</p> <p>There was no difference among the groups for time to completion on the OSCE, <math>p = 0.723</math>. <i>Means and SD are shown both VR groups combined and the traditional control group only.</i></p> <p>CONTROL: n = 43, M = 260 [180, 360]</p> <p>EXPERIMENTAL: n = 43, M = 260 [180, 360].</p>	X No difference 3 (skills knowledge) (performance skills - time) X No difference skills -other specific decontamination mannikin) - Performance skills	skills knowledge (performance skills - time) X No difference Mixed skills -other specific decontamination mannikin) - Performance skills

Stepan et al.	2017	VR	medical - neuroanatomy	66 medical students	1 (Electronic)	<p>The computer/mouse VR groups showed superior performance measures compared to the control group on the immediate post-test, <math>p = 0.017</math>.  <i>Means and SD are shown both VR groups combined and the traditional control group only.</i>  <b>CONTROL:</b> n = 43, M = 54 [46, 57].  <b>EXPERIMENTAL:</b> n = 43, M = 55 [46, 57].</p>			
Sultan et al.	2019	VR	Medical - tube placement	169 medical students	1 (Practice)	<p>Students completed preintervention, postintervention, and retention tests for assessment of knowledge. Medical students were assigned to either a control (online textbook) or an experimental (3D imaging VR interactive model) group for the learning of neuroanatomy. No significant differences in anatomy knowledge assessments were found between the control and VR groups, <math>p = 0.87</math>.  <b>CONTROL:</b> n = 33, M = 0.76 (SD = 0.14)  <b>EXPERIMENTAL:</b> n = 33, M = 0.75 (SD = 0.16)</p>	<p>1 (skills knowledge)</p>	<p>X No difference skills knowledge</p>	<p>No difference difference</p>

Watari et al.	2020	VR	medical students - basic clinical knowledge	210 medical students	1 Nothing (pre-post)	intervention assessment included an MCQs score and an Objective Structured Clinical Examinations (OSCE) score.	the conventional learning group, $p < 0.001$ . CONTROL: n = 112, M = 9.8, (SD = 4.2). EXPERIMENTAL: n = 57, M = 12.9, (SD = 4.1).  Students showed a significant increase in post-test scoring on both knowledge, $p = 0.003$ . EXPERIMENTAL <sub>pre</sub> : n = 169, M = 4.78, [4.55, 5.01]. EXPERIMENTAL <sub>post</sub> : n = 169, 5.12, [4.90, 5.43].	+ Skills knowledge 2 (skills knowledge) (clinical reasoning) + Clinical reasoning	Positive
Wolf et al	2021	AR	Medical - surgery	21 medical students	1 (Print)	Medical students were recruited for training in extracorporeal membrane oxygenation (ECMO) cannulation. They were split into two groups: (1) conventional training instructions for the first procedure, and AR instructions for the second; (2) reverse order (AR instructions for the first procedure). Participants performed the two ECMO cannulation procedures on a simulator. Training times and a detailed error protocol were used for assessment.	AR group showed minimally higher training times compared to the control group, <i>no p values given. Means and SD in figures only.</i>  AR group had significantly less errors when performing the second (more complex) simulation procedure, <i>no p values given. Means and SD in figures only.</i>	2 (performance skills - error) (performance skills - time) -Performance skills + Performance skills -Performance skills	Mixed

Yang & Oh	2022	VR	nursing - infant respiration	83 nursing students	2 (Teacher) (Practice)	<p>Knowledge scores increased for all groups post-intervention, but the VR and simulation group showed significantly higher knowledge after intervention compared to the control group, <math>p = 0.004</math>. Means and SDs are reported for the VR and control groups only.</p> <p>CONTROL: <math>n = 26</math>, <math>M = 11.85</math>, <math>(SD = 5.43)</math>.</p> <p>EXPERIMENTAL: <math>n = 29</math>, <math>M = 18.00</math>, <math>(SD = 2.55)</math>.</p>	+ Skills knowledge		
							- other specific resuscitation)	+ Problem solving (problem solving) Mixed Positive	
Yeo et al.	2018	AR + Ultrasound	medical- lumbar puncture and facet joint injection	36 medical students	1 (Practice)	<p>Nursing students were separated into three groups to undergo neonatal resuscitation training. These groups included: a virtual reality group, high-fidelity simulation group, and a control (online lectures only) group. Pre and post-test scores were analyzed on neonatal resuscitation knowledge, problem-solving ability, and clinical reasoning ability.</p> <p>The VR group showed significantly improved problem-solving ability scores in comparison to both the simulation and control groups, <math>p = 0.038</math>.</p> <p>CONTROL: <math>n = 26</math>, <math>M = 106.24</math>, <math>(SD = 24.52)</math>.</p> <p>EXPERIMENTAL: <math>n = 29</math>, <math>M = 122.72</math>, <math>(SD = 15.68)</math>.</p> <p>Clinical reasoning ability showed significant improved performance, but the none of the groups differed statistically, <math>p = 0.123</math>.</p> <p>CONTROL: <math>n = 26</math>, <math>M = 53.69</math>, <math>(SD = 12.02)</math>.</p> <p>EXPERIMENTAL: <math>n = 29</math>, <math>M = 59.66</math>, <math>(SD = 9.44)</math>.</p>	X No difference clinical reasoning	+ Problem solving (clinical reasoning)	Mixed Positive
							5 (performance skills - time) (performance skills - other specific time inside)	+ Performance skills - + Performance skills Positive	

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with ultrasound and AR using Perk Tutor software. Training involved learning lumbar puncture and facet joint injection skills on five different tasks for two tasks, a simpler and hard one (ultrasound-guided facet joint injection).	CONTROL: n = 10, M = 37.5, SD = <i>not reported</i> . EXPERIMENTAL: n = 10, M = 62.5, SD = <i>not reported</i> .	(performance skills - other specific damage) (performance skills - other specific success)	+ Performance skills + Performance skills + Performance skills
	AR was also better than control post-training for total time, $p < 0.001$ .	CONTROL: n = 10, M = 103, (SD = 13). EXPERIMENTAL: n = 10, M = 47, (SD = 3).	
	AR was also better than control post-training for time inside the phantom body, $p < 0.01$ .	CONTROL: n = 10, M = 31, (SD = 5). EXPERIMENTAL: n = 10, M = 14, (SD = 2).	
	AR was also better than control post-training for path distance inside the phantom body, $p < 0.01$ .	CONTROL: n = 10, M = 266, (SD = 76). EXPERIMENTAL: n = 182, M = 47, (SD = 36).	
	AR was also better than control post-training for potential tissue damage, $p = 0.03$ .	CONTROL: n = 10, M = 3217, (SD = 1173). EXPERIMENTAL: n = 10, M = 2376, (SD = 673).	

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Author	Year	Intervention	Setting	Sample	Design	Outcomes	Conclusion		
							Outcomes	Conclusion	
Yu et al	2022	VR + Practice (BOX)*	Medical - surgery	51 medical students	1 Nothing (pre-post)	Medical students completed four pre and post experiments with a box trainer based laparoscopic surgery simulators (VRLS). Students were assessed by expert surgeons using the Global operative assessment of laparoscopic skills (GOALS) standards on performance and time on two tasks, fundamental task (FT) and the color resection task (CRT)	Post-test assessments showed a significantly faster task completion post training on both tasks, $p < 0.01$ . <i>Means and SDs are shown pre and post for one task (FT).</i> EXPERIMENTAL <sub>pre</sub> : n = 51, M = 21.95, SD = <i>not reported</i> . EXPERIMENTAL <sub>post</sub> : n = 51, M = 14.04, SD = <i>not reported</i> . Post-training also had improved performance on the GOALS (heart rate) for both tasks, $p < 0.05$ . <i>Means and SDs are shown pre and post for one task (FT).</i> EXPERIMENTAL <sub>pre</sub> : n = 51, M = 94.96, (SD = 11.14). EXPERIMENTAL <sub>post</sub> : n = 51, M = 92.71, (SD = 11.67).	2 (performance measures -GOALS) (performance skills - time)	+ Performance skills + Performance measures -
Zackoff et al.	2020	VR + Teacher + Practice	Med students- infant respiratory distress	168 medical students	1 (Teacher)	Medical students completed standard respiratory distress training using traditional didactic material, as well as a mannequin simulation. A randomized group of students also completed an additional 30-minute training using immersive virtual reality with various infant simulations (no distress, respiratory distress, and impending respiratory failure). After training, all students completed a	Students who underwent additional VR intervention showed significantly superior status interpretation across all assessed dimensions and all cases, $p < 0.01$ . <i>Means and SDs for one case are shown (no distress).</i> CONTROL: n = 90, M = 104 (SD = 61.2). EXPERIMENTAL: n = 78, M = 124 (SD = 80). The additional VR status groups also had a higher recognition for the need	2 (skills knowledge- respiratory) (skills knowledge - level of care) + Skills knowledge + Skills knowledge	

Zhou et al.	2022	AR	Nursing - stroke	36 nursing students	1 (Practice)	<p>free-response test for increased care, <math>p = 0.0004</math>.  regarding various video questions including <i>Means and SDs for one case</i>  mental status, breathing, <i>are shown (respiratory</i>  breadth sounds, and <i>failure).</i>  vital signs for three CONTROL: n = 90, M = 41,  different cases, no (SD = 45.6).  distress, respiratory EXPERIMENTAL: n = 78,  distress, and respiratory M = 56, (SD = 72.7).  failure. In addition, the  need for escalation of  care was assessed in  each case.</p>	<p>The AR group spent less time than the control group in the critical phase of the stroke simulation, <math>p &lt; 0.05</math>.  Nursing students were divided into two groups: a mannequin-based simulation only (control) or a mannequin-based simulation with AR. They were assessed for clinical judgment with the LCJR. AR outperformed the control group on the LCJR for one of three sections (noticing), <math>p &lt; 0.05</math>. <i>Means and SDs are only shown in figure.</i></p>	<p>+ Performance skills  2 (performance skills - time) (clinical reasoning - LCJR)</p>	Positive

<sup>1</sup>. Note: Means (M) or Medians, and standard deviations (SD) or Interquartile Ranges (IQRs) are reported when available. Confidence intervals are reported in place of SD where available and appear in [ ].

## 2.5. Assessing Article Quality and Bias

There are several ways in which article quality and bias can be addressed, including the Critical Appraisal Skills Program (2024) instrument (Chen et al., 2017), the Mixed Methods Appraisal Tool (MMAT) (Hong et al., 2018), the Joanna Briggs Institute (JBI) tool (2014), the Medical Education Research Study Quality Instrument (MERSQI) (Jaros & Dallaghan, 2024), the Newcastle-Ottawa Scale for Education (NOS-E) (Wells et al., 2014) [48], and the Quality Assessment of Diagnostic Accuracy Studies revised (QUADAS-2) (Whiting et al., 2011). In our exploration of the literature, we found that the most common assessments were the MERSQI, NOS-E, and QUADAS-2. Because both the MERSQI and the NOS-E evaluate different aspects of study design and quality, we chose to evaluate each study on both.

The MERSQI is used as a measure to assess the quality of educational studies across eight domains: 1) study design, 2) sampling, 3) response rate, 4) type of data, 5) validity of evidence for evaluation measures, 6) data analysis sophistication, 7) data analytic appropriateness, and 8) outcome. The maximum score for study design, validity of evidence, outcome, and type of data is "3". For sampling design and response rate, the maximum score is "1.5". For data analysis sophistication, the max score is "2", and for data analysis appropriateness, the max score is "1". Scores are summed, with the max score being 16. We also included the items from the NOS-E, which evaluates a study on sample representation, comparison groups, retention, and blinding conditions. Sample representation, selection of the comparison group, retention, and blinding conditions were all scored with a maximum of "1", whereas the comparably of the comparison group was divided into randomized and non-randomized studies, each having a max score of "2". Scores are summed, with the max score being eight. The max score for the total of both scales was thus "24". The average MERSQI-2 plus NOS-E score across the 56 studies was 18.5, showing generally high quality of articles included (range: 15-22) (see Table 2).

To assess bias, we chose the QUADAS-2. The QUADAS-2 includes 11 items that assess risk of bias with "yes"/ "no"/ "unclear" marks. We included a total "yes" score that summed across the 11 statements, indicating bias. The average QUADAS-2 score across studies was 7.12 (range: 2-11), showing moderately high bias despite high study quality (see Table 2).

**Table 2.** Learning Theories Noted and Calculated MERSQI/NOS-E and QUADAS scores.

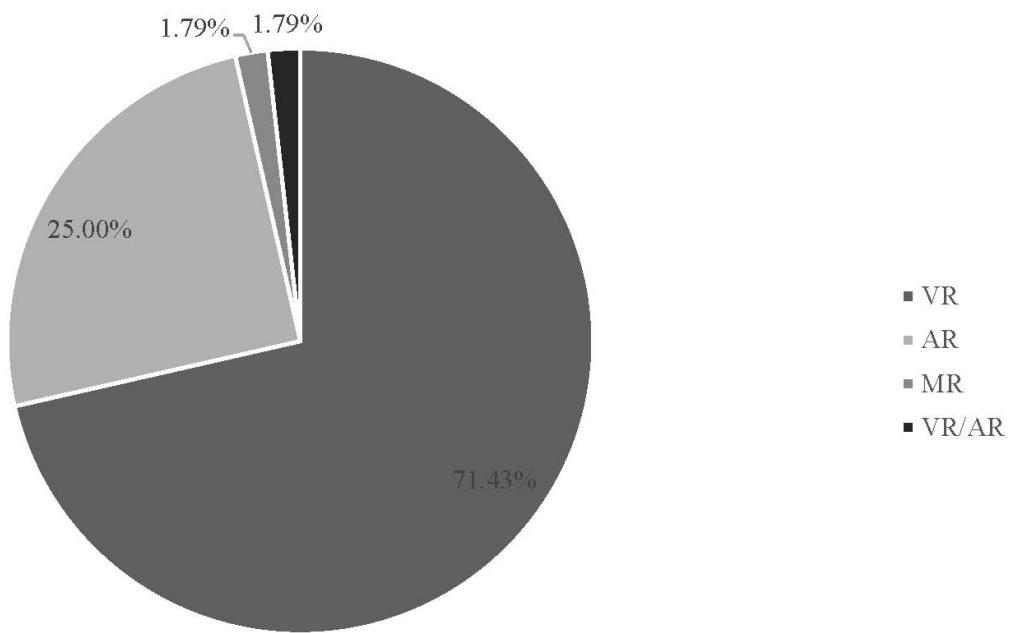
Author, Date	Learning Theory Mentioned	MERSQI + NOS-E score (max 24)	QUADAS score (max 11)
Aebersold et al., 2018	Situated learning theory	22	10
Andersen et al., 2021	No	21.5	11
Andersen et al., 2022	No	21.5	9
Arents et al., 2021	Cognitive load	15.5	6
Azimi et al., 2018	No	12.5	4
Banaszek et al., 2017	No	21	10
Bayram & Caliskan, 2019	NLN/Jeffries Simulation Theory	17.5	7
Blumstein et al., 2020	No	21	8
Bogomolova et al., 2020	Constructive Alignment Theory & Blooms	21	8
Bork et al., 2019	No	13.5	7
Brinkmann et al., 2017	No	17.5	8
Bube et al., 2020	Directed self-regulated learning (DSRL) theory; Simulation-based mastery learning	19	8
Butt et al., 2018	Deliberate practice theory	19.5	9

Cevallos et al., 2022	No	20	8
Chao et al., 2021	No	17.5	8
Chao et al., 2023	Cognitive load theory	21.5	8
Chen & Liou, 2023	No	19	7
Ekstrand et al., 2018	No	16.5	9
Fu et al., 2020	No	16.5	9
Haerling, 2018	NLN/Jeffries Simulation Theory	16.5	9
Han et al., 2021	No	16	7
Henssen et al., 2019	Cognitive load theory; Blooms	14.5	8
Hu et al., 2020	No	19	7
Issleib et al., 2021	No	20	3
Jaskiewicz et al., 2020	No	15	3
Jung & Park, 2023	No	16.5	9
Kane et al., 2022	No	21	5
Kowalewski et al., 2019	No	21	7
Küçük et al., 2016	Cognitive load theory	18.5	9
Lau et al., 2023	No	14	10
Lemke et al., 2020	No	18.5	10
Logishetty et al., 2018	No	20.5	9
Maresky et al., 2018	No	17	3
Moll- Khosrawi et al., 2022	No	20	8
Moro et al., 2017	Cognitive load theory	17.5	3
Nagayo et al., 2022	No	21	6
Neumann et al., 2019	No	21	4
Nielsen et al., 2021	No	19.5	8
Noll et al., 2017	No	20.5	9
Orland et al., 2020	No	21	4
Plotzky et al., 2023	No	21	8
Price et al., 2018	No	17	2
Ros et al., 2020	No	17.5	3
Ros et al., 2021	Kolb	20.5	4
Schoeb et al., 2020	No	20.5	10
Shao et al., 2020	No	19.5	3
Smith et al., 2018	No	19	9
Stepan et al., 2017	No	20.5	7
Sultan et al., 2019	Experiential learning theory	19.5	8
Watari et al., 2020	No	16	7
Wolf et al., 2021	No	15.5	7
Yang & Oh, 2022	No	19	7
Yeo et al., 2018	No	19	6
Yu et al., 2022	Cognitive load theory	18	6
Zackoff et al., 2020	No	17.5	9
Zhou et al., 2022	No	19	9

### 3. Results

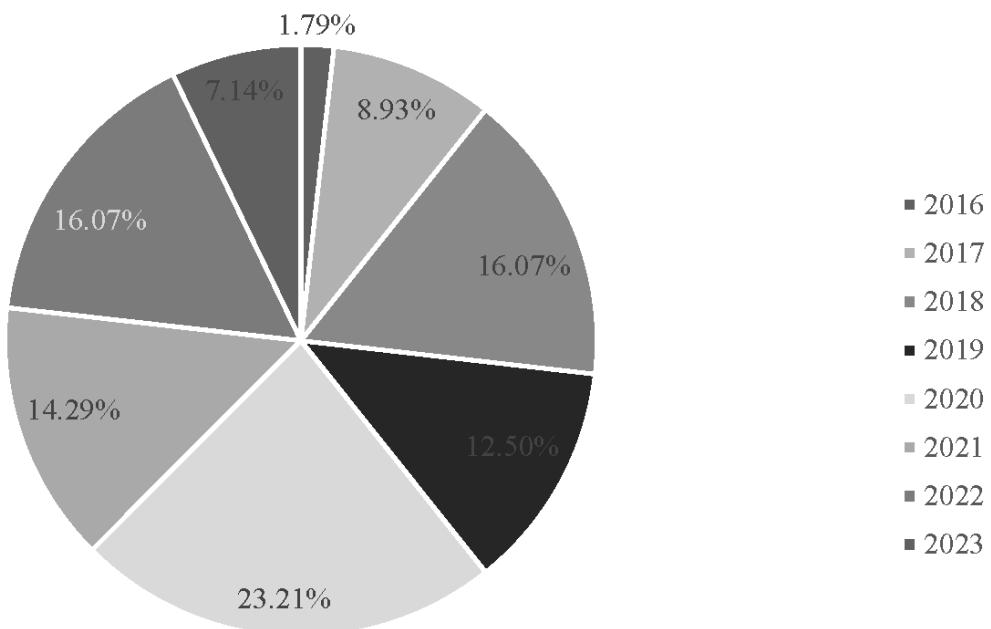
#### 3.1. Descriptive Findings

Of the 56 articles reviewed, 40 (71.43%) were VR studies, 14 were AR (25.0%), one study was an MR study, and one study was a VR/AR study because it was unclear whether the training was VR or AR based on the description of the apparatus (1.8%) (see Figure 3).



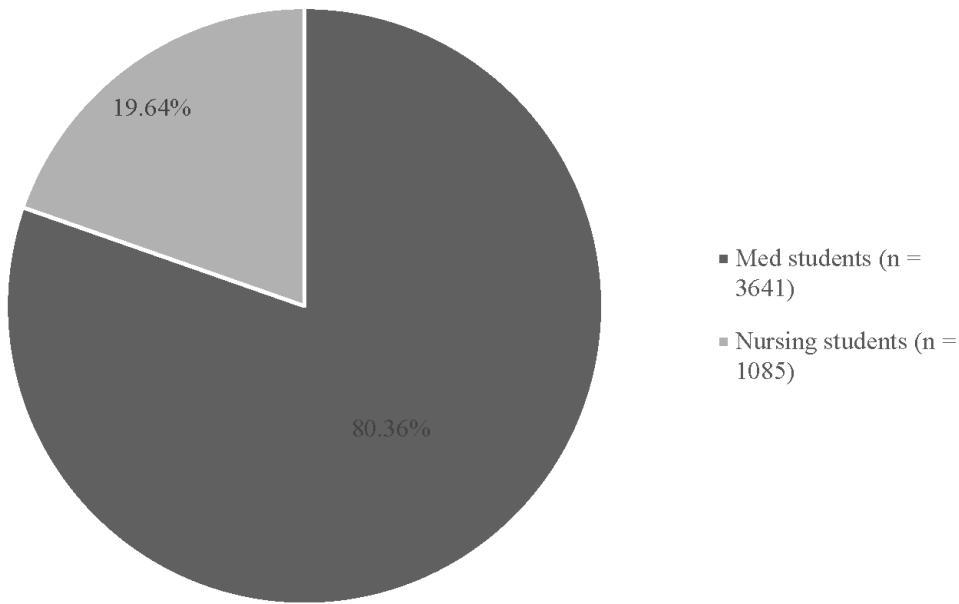
**Figure 3.** Percentages of Immersive Techniques Used (I-XR.).

Most studies were published in 2020 ( $n = 13$ , 23.2%) (Figure 4).



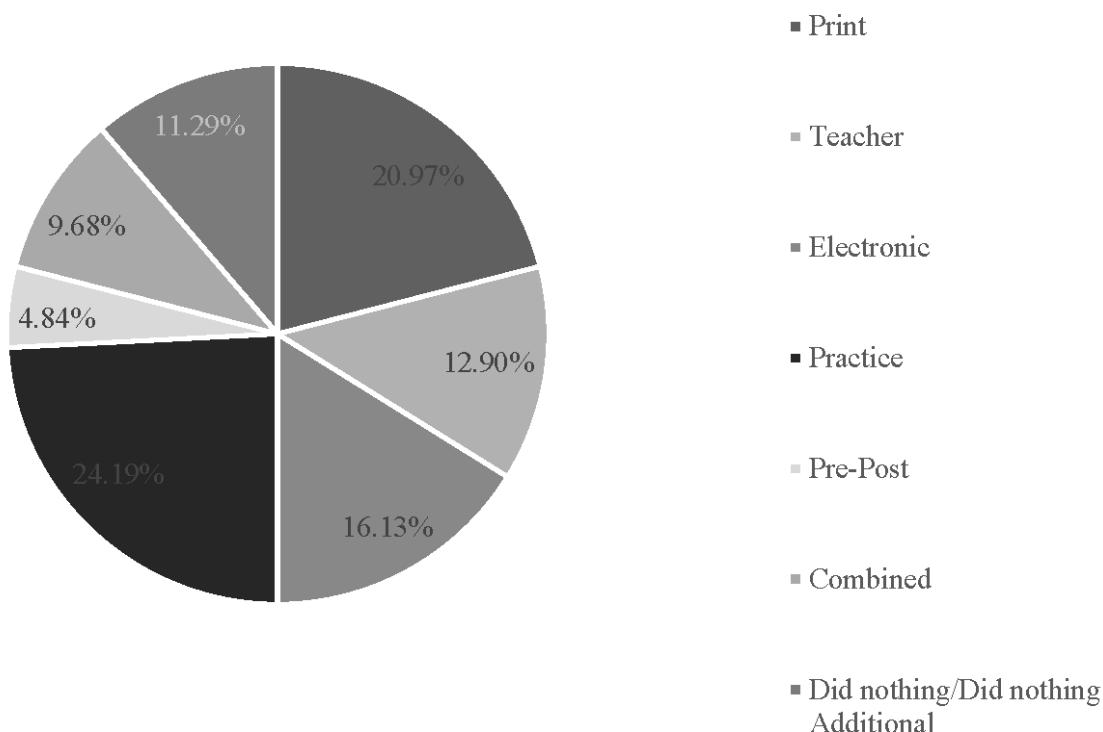
**Figure 4.** Percentage of Studies by Year.

80.4% (n = 45 studies) were performed on medical graduate students (n = 3641 individuals), compared to 19.6% of studies (n = 11) that were performed on nursing graduate students (n = 1085 individuals) (Figure 5).



**Figure 5.** Percentage of Student Type Reviewed.

The types of comparison techniques used across studies included: “print” (n = 13, 21.0%), “teacher” (n = 8, 12.9%), “electronic” (n = 10, 16.1%), “practice” (n = 15, 24.2%), “combined methods” (n = 6, 9.7%) , “pre-post” (n = 3, 4.8%), or “did nothing additional” (n = 7, 11.3%) (Figure 6).



**Figure 6.** Percentage of Types of Control Groups (n = 62 Control group across 56 Studies).

“Performance skills” variables across studies included: “time”, n = 13, 11.3%; “errors”, n = 5, 4.4%; “dexterity”, n = 0, 0%; “injury”, n = 1, 0.9%; “other specific/DOPS”, n= 46, 40.0% “skills knowledge’ (n = 28, 24.4%), “grade in internship” (n = 1, 0.9%), “performance measures” (n = 15, 13.0%), clinical judgment/reasoning skills” (n = 5, 4.4%), “problem-solving skills” (n = 1, 0.9%) (Figure 7). The average number of variables assessed in a study was two (range: 1- 12).

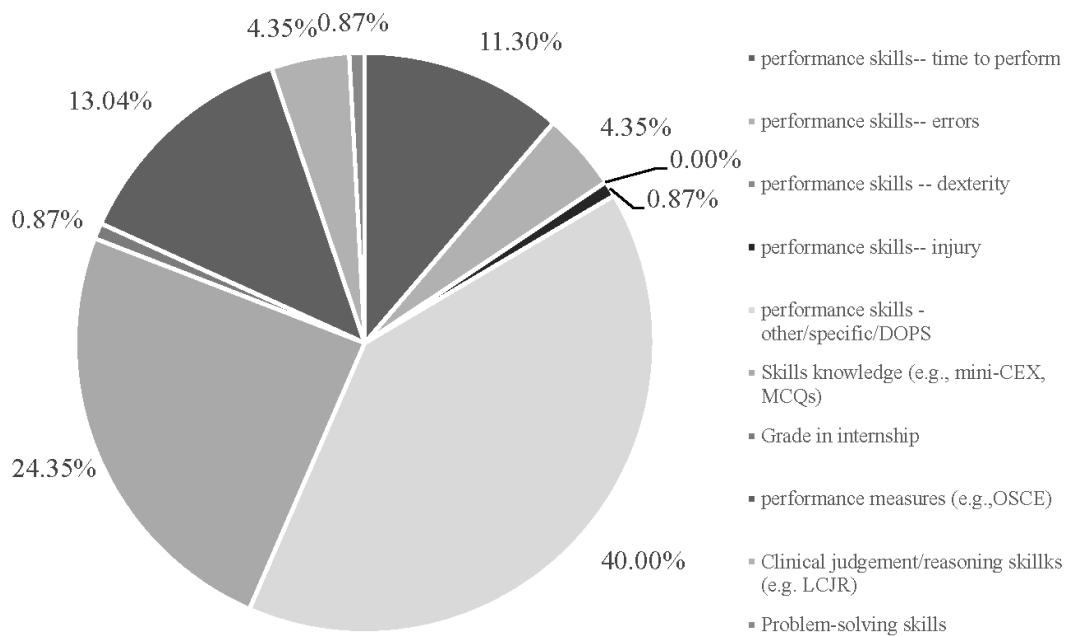
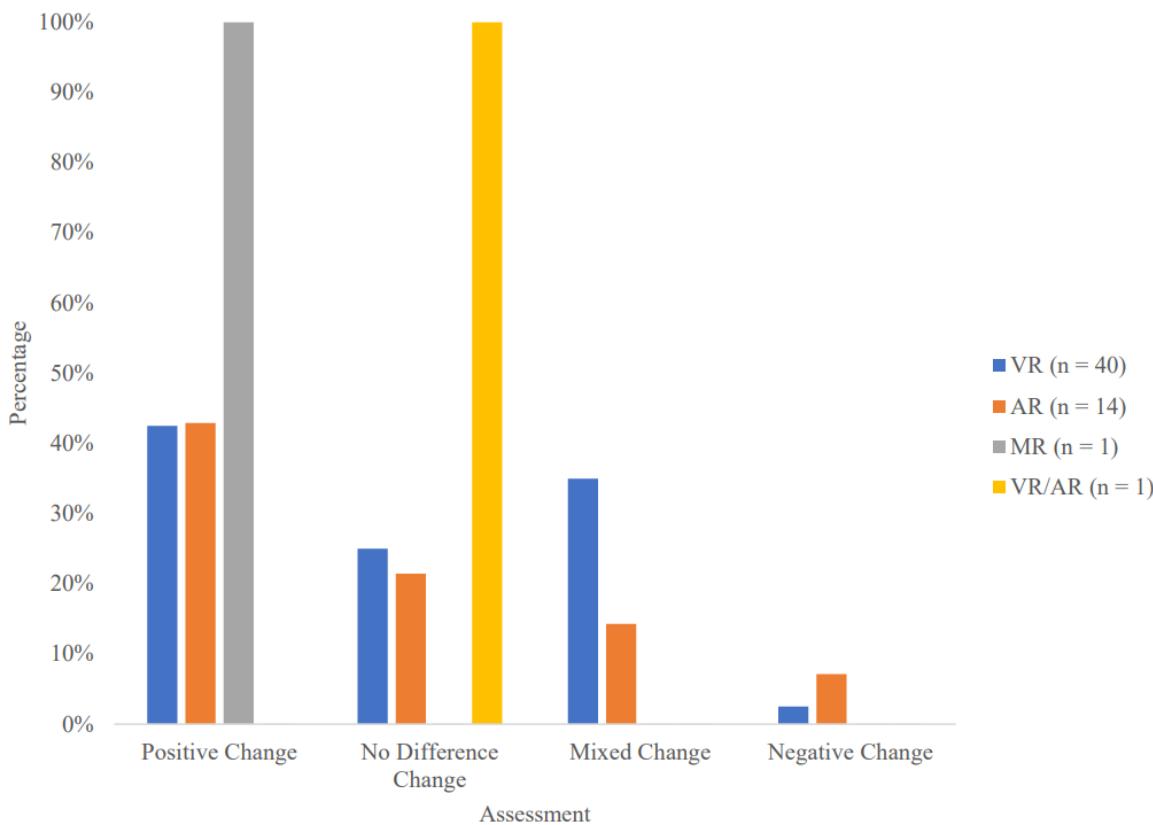
**Figure 7.** Percentage of Outcomes (115 over 56 Studies).

Table 2 also shows whether any learning theories were mentioned within the included studies. Approximately seventy-five percent of the studies did not mention any learning theory. Thirteen studies included one or more theories. The most commonly mentioned theory was cognitive load (n = 6). Simulation theory/NLN/Jeffries theory was mentioned three times, and self-regulated theory and Bloom’s theory were mentioned twice. Situated learning theory, directed self-regulated theory, simulation-based mastery learning, constructive alignment theory, deliberate practice, Kolb’s theory, and experiential learning theory were each mentioned once.

### 3.2. Overall Study Assessments of I-XR Skill Competency

Overall, we found that 42.5% of studies reported that VR was more effective than comparison (non-immersive) methodologies (Positive), 42.9% of AR studies were more effective than non-immersive methodologies (Positive), and the one MR (100%) was more effective than the comparison (Positive) (Figure 8). In addition, 25.0%, 21.4%, 0.0% of studies reported no difference between the VR, AR, MR group and the comparison groups, respectively (No difference). The one study classified as VR/AR also showed no change compared to the comparison (No difference). These percentages are compared to 35.0% (VR), 14.3% (AR) and 0.0% (MR) of studies showed mixed results (Mixed Evidence Positive or Mixed Evidence Negative). Finally, only two studies (one VR (2.5%) and one AR (7.14%)) favored the training comparisons (Negative).



**Figure 8.** Percentage of VR/AR/MR Studies by Outcome Success of Clinical Skills.

#### 4. Discussion

This review provides compelling evidence for the effectiveness of I-XR training in healthcare education. A substantial majority of studies (42.9%) demonstrated that I-XR methodologies led to universally improved outcomes compared to traditional training comparison methods. These percentages reflect improved outcomes on all measures, thus underscoring the findings' robustness. Only a small minority (3.6%) of studies universally favored comparison (non-immersive) approaches.

It is important, however, to acknowledge that 26.8% of I-XR studies showed mixed results (Mixed Positive, n = 10, 17.9%; Mixed Negative, n = 4, 7.1%; Mixed, n = 1, 1.8%). The effectiveness of I-XR is therefore not universal and may depend on various factors such as the specific technology used, the type of training, and the implementation context. Further research is needed to identify the conditions under which I-XR is most effective and to develop evidence-based guidelines for its optimal use.

A significant gap in the theoretical foundation of I-XR teaching and learning approaches was also revealed. Many studies lacked a clear articulation of the learning theories guiding their pedagogical strategies. This omission hinders the development of effective, evidence-based practices and may stem from a limited understanding of how best to teach and learn in these emerging digital environments. While I-XR offers exciting possibilities for simulation, interaction, and experiential learning, traditional pedagogical models designed for physical classrooms may not translate effectively to the virtual realm. For example, some studies have indicated challenges in maintaining student engagement, fostering collaboration, and ensuring equitable access in virtual learning environments (Zweifach & Triola, 2019).

These challenges underscore the need for further research to explore pedagogical approaches specifically for virtual learning environments. Future studies should prioritize the development of robust pedagogical frameworks specifically for I-XR, grounded in established learning theories. Additionally, more research is needed to determine the conditions under which I-XR training is most

effective, considering factors such as the type of technology, the learning objectives, and the characteristics of the learners. By addressing these critical areas, the full potential of I-XR can be unlocked, optimizing its use in healthcare education to enhance the training of future healthcare professionals.

#### 4.1. Limitations

While I-XR technology holds immense promise for revolutionizing healthcare education, a critical appraisal of the current landscape necessitates overcoming significant limitations. First, the existing body of research is hampered by methodological weaknesses, notably the prevalence of high bias in published studies. This bias, a well-documented phenomenon across academic research, may lead to an inflated perception of I-XR's positive effects, obscuring a clear and objective understanding of its true impact on learning outcomes and clinical performance.

Second, the inherent limitations of I-XR technology itself pose substantial barriers to widespread adoption. The acquisition and maintenance of I-XR equipment, software, and dedicated simulation spaces often entail significant financial investment, potentially creating disparities in access to this technology, particularly for institutions with limited resources.

Moreover, users may experience physiological side effects, such as cybersickness (including nausea, dizziness, and disorientation), eye strain, and headaches, which can hinder learning, diminish user engagement, and impede the seamless integration of I-XR into educational practices (Kolcun et al., 2023). Furthermore, the effective implementation of I-XR necessitates careful planning, technical expertise, and ongoing support. Challenges include the time and effort required for content creation, customization, and integration with existing curricula, as well as the need for dedicated technical support to troubleshoot issues and ensure smooth operation.

Addressing these limitations will require a multi-pronged approach. Future research should prioritize rigorous methodologies, including randomized controlled trials with well-defined comparison groups and objective outcome measures, to mitigate bias and provide a more robust evidence base for I-XR's efficacy. Technological advancements are urgently needed to reduce costs, enhance user comfort, and minimize physiological side effects, making I-XR more accessible and user-friendly.

Finally, fostering collaborative partnerships between educators, technology developers, and institutional stakeholders is crucial to addressing logistical challenges, developing best practices for implementation, and facilitating the seamless integration of I-XR into healthcare education. By acknowledging and proactively addressing these limitations, we can pave the way for the successful and impactful integration of I-XR, ultimately transforming healthcare education and improving patient care.

#### 4.2. Future Directions

The future of healthcare education stands at the precipice of a transformative era fueled by the remarkable potential of I-XR technologies. These technologies, encompassing VR, AR, and MR, offer unprecedented opportunities to create engaging and effective learning experiences that bridge the gap between theoretical knowledge and practical application. However, successfully integrating I-XR into healthcare education requires a thoughtful and comprehensive approach beyond mere technological adoption. It demands a fundamental shift in pedagogical thinking, prioritizing the application of established learning theories, such as experiential learning and embodied cognition, to guide the design and implementation of I-XR curricula. Experiential learning emphasizes the importance of active engagement and hands-on experiences in the learning process, while embodied cognition highlights the interconnectedness of mind and body, suggesting that learning is enhanced when it involves physical interaction and sensory immersion (Macrine & Fugate, 2022). By grounding I-XR experiences in these theoretical frameworks, educators can create immersive and impactful learning environments that foster deep understanding, critical thinking, and skill acquisition.

Furthermore, the effective integration of I-XR necessitates a collaborative ecosystem where medical education leaders, faculty, and technology developers work in concert. This collaboration should focus on establishing a shared vision for I-XR integration that aligns with the goals of healthcare education, providing comprehensive faculty training to empower them to adapt their curricula and pedagogy, and encouraging active faculty participation in the development of future I-XR applications. To ensure consistency and quality, standardized methods for educating, assessing, and certifying I-XR instructors are essential.

While dedicated virtual simulation centers currently represent a significant step towards realizing the potential of I-XR in healthcare education, the future points towards increased accessibility and versatility. Imagine a future where I-XR transcends the confines of specialized centers, seamlessly integrating with mobile devices and wearable technology. This would empower healthcare professionals with a powerful toolkit readily available at their fingertips, enabling them to access course content, patient information, and real-time patient data during rounds using wearable devices or smartphones.

Problem-based learning, a cornerstone of medical education, could be further enhanced through I-XR, with virtual patients and dynamic case scenarios providing a safe and engaging environment for students to hone their clinical reasoning and decision-making skills. As I-XR technology becomes more affordable and accessible, collaborations will drive the development of innovative applications tailored to the diverse needs of individual learners and healthcare disciplines, fostering personalized learning experiences.

The future of I-XR also holds the promise of even richer and more immersive experiences. Haptic feedback devices will enable students to simulate procedures with unparalleled realism, developing muscle memory and refining their dexterity. Realistic visuals and soundscapes will transport students to diverse clinical settings, from bustling emergency rooms to serene operating theaters, fostering a deeper understanding of contexts and enhancing their ability to adapt to different environments. The key to unlocking the full potential of I-XR lies in integrating learning theory into I-XR experiences and creating immersive and impactful learning environments that optimize teaching, learning, and evaluation. This, coupled with ongoing innovation and collaboration, will ensure that healthcare education remains at the forefront of technological advancement and pedagogical excellence.

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