

ARTIFICIAL INTELLIGENCE IN CLINICAL MEDICAL EDUCATION

Enhancing the Student and Preceptor Experience

Michael A Edwards DO
LMU-DCOM Orange Park
April 17, 2026



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4th Friday Preceptor Development Series

DISCLOSURES: NONE*
CONFLICTS: NONE



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- OBJECTIVE: 1. NAME AND DEFINE LEARNING THEORIES THAT AFFECT MEDICAL EDUCATION**
- 2. IDENTIFY THE OPPORTUNITIES OF AI AS IT RELATES TO TEACHING MEDICAL STUDENTS**
- 3. RECOGNIZE THE PITFALLS OF AI AS IT RELATES TO TEACHING MEDICAL STUDENTS**
- 4. DEVELOP STRATEGIES TO UTILIZE AI AS IT RELATES TO TEACHING MEDICAL STUDENTS**

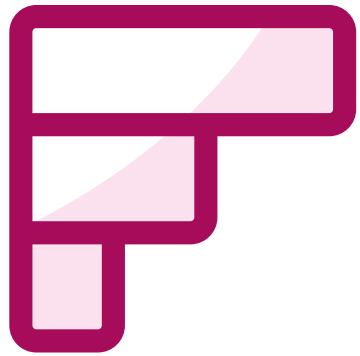


What challenges are you facing to educating students?

CHANGING LANDSCAPE OF MEDICAL EDUCATION



- Explosion of medical knowledge
- Students trained in digital-first environments
- Increasing documentation burden
- Less bedside teaching time



What kind of learner are you?

LEARNING THEORY IN MODERN MEDICAL EDUCATION



- Constructivism: learners build knowledge through experience
- Cognitive apprenticeship: modeling expert reasoning
- Experiential learning (Kolb)
- Self-directed learning (Knowles)

EXPERIENTIAL LEARNING (KOLB) VS SELF-DIRECTED LEARNING (KNOWLES)

Kolb – Experiential Learning

Learning through experience and reflection

4-stage cycle: Experience → Reflect → Conceptualize → Experiment

Teacher designs meaningful experiences

Common in simulations, clinical rotations, case learning

Knowles – Self-Directed Learning

Adults take responsibility for their learning

Learner identifies needs and goals

Finds resources and learning strategies

Evaluates progress independently

LINKING KOLB & KNOWLES TO AI-ASSISTED LEARNING

- AI accelerates reflection in Kolb's cycle by providing immediate feedback after clinical encounters or simulations.
- AI tools help learners identify knowledge gaps, supporting Knowles' self-directed learning model.
- Adaptive AI tutoring systems personalize learning pathways based on learner performance.
- Clinical learners can simulate cases with AI, reflect, refine concepts, and re-apply knowledge quickly.
- AI acts as a 'cognitive coach,' augmenting metacognition and lifelong learning habits.

INTEGRATED LEARNING FRAMEWORK: KOLB + KNOWLES + MASTER ADAPTIVE LEARNER

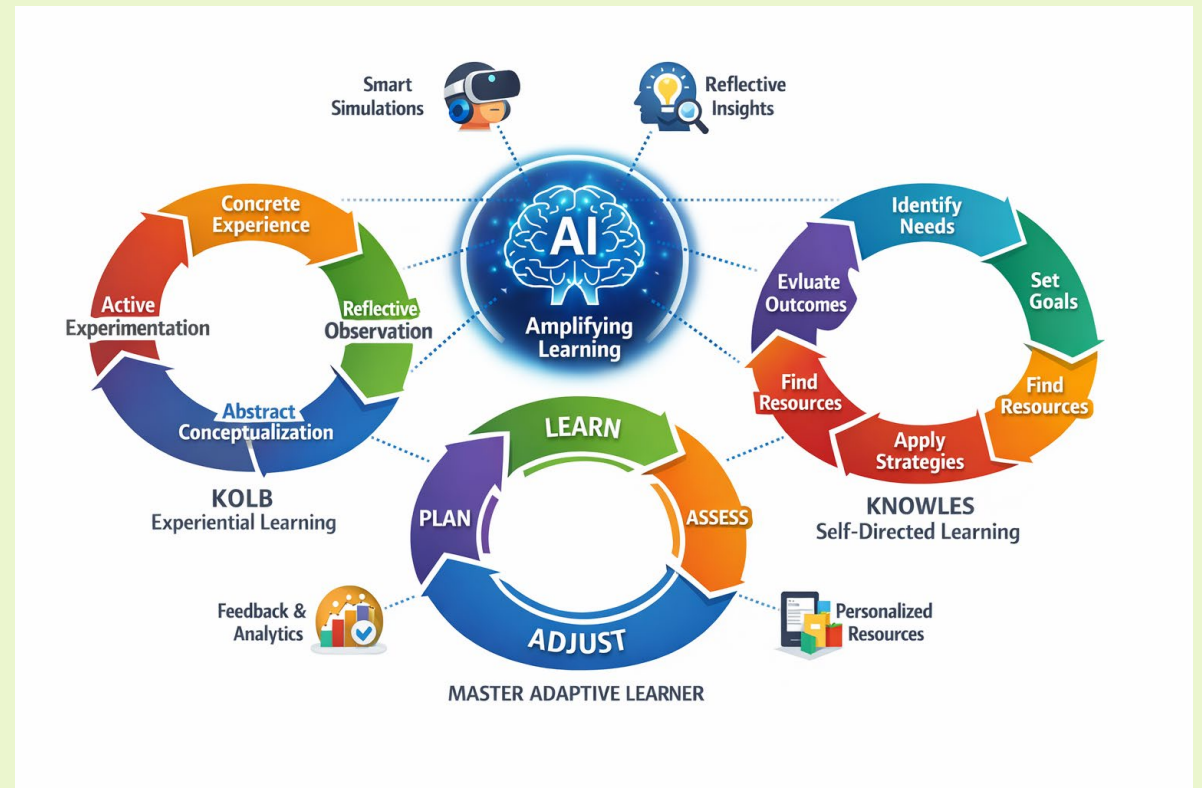
Master Adaptive Learner (MAL): Plan → Learn → Assess → Adjust

Knowles contributes learner autonomy: identifying learning needs and goals.

Kolb contributes the experiential cycle: experience, reflection, conceptualization, experimentation.

AI enhances each stage: gap analysis, adaptive resources, simulation practice, and performance feedback.

Together they create a continuous professional learning loop for clinicians.





How comfortable are you using any form of AI?

LARGE LANGUAGE MODELS



- Neural networks trained on large text corpora
- Generate explanations, summaries, and reasoning dialogue
- Examples: ChatGPT, Claude, Gemini, Llama, Mistral
- ❖ Pitfalls: they are language models not "non-cognitive processors" nor "probabilistic engines"

CLINICAL EVIDENCE AI TOOLS



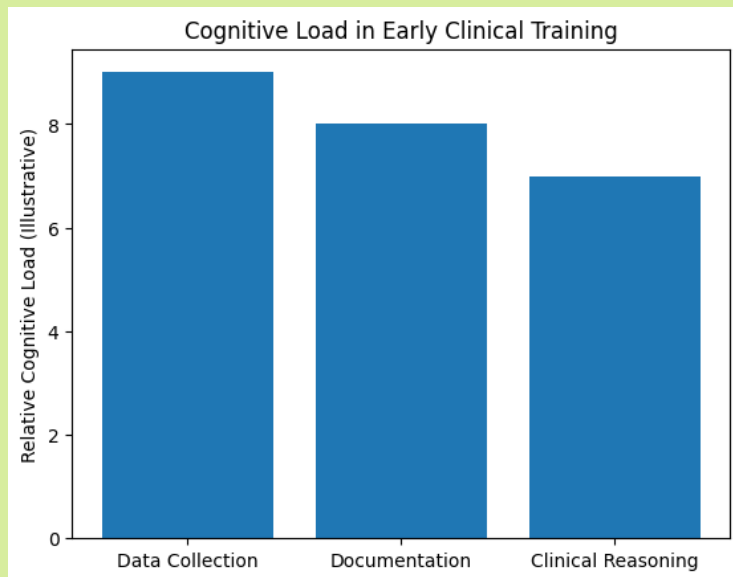
- OpenEvidence: evidence synthesis assistant
- Doximity AI: workflow and documentation support
- Overlapping End User Experience
- Differing roles in clinical education

AI TRANSCRIPTION IN CLINICAL TRAINING



- Gain: Automated note generation
- Gain/Pitfall: Clinical encounter summarization
- Gain/Pitfall: Reduced documentation burden
- Gain: Improved learner focus on patient interaction
- ❖ Pitfalls:
 - ❖ overreliance on tools over skills
 - ❖ AI transcription Hallucination
 - ❖ tendency to include data that is not accurate
 - ❖ increased clinician review

COGNITIVE LOAD IN EARLY CLINICAL TRAINING



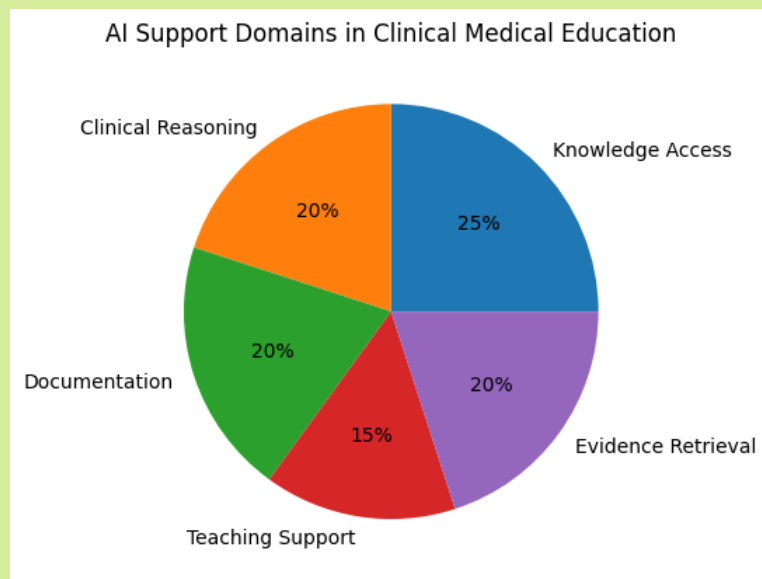
- Students must learn data collection and interpretation simultaneously
- Documentation requirements increase cognitive load
- AI tools can reduce transcription burden

NOTEBOOKLM AND AI LEARNING TOOLS



- AI-powered research assistant
- Generates summaries, quizzes, and study guides
- Grounds answers in uploaded documents
- Supports active learning

AI SUPPORT DOMAINS IN CLINICAL EDUCATION



- Knowledge retrieval
 - Clinical reasoning development
 - Documentation support
 - Teaching augmentation
 - Evidence synthesis
- ❖ Pitfall: students may skip the part where they learn how to do all of the above and may develop a false sense of skill

CLINICAL CASE EXAMPLE 1 – CHEST PAIN



- Patient: 54-year-old male with exertional chest discomfort
- Student obtains history and physical and then asks AI for differential diagnosis
- While AI generates prioritized differential list
- ❖ Preceptor reviews the case and verbalizes a differential list
 - ❖ Student and preceptor critiques AI reasoning

CLINICAL CASE EXAMPLE 2 – DIABETES MANAGEMENT



- Patient with uncontrolled Type 2 diabetes
- Student performs H&P and recognizes poorly managed disease then queries AI for guideline-based therapy
- Compare response with ADA recommendations
- Discuss medication selection and comorbidities

CLINICAL CASE EXAMPLE 3 – DIAGNOSTIC REASONING



- Young woman with fatigue and anemia
- Student explores differential with AI
- AI suggests iron deficiency, chronic disease, B12 deficiency
- Student uses clinical data to refine diagnosis

RISKS AND LIMITATIONS



- ❖ Hallucinated information
- ❖ Overreliance by learners
- ❖ Privacy considerations
- ❖ Need for critical appraisal
- ❖ Malpractice implications of we abdicate responsibility

FUTURE OF AI IN MEDICAL EDUCATION



- Personalized AI tutoring
- Real-time clinical evidence retrieval
- Improved clinical simulations
- Expanded teaching capacity

CURRENT USE OF AI IN MEDICAL EDUCATION

*DISCLOSURES:



- Most of this presentation was created with ChatGPT via entering prompts
- Substantial reduction in burden of composition
- Allowed for rapid retrieval of models and rough draft writing
- Revisions correct and elevate to personal
- Final Draft was proofread by Gemini



**Please share your thoughts on
utilizing AI in the future of Medical
Education?**

THANK YOU

"Mickey" Michael A Edwards DO
Michael.edwardso2@Imunet.edu

Citations:

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