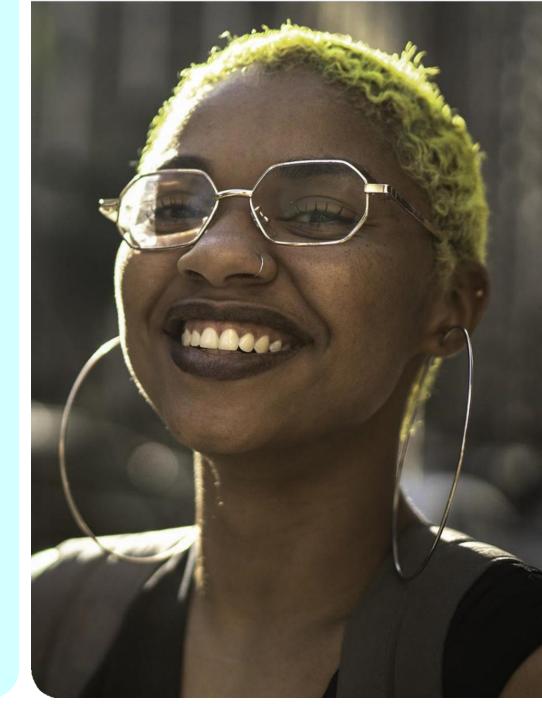
¬PressGaney **Using Artificial** Intelligence to **Improve Safety: Risks and Opportunities**

Prepared by

Tejal Gandhi, MD, MPH, CPPS Chief Safety & Transformation Officer



Caveats

- I have a long history in patient safety research and operations
- -I have studied and implemented a wide range of clinical decision support tools in EHRs
- -I am constantly learning, as all of you are, about AI and generative AI
- -I am not an expert in AI algorithms!

Al is not new



Artificial intelligence (AI) is any form of intelligence shown by a machine, which resembles natural (human) intelligence such as planning, learning, problem solving, etc.



Machine learning (ML) is the ability of machines to predict outcomes by mapping the outcome to representation



Deep learning (DL) is the ability of machines to predict an outcome by learning the representation of the outcome

General (Strong) AI is AI that exhibits human-level intelligence

		Narrow (Weak) Al	Super Al is Al that surpasses human intelligence and ab	ility.
1950	1980	2010 Now 204	40? 2070?	nvidia.com

What is AI?

- A machine-based system that can, for a given set of human-defined objectives, make predictions, recommendations, or decisions influencing real or virtual environments.



The Artificial Intelligence Spectrum

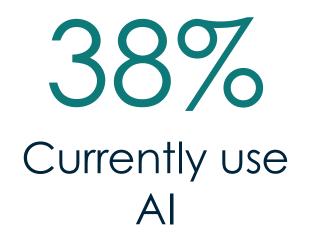
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Slide courtesy of Dr. Raj Ratwani

AI in healthcare

1 in 4 dollars invested in healthcare going toward companies leveraging AI

- Silicon Valley Bank report





definitivehc.com

AI in healthcare







27%

Computer-aided image detection for disease states



25% Care guideline consultation/ suggestive care options

definitivehc.com

The safe, secure, and trustworthy development and use of artificial intelligence

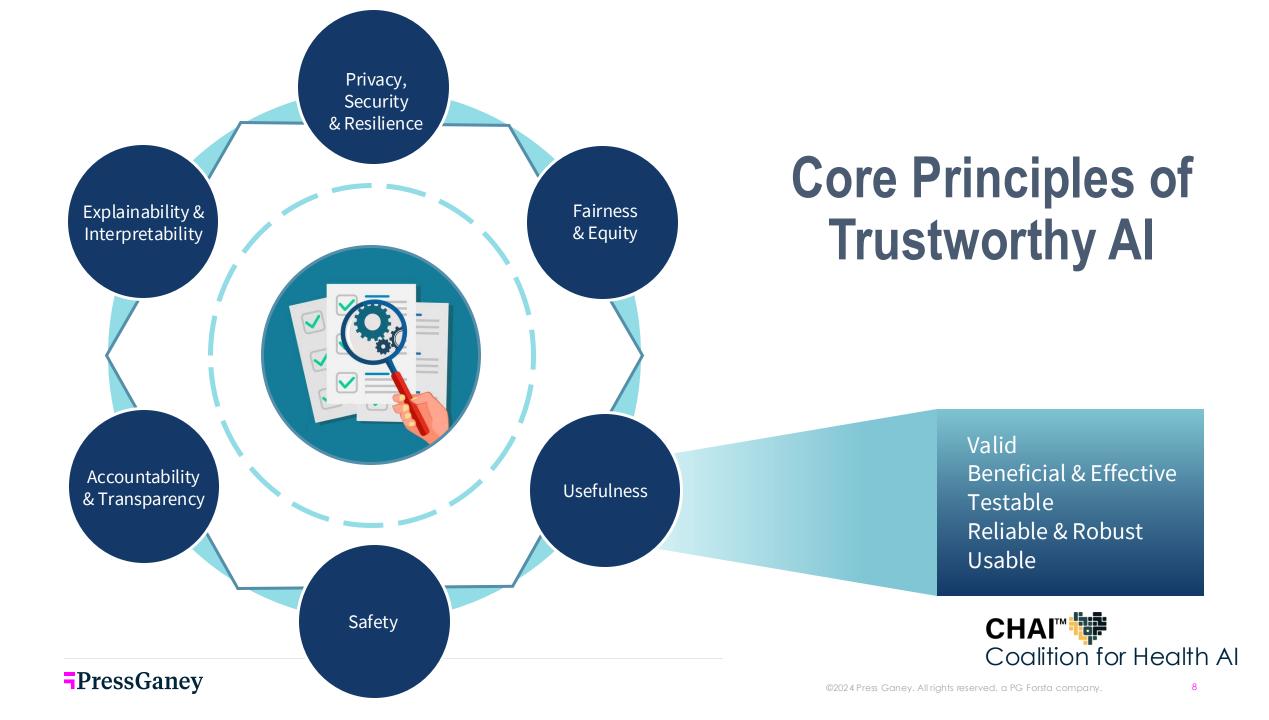
PRESIDENT BIDEN ISSUES EXECUTIVE ORDER (EO)

- Directs federal agencies to create standards and regulations for the use and oversight of AI across sectors like energy, infrastructure, criminal justice, healthcare, housing, education, and labor, with a focus on protecting civil rights and liberties.
- Guiding principles are reflected across several sections:
 - Ensuring the safety and security of AI technology
 - Promoting innovation and competition
 - Supporting workers
 - Advancing equity and civil rights
 - Protecting consumers, patients, passengers, and students
 - Protecting privacy
 - Advancing federal government use of AI
 - Strengthening American leadership abroad

Federal agencies to watch

- National Institute of Standards and Technology (NIST) within the Department of Commerce: tasked with developing guidelines and best practices for AI safety and security
- The Department of Health and Human Services (HHS): responsible for new strategy and safety programs for Al in healthcare programs
- Office of Management and Budget (OMB) coordinates interagency activity and use of AI in government

"Harnessing AI for good and realizing its myriad benefits requires mitigating its substantial risks. This endeavor demands a society-wide effort that includes government, the private sector, academia, and civil society."



Questions to ask:









Oversight Criteria

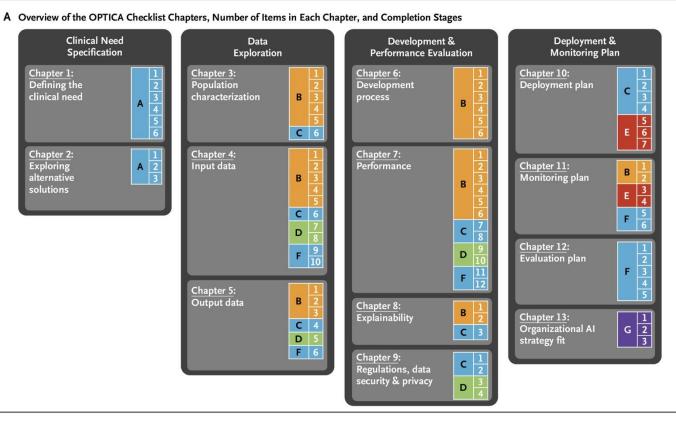
Transparency Regulatory **Clinical Value Usability &** Fairness & & Accountability Compliance Adoption & Safety Algorithmic tools have defined Algorithmic technologies have Algorithmic tools are fair, Algorithmic tools adhere to Quality & Algorithmic tools improve clinical and technical clinical care (e.g., patient and biases are managed regulatory requirements clear clinical and technical **Ethical** such that impact of tool ownership, and are reliable and best practices ownership. Relevant outcomes, patient Implementation Guide and usable wihin the information on development, satisfaction), clinician is equitable. throughout the lifecycle. Principles designated clinical workflow. validation, and use are experience (e.g. burnout), documented, available and or healthcare operations, while being safe. accessible to stakeholders depending on their role. Evaluation Criteria Submission Material Evaluation Process Algorithmic Model Effectiveness Silent General Deployment **Tool Lifecycle** Development Evaluation Evaluation

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Duke University Health System; Economou-Zavlanos N. JAMIA brief communication 2023

Evaluation of AI Solutions in Health Care Organizations — The OPTICA Tool

Dagan et al. NEJM AI 2024



B Description of the Completion Stages (Marked by Capital Letters) and the Stakeholders who Participate in Completing Each Stage (Marked by Colors)

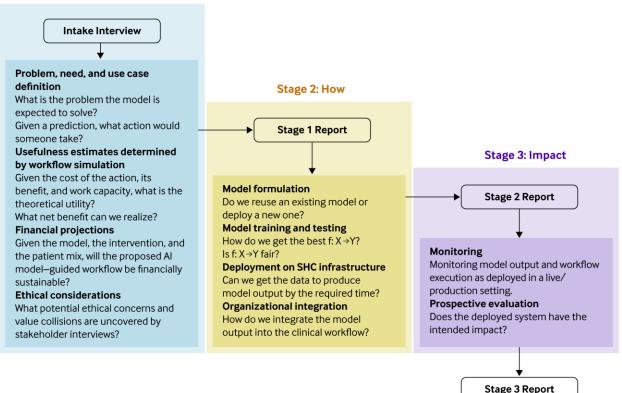
Clinical Expert	AI solution Developer	Clinical Expert	Organizational Data Lead	MLOps Expert	Clinical Expert	Organizational AI Lead
A	• •	• •		• •	• •	• G
Defines the clinical need and the performance requirements in detail + explores alternative solutions	Describes the solution "under the hood," including aspects such as the training data, algorithmic methods, performance metrics, and explainability	Assesses items completed by the developer to estimate the solution's appropriateness from a clinical perspective + given satisfactory results, completes an initial deployment plan	Evaluates the data availability in the organization + provides a comparison to the training data + assesses the privacy and cybersecurity compatibility	Refers to the technical aspects of model deployment, real-time data and server availability + plans the monitoring to ensure model stability	Provides a final assessment of the solution's suitability to the local data and requirements given all prior completed items + provides a postdeployment plan for ongoing updates and evaluation	Provides an overa assessment of th checklist answers provides recommendation in view of the organizational A strategy

FIGURE 1

The Fair, Useful, and Reliable AI Model Assessment Process

Each Fair, Useful, and Reliable AI Model (FURM) assessment consists of three stages: evaluating the what and why behind a particular AI use case; how a given AI model will be formulated, evaluated, and integrated into a given health care system workflow; and the impact of the proposed implementation. Each of these three stages has multiple components (as shown in the boxes), each addressing a different aspect of the use case. After each of the three stages, a report is prepared, which recommends whether or not to proceed.

Stage 1: What and Why



AI = artificial intelligence, SHC = Stanford Health Care.

Note: f: $X \rightarrow Y$ denotes a function that maps some values of X to some values of Y.

Source: The authors

NEJM Catalyst (catalyst.nejm.org) © Massachusetts Medical Society

2024

Standing on FURM

Ground: A Framework

Health Care Systems

Callahan et al. NEJM Catalyst

forEvaluating Fair, Useful,

and Reliable AI Models in

Safe Use

UukeHealth

- LLMs can produce false and/or inaccurate answers while sounding very convincing; this is called "hallucination." In other words, LLMs can be wrong and patient injury or other problems may result.
- LLMs have been shown to propagate social biases and inequities.
- LLMs can "forget" earlier parts of your interaction with them and behave erratically.

Principles of Appropriate Use

- Duke Health employees must carefully review, verify, and take responsibility for any LLMgenerated text or actions that result from it. Even small errors could have serious consequences if not caught.
- 2. LLMs should not be used in situations where the user is not adequately trained to, unwilling to or otherwise unable to validate the output. Adequate training is defined by the knowledge and ability required to understand all the LLM's output at an expert level and make changes such that the result is factually and medically accurate and appropriate. Duke Health employees are fully responsible for validating LLM output that they use in their work.
- 3. Duke Health employees should only use Duke-approved LLMs such as Bing Chat Enterprise for tasks related to clinical activities. Duke Health employees must not enter protected health information (PHI) or Duke Health intellectual property into LLMs unless specifically authorized (Figure 1). PHI is any information that was created, used, or disclosed while providing a health care service such as diagnosis or treatment that can be used to identify an individual. Intellectual property is work like business plans, contracts, ideas, patents, and other creative work produced at Duke that should not be shared outside of Duke.
- Duke Health employees must not use LLMs as the primary basis of clinical diagnostic or therapeutic decisions. LLMs should be used as an adjunct to standard tools such as UpToDate and PubMed, not as a replacement.
- 5. Any patient-facing content that was drafted by LLMs for human review should have a disclosure statement informing patients that an LLM was used to generate the content. e.g. This message was auto-generated and carefully reviewed for accuracy by Dr. Erica Smith

Courtesy of Dr. Eric Poon

Questions to ask:









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AI & Its Role in Patient Safety: Specific Safety Issues

npj | Digital Medicine

www.nature.com/npjdigitalmed

Check for updates

REVIEW ARTICLE OPEN

The potential of artificial intelligence to improve patient safety: a scoping review

David W. Bates (1)^{2,3 SI}, David Levine^{1,2}, Ania Syrowatka (1)^{1,2}, Masha Kuznetsova⁴, Kelly Jean Thomas Craig (1)⁵, Angela Rui¹, Gretchen Purcell Jackson (1)^{5,6} and Kyu Rhee⁵

Majority of use cases are prediction models- ADEs, sepsis, falls, pressure injuries, surgical complications, decompensation; key to tie into workflow and action

Patient safety domain	Likelihood of impact
Healthcare-associated infections	Al may have a moderate impact on the reduction of HAIs given that current evidence-based practices are already effective when applied well.
Adverse drug events	Al can play a major role in ADE prevention. As more patients at risk of ADEs are accurately identified before a medication is administered or prescribed, a greater proportion of these events will become preventable. However a key challenge lies in the lack of integrated high-quality datasets in which ADEs have been accurately captured. A variety of automated approaches have been effective at identifying patients likely to have experienced an ADE, but typically clinician adjudication is still required. ML could also help identify patients who may benefit from additional testing for specific single nucleotide polymorphisms to guide optimal drug therapy. These methods may also help to identify signals from the remainder of the genome beyond single nucleotide polymorphisms which may have prognostic impact.
Venous thromboembolism	We believe that AI will have a moderate effect on the reduction of VTE, as current evidence-based preventive strategies are already effective. AI solutions could provide further insights by identifying patients who could benefit from diagnostic testing for inherited thrombotic disorders to inform management of their condition.
Surgical complications	Al can be expected to have a moderate impact on the prediction and prevention of surgical complications both in the operating room and during recovery. Most complications felt to be preventable today are related to delayed diagnoses or intervention, technical issues, and infections. Given the overlap with other harm domains, focusing or advances in these other areas will likely also improve surgical safety.
Pressure ulcers	Pressure ulcers represent an attractive target with moderate to high potential for AI to prevent harm. Novel data sources such as motion and fluid sensors are now available, and large numbers of traditional clinical variables can be combined with the sensing data to predict who is at risk to guide evidence-based prevention.
Falls	Al is anticipated to have a moderate impact on fall prevention, given that this area has already received substantia attention and current risk mitigation strategies are effective. As with pressure ulcers, clinical data combined with sensing data can be used to predict when falls may occur, and which ones are likely to be associated with the most harm.
Decompensation	Leveraging novel data sources and AI has high potential to improve the prediction of decompensation to guide preventive strategies as well as early intervention to mitigate the impacts including premature death, given that current approaches are not effective. Given the serious nature of these events, preventing decompensation is a particularly attractive target. ML can deeply analyze data, beyond the standard values of heart rate or heart rate variability and will be critical to improving detection of decompensation and subsequent intervention.
Diagnostic errors	Diagnostic error is the most complex of the eight harm domains with vast opportunities for improvement using novel data sources and AI. ML could help to reduce the frequency of diagnostic errors by leveraging pattern recognition, bias minimization, and infinite capacity, areas where diagnosticians often falter. Although this area has gamered a lot of attention, many outstanding challenges remain, and current solutions only address a small fraction of what is possible. Most crucial to constructing valuable ML algorithms that help to reduce diagnostic errors.

AI & Its Role in Patient Safety: Decreasing Cognitive Burden

JAMIA Open, 2023, 6(3), ooad079 https://doi.org/10.1093/jamiaopen/ooad079 Perspective

Perspective

How can artificial intelligence decrease cognitive and work burden for front line practitioners?

Tejal K. Gandhi, MD, MPH^{1,*}, David Classen, MD, MS², Christine A. Sinsky, MD³, David C. Rhew, MD⁴, Nikki Vande Garde, BSN⁵, Andrew Roberts ^(b), PhD⁶, Frank Federico, RPh⁷

- Data gathering
- Data synthesis
- Documentation
- Taking appropriate action
- Structure asynchronous activities, outcomes and activities

Table 1. Clinical and cognitive processes potentially impacted by Al.

Function	Description	Example
Data gathering.	AI-based tools such as natural language processing (NLP), large language models, and image recognition can help clinicians comb through large volumes of informa- tion sources and discrete or unstructured data to help locate, identify, and surface the most relevant pieces of information, as well as missing information, to reduce cognitive and work burden on clinicians.	Data gathering: Using NLP, clinicians can more effi- ciently search genomic and clinical trial databases medical literature, and other sources of informa- tion to rapidly identify recommended treatments specific to cancer patients. Data gathering and synthesis: Solutions identify and compile both structured and unstructured EHR data and present that information to the provider to identify potential missing diagnoses, diagnoses that lack elinical evidence. Presenting this informa- tion directly to providers within their workflow eliminates the need to address manual queries hours, days, or weeks later; thus, reducing the cog nitive burden on the provider.
Data synthesis. rapidly collect, organize, manage, and making sense of the datasets from clinical assessments, physiologic observations, and documentation.	AI may be used to help support a new clini- cal workflow (eg, command center; rapid response team) or to improve an existing one (eg, prioritize existing care manage- ment outreach based on risk)	Data gathering and synthesis: The future of health- care includes the use of wearable devices and remote monitoring which will present clinicians with an overwhelming amount of information to sift through to make a clinical decision. AI can help filter through data, identifying critical data points or information that may indicate a change in a patient's condition or sudden deterioration. A can also suggest treatments based on scientific evi- dence, as well as offer customization based on the patient's condition, ability to adhere to treatment, and personal desires.
Documentation.	AI may be used to create or enter a summary of or provide specific details regarding a patient encounter into an EHR or other system of record using technology. AI can help simplify the billing documentation process, and generative AI can help reduce documentation burden.	Documentation: Ambient clinical intelligence (ACI) allows patients and clinicians to engage in conver- sation without the clinician needing to focus his/ her time on a keyboard or screen. When used in ACI, AI helps identify the speaker, uses NLP to convert the voice conversation into text, maps the medical terms and phrases to standardized nomen clatures, and organizes the conversation into a properly formatted clinical note, that can then be reviewed by the clinican for accuracy and rele- vance. Once finalized, the note is integrated into the EHR. Voice can also be leveraged to navigate and act within the EHR.
Taking appropriate action.	AI can provide decision support, prediction tools, targeted outreach, and guidance in response to signals or deploy an "intelligent" command center to help manage populations	Data gathering, synthesis and taking action: In 2018 Ochsner Health System (New Orleans, LA) lever- aged a pilot program to redesign care models at several campuses to include virtual nursing. ^{14,15} With trained clinicians working from remote com- mand centers or virtual bunkers, there is opportu- nity to leverage Al to remove interruptions for nurses and physicians at the patient's bedside, and to centralize Al and ML model monitoring and management while allowing associated interven- tions to continue to be carried out at the bedside. <i>Taking action</i> : Clinical decision support, particularly related to medication prescribing, has been plagued with issues related to over-alerting, contri buting to cognitive burden. In a recent study, researchers at one academic medical center found that ML techniques could enable intelligent filter- ing of medication alerts and reduce alert volume by 54%. ¹⁶
Structure asynchronous activities, outcomes and activities.	The care team's clinical and operational activities, including patient rounds and communication, are unstructured and asynchronous. AI can help manage coor- dination and synthesis of the information for updates and the next best course of	Data gathering, synthesis and taking action: A com- puter assisted management program for antibiotic and other anti-infective agents showed improved patient and antibiotic outcomes. ¹⁷

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action



From: Using Artificial Intelligence to Improve Primary Care for Patients and Clinicians

JAMA Intern Med. Published online February 12, 2024. doi:10.1001/jamainternmed.2023.7965

Table. Potential Use Cases for AI in Primary Care		
Use case	Examples of AI role	
Inbox management	Prioritize patient messages	
	 Generate draft responses 	
	• Edit physician messages to optimize communication, including for literacy appropriateness	
Clinician	With transcription software:	
documentation	 Draft progress notes in real time during visits 	
	 Draft prior authorization, disability, and durable medical equipment requests 	
	 Draft a list of billing codes for visits 	
Between-visit panel management	• Accurately identify patients in need of cancer screening using unstructured and structured EHR data to determine exclusions	
	• Identify patients with incomplete cancer screening (such as missed appointments), automate communication with patients, and provide scheduling and/or staff notification	
	 Generate tailored messages to patients related to needed between-visit care needs 	
Individualized decision support	 Identify relevant information in structured and unstructured EHR data to prioritize differential diagnoses for new symptoms 	
	• Recommend medication options for chronic conditions, considering prior medication prescriptions, allergies, and intolerances noted in structured and unstructured EHR data	

Abbreviations: AI, artificial intelligence; EHR, electronic health record.

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Patient Communication

Research

JAMA Internal Medicine | Original Investigation

Comparing Physician and Artificial Intelligence Chatbot Responses to Patient Questions Posted to a Public Social Media Forum

Invited Commentary page 596

Related article page 507

Supplemental content

John W. Ayers, PhD, MA; Adam Poliak, PhD; Mark Dredze, PhD; Eric C. Leas, PhD, MPH; Zechariah Zhu, BS; Jessica B. Kelley, MSN; Dennis J. Faix, MD; Aaron M. Goodman, MD; Christopher A. Longhurst, MD, MS; Michael Hogarth, MD; Davey M. Smith, MD, MAS

IMPORTANCE The rapid expansion of virtual health care has caused a surge in patient messages concomitant with more work and burnout among health care professionals. Artificial intelligence (AI) assistants could potentially aid in creating answers to patient questions by drafting responses that could be reviewed by clinicians.

OBJECTIVE To evaluate the ability of an AI chatbot assistant (ChatGPT), released in November 2022, to provide quality and empathetic responses to patient questions.

DESIGN, SETTING, AND PARTICIPANTS In this cross-sectional study, a public and nonidentifiable database of questions from a public social media forum (Reddit's r/AskDocs) was used to randomly draw 195 exchanges from October 2022 where a verified physician responded to a public question. Chatbot responses were generated by entering the original question into a fresh session (without prior questions having been asked in the session) on December 22 and 23, 2022. The original question along with anonymized and randomly ordered physician and chatbot responses were evaluated in triplicate by a team of licensed health care professionals. Evaluators chose "which response was better" and judged both "the quality of information provided" (*very poor, poor, acceptable, good, or very good*) and "the empathy or bedside manner provided" (*not empathetic, slightly empathetic, moderately empathetic*,

Chatbot had longer responses or higher quality and empathy compared to doctors



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COMMENTARY

Ambient Artificial Intelligence Scribes to Alleviate the Burden of Clinical Documentation

Aaron A. Tierney, PhD, Gregg Gayre, MD, Brian Hoberman, MD, MBA, Britt Mattern, MBA, Manuel Ballesca, MD, Patricia Kipnis, PhD, Vincent Liu, MD, MS, Kristine Lee, MD Vol. 5 No. 3 | March 2024 DOI: 10.1056/CAT.23.0404

Clinical documentation in the electronic health record (EHR) has become increasingly burdensome for physicians and is a major driver of clinician burnout and dissatisfaction. Time dedicated to clerical activities and data entry during patient encounters also negatively affects the patient-physician relationship by hampering effective and empathetic communication and care. Ambient artificial intelligence (AI) scribes, which

Significantly larger decrease in time spent in the EHR outside 7 a.m. to 7 p.m. among users starting to use the ambient AI scribe than among those who did not use the tool

PressGaney



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Use cases for Safety: Bread and Butter vs. Shiny Object

1. Automate low priority tasks and decrease administrative burden and cognitive load -> more time to focus on delivering safe care

Diagnosis 2.

- Follow-up of test results and referrals/complaints without follow-up
- Imaging —
- Synthesizing information in the workflow

3. Medication Decision Support/Med Reconciliation

- Smarter decision support using more robust algorithms _
- Identifying discrepancies

4. Predicting harm/proactive

Early deterioration/early detection/risk prediction (finding high risk situations through a range of data sources)

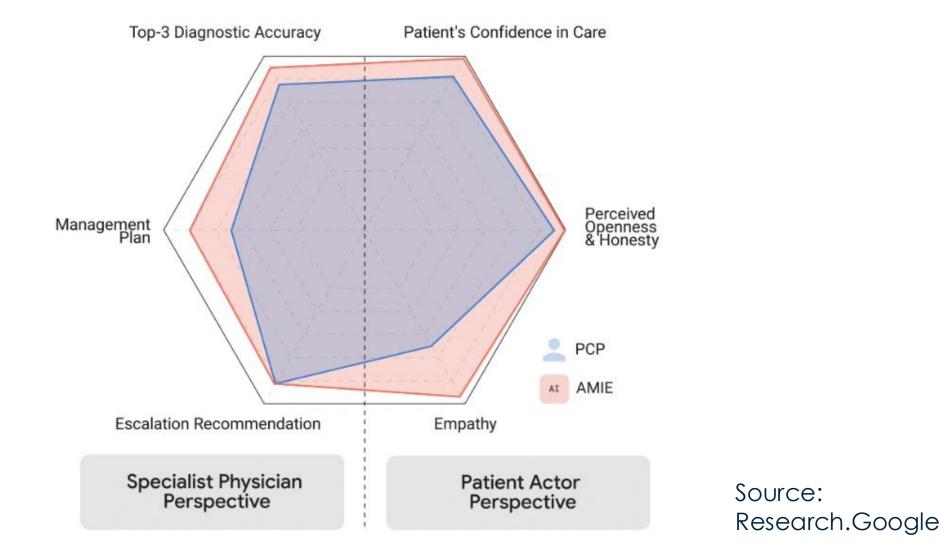
5. Measuring harm

- Triggers _
- Safety event reporting/PSOs
 - Easing burden of reporting by leveraging text to minimize data entry
 - Analyzing to find important trends and helping prioritize

6. Patient engagement

- Identifying safety themes in patient comments Communication tools

Diagnostic Dialogue Evaluation



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What to worry about

Problem	Example
Design Implementation/Use of Findings	 Black box Bias in data/bad data Perpetuating inaccurate information in the EHR (falsehood mimicry) Hallucinations/complete the narrative errors Sycophancy bias Variability (LLMs are probabilistic) Other unintended consequences Patients using AI generated medical advice and experiencing harm Bias in how used Generalizability Drift in how used Decay in algorithm accuracy Inequitable access/widening gaps Depersonalized care/patient concerns
Cognition Distraction De-skilling/automation bias/induced belief revision Re-skilling	 Too much attention to computer monitors and smart-phones, and not enough to patients and colleagues Al does not say patient is deteriorating- do we ignore clinical intuition? What new teaching and competencies do clinicians need?
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Bias/Equity Considerations

Algorithms can perpetuate existing bias in the data

- biased data in= biased data out

Algorithms can potentially reduce existing bias if designed well

- language/access/bias removal

Access/cost of AI can lead to inequities

De-skilling and Re-skilling

- Deskilling/Automation bias/Alert fatigue

- Rely too much on the algorithm/decision support
- Lose clinical judgment and/or skills
- Has been a concern for years
 - E.g. Drug interactions, barcode technology
 - E.g. Cookbook medicine

- Reskilling

- How do we teach about what the algorithms can and cannot do?
- How do we highlight areas where the algorithm may be less certain?
- How do we teach clinicians how to effectively oversee?
 - E.g. review of clinical notes- are we reducing or adding to workload?
- How do we talk to patients about it?

Designing AI to Promote Clinician Vigilance



Visual Cues

Highlight Al output that is more uncertain and potentially faulty.



Clinical-Level Measures

The basis for a system to assess whether a clinician is exhibiting automation bias.



Avoid Throughput Expectations

Al-generated practice efficiencies should not be converted into throughout expectations.



Deliberate Shocks

Program AI to introduce faulty outputs to stimulate vigilance.



Paradigm Shift

Design AI to exercise vigilance over clinicians (e.g., AI only shows if different from clinician choice).

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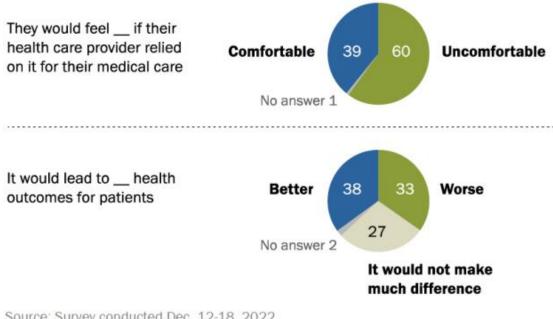
Adler-Milstein J, Redelmeier DA, Wachter RM. The Limits of Clinician Vigilance as an Al Safety Bulwark. *JAMA*. Published online March 14, 2024. doi:10.1001/jama.2024.3620

What Do Patients Think About Al

- Security/privacy
- Consent?
- Depersonalized care vs Increased empathy?
- Increased access to care (chatbots)
- Reduce bias
- Reduce errors
- Transparency
 - Do we have to say when using AI? Do we say when using other tools (e.g. UpTo Date)?
- What are best ways to educate patients?

Fewer than half in U.S. expect artificial intelligence in health and medicine to improve patient outcomes

% of U.S. adults who say that thinking about the use of artificial intelligence in health and medicine to do things like diagnose disease and recommend treatments ...



Source: Survey conducted Dec. 12-18, 2022.

"60% of Americans Would Be Uncomfortable With Provider Relying on Al in Their Own Health Care"

PEW RESEARCH CENTER

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Americans tilt positive on Al's ability to reduce medical errors; greater concern around data security, patient-provider relationships

% of U.S. adults who say the use of artificial intelligence in health and medicine to do things like diagnose diseases and recommend treatments would make each of the following ...

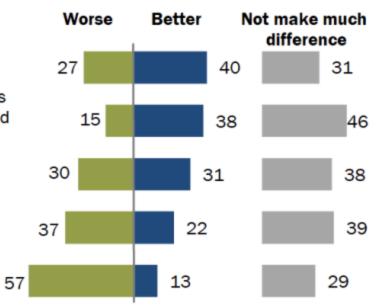
The number of mistakes made by health care providers

The job that health care providers do treating people of all races and ethnicities fairly

The quality of health care for people like you

The security of patients' personal health records

Patients' personal relationship with their health care providers



Note: Respondents who did not give an answer are not shown.

Source: Survey conducted Dec. 12-18, 2022.

"60% of Americans Would Be Uncomfortable With Provider Relying on Al in Their Own Health Care"

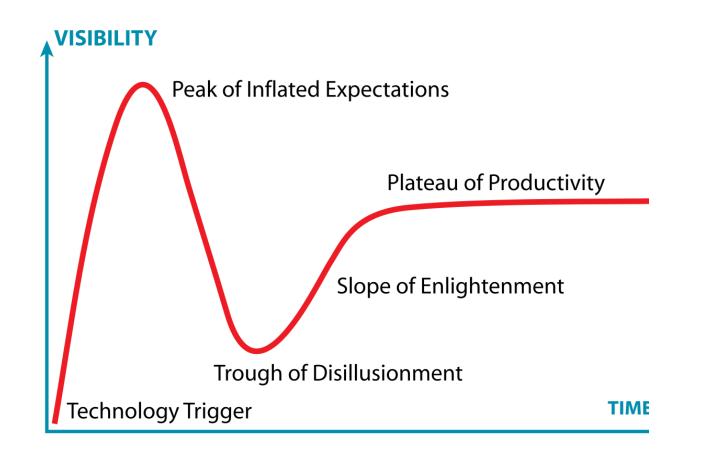
PEW RESEARCH CENTER

How Can Al Improve Patient Safety (per Chat GPT)

- 1. Early Detection and Diagnosis: AI-powered algorithms can analyze large volumes of patient data, including medical records, imaging scans, and genetic information, to detect patterns indicative of potential health risks or diseases at an early stage. This can lead to quicker diagnosis and intervention, reducing the likelihood of medical errors and improving patient outcomes.
- 2. Predictive Analytics: AI can utilize predictive modeling techniques to forecast adverse events, such as hospital-acquired infections or medication errors, by analyzing patient data and identifying risk factors. Healthcare providers can then implement proactive measures to mitigate these risks, thereby enhancing patient safety and preventing avoidable harm.
- 3. Personalized Treatment Plans: AI algorithms can analyze patient characteristics, including genetic factors, medical history, and treatment responses, to tailor personalized treatment plans that optimize efficacy while minimizing adverse effects. By considering individual patient variability, AI-driven decision support systems can help reduce the likelihood of adverse drug reactions or treatment complications, thereby improving patient safety.
- 4. Real-time Monitoring and Alerts: AI-powered monitoring systems can continuously analyze patient data, such as vital signs, medication adherence, and physiological parameters, in real-time. These systems can automatically detect deviations from normal ranges and issue alerts to healthcare providers, enabling timely intervention and preventing adverse events such as sepsis or cardiac arrest.
- 5. Workflow Optimization: AI can streamline healthcare workflows by automating routine tasks, optimizing resource allocation, and prioritizing patient care activities. By reducing administrative burden and minimizing human error, AI technologies free up healthcare professionals to focus on delivering high-quality care, ultimately enhancing patient safety across the care continuum.

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Gartner Technology Hype Cycle



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How to go forward

1. Go forward

- Don't hold AI to impossible standards get wiser about how to use it.
- Transparency about algorithms
- Governance/oversight rapidly evolving

2. Don't be afraid

- It is mainly going to do the easy and boring stuff
- Solve the common problems that have been challenging
- Clinicians have "superpowers" of critical thinking and relationship building

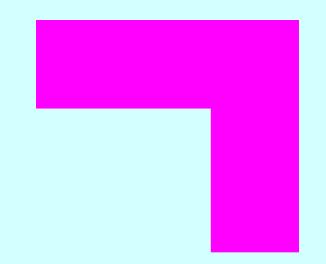
3. Think of AI as a new team member

 Thoughtful, intelligent, able to do some things very well, and others not-so- well

4. Understand/proactively assess and monitor safety implications on many levels

"When will we feel unsafe if we are not using Al?"

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Thank you

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