

# Information is Different Now That You're a Doctor

William Hersh, MD

Professor

Department of Medical Informatics & Clinical Epidemiology

School of Medicine

Email: [hersh@ohsu.edu](mailto:hersh@ohsu.edu)

Web: <http://www.billhersh.info/>

Blog: <https://informaticsprofessor.blogspot.com/>

Twitter: [@williamhersh](https://twitter.com/williamhersh)

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William Hersh, MD  
Professor

Department of Medical Informatics & Clinical Epidemiology  
School of Medicine

Email: [hersh@ohsu.edu](mailto:hersh@ohsu.edu)

Web: <http://www.billhersh.info/>

Blog: <https://informaticsprofessor.blogspot.com/>

Twitter: [@williamhersh](https://twitter.com/williamhersh)

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## About me

- Professor in Department of Medical Informatics & Clinical Epidemiology (DMICE)
- Medical school and residency in internal medicine at University of Illinois Chicago, followed by fellowship in medical informatics at Harvard University
- At OHSU since 1990
- Served as Inaugural Chair of DMICE from 2003-2022
- Have developed informatics educational programs for informaticians, physicians, and others over the years
- Disclosures/Conflict of Interest – None



2



# Session Objectives

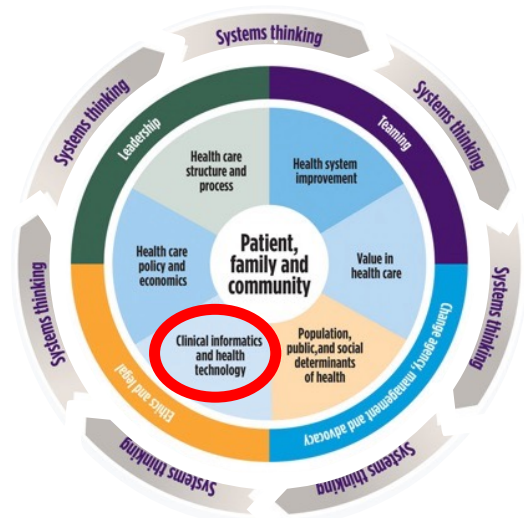
- Define the field of clinical informatics and the central role that data and information play in medicine and healthcare
- Describe how information is different from a medical professional perspective, including how it is used for care and other purposes, kept private and secure, and shared with patients
- Discuss the emerging impact and challenges of artificial intelligence in medicine
- Describe the discipline of clinical informatics as it pertains to healthcare professionals, including those who work professionally in it

## Clinical informatics is part of (but not limited to) Health Systems Science

Hersh and Ehrenfeld, *Clinical Informatics*, Chapter 10 in *Health Systems Science* (Skochelek et al., 2<sup>nd</sup> edition, 2020)

For more information:

- Hersh (Ed.), *Health Informatics: Practical Guide*, 8<sup>th</sup> Edition, Lulu.com, 2022
  - <http://www.informaticsbook.info/>
- What is Biomedical Informatics?
  - <http://informatics.health>



## Information and the new medical student (Shortliffe, 2010)

William Hersh, MD

**W**HEN I FIRST MEET WITH PRECLINICAL MEDICAL students, I make a point of asking them what they believe will receive the greatest focus of their attention once they are in clinical practice. The most common response, not surprisingly, is patients, and yet it is clear to experienced practitioners that the correct answer is information—in the service of their patients. The need for information underlies essentially all clinical work: the questions asked during a patient history, the tests ordered, the books read, and the questions asked of colleagues. A key correlate to information is knowledge, that elusive concept that justifies all the years of education and training, and that provides the background sense of what is true that allows gathering and interpreting information appropriately. Clinicians often start with data (eg, “Mr Jones’ creatinine is 5.2 mg/dL”), those individual elements that combine to allow a synthesis of observations with what is known in order to create summary statements of information (eg, “Mr Jones has renal failure”).

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## Information skills are essential for medical practice (Glasziou, 2008)

The search engine is now as essential as the stethoscope

What we know about diseases, diagnosis, and effective treatments is growing rapidly. Today health professionals cannot solely rely on what they were first taught if they want to do the best for their patients. It has repeatedly been shown that clinical performance deteriorates over time.<sup>1</sup> A commitment to lifelong learning must be integral to ethical professional practice. However, the speed of the increase in knowledge—more than 2000 new research papers are added to Medline each day—represents a challenge.<sup>2</sup> The skills needed to find potentially relevant studies quickly and reliably, to separate the wheat from the chaff, and to apply sound research findings to patient care have today become as essential as skills with a stethoscope.

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## Most of you are “digital natives” but

- Not the same as competence in clinical informatics
- Your relationship with information changes as you become a medical professional
- You become responsible not only for “knowing” information, but also
  - Using it to provide better care of patients
  - Leveraging it to improve the healthcare system
  - Protecting privacy and confidentiality of patients
  - Acting professionally with information
  - Critically analyzing data sets and their potential biases
- Computer literacy is a prerequisite, not an end

## Why is information different now that you’re in medicine?

- Growth of medical knowledge
  - 75 new clinical trials and 11 systematic reviews published each day (Bastian, 2010)
    - To say nothing of the basic science, especially genomics
- Medical knowledge no longer the exclusive purview of physicians
  - >80% of all Internet users search for personal health information (Fox, 2013)

## Many problems in healthcare have information-related solutions

- Quality – not as good as it could be; slightly more than half of patients get care they should get (McGlynn, 2003; McGlynn, 2020)
- Safety – errors cause morbidity and mortality; many preventable (IOM, 2000; Leape, 2021)
- Cost – US spends more and gets less (Gunja, 2023)
- Inaccessible information – missing information frequent in primary care (Smith, 2005)

## EHR is more than “charting”

- Physicians must be able to
  - Move from one vendor system to another
  - Effectively use *clinical decision support* to remind us of things to do and warn us about things not to do (Greenes, 2023)
  - Access information from other settings where patient received care through *health information exchange* (Dixon, 2022)
  - Apply *data analytics*, especially in setting of population health management, to achieve quality, safety, and cost-effectiveness
  - Integrate *artificial intelligence* (AI) in care of patients

## Patients want more access to data and information too

- They have access to just about all of the same knowledge resources we can access through the *personal health record* (PHR)
  - And increasingly all of their medical record
- They want to interact with us digitally and want to interact with healthcare the way they interact with airlines, retailers, banks, etc.
- They want access to and control over their data
  - We must educate them in the risks and benefits



## Those who pay for care want more accountability from us

- Purchasers (employers, government) and payors (insurers) want assurance that care provided is high-quality and cost-effective
  - *Clinical decision support* aims to help physicians make best choices and avoid errors
  - Use of *quality measurement and improvement*
- Leading to calls for a *learning health system*, where we learn from data to improve care (IOM, 2012)

## We also have responsibilities around data and information

- Patients expect us to keep their information private and secure
  - *Health Insurance Portability and Accountability Act* (HIPAA) regulations guide our actions
    - Treatment, payment, and operations (TPO) allow disclosure
    - Other uses require patient consent
- Our public-facing persona must be professional, especially on social media
- Growing recognition of bias in data and algorithms
  - Algorithms mis-appropriating resources (Obermeyer, 2019; Kakani, 2020)
  - Companies and others “monetizing” our personal health data (McGraw, 2020)
  - AI may compromise care (DeCamp, 2023)
  - Implementing responsible (Dorr, 2023) and fair (Chen, 2023) AI

## We must also manage and lead the introduction of AI in medicine

- AI – “information systems and algorithms capable of performing tasks associated with human intelligence” (Rajpurkar, 2022; Sahni, 2023)
  - Modern success from advances machine learning (ML) – “computer programs that learn without being explicitly programmed” (Alpaydin, 2020)
- AI is not new (Shortliffe, 2019)
  - First usage began in 1960s, aiming to build computer programs based on human-constructed knowledge bases
  - Some successes but difficult to scale, leading to “AI winter” of 1990s
- Re-emergence of AI in 2000s with advances in ML, much larger data availability, and more powerful and Internet-based computers
  - Major advances in ML from *deep learning* (Esteva, 2019; Topol, 2019)
  - Initial successes in *predictive AI* but now success (and hype) of *generative AI*, e.g., ChatGPT (Lee, 2023)
  - Has led to new field of *data science* (Donoho, 2017; in biomedicine: Hoyt, 2019)



# Impressive results of predictive AI on various types of data

- Most success has been with image interpretation (Rajpurkar, 2023); examples include
  - Radiology – chest x-rays for diagnosis of pneumonia and tuberculosis
  - Ophthalmology – retinal images for diagnosis of diabetic retinopathy
  - Dermatology – skin lesions for diagnosis of cancer
  - Pathology – breast cancer slides to predict metastasis
- Achievements in other areas
  - Predicting adverse events in hospitalizations (Rajkomar, 2018)
  - Generating clinical notes from patient and physician verbal interaction (Rajkomar, 2019)
  - Predicting protein folding from amino acid sequences (Jumper, 2021)

# Including ability to “see” where humans cannot

- Retinal images
  - Age, sex, and cardiovascular risk determination from retinal images (Poplin, 2018)
  - Race (Coyner, 2023)
- ECG
  - Age and sex determination (Attia, 2019)
  - Hyperkalemia from 2 (of 12) leads (Galloway, 2019)
  - Diagnosis and risk stratification of occlusive myocardial infarction (Al-Zaiti, 2023)
  - Chronic kidney disease (Holmstrom, 2023)
  - Left ventricular systolic dysfunction from ECG images (Sangha, 2023)
- Chest x-ray
  - Race (Gichoya, 2022)
  - Cardiac function and valvular heart diseases (Ueda, 2023)
  - Diabetes (Pyrros, 2023)
  - Correlation with chronological age in healthy cohorts and, for various chronic diseases, difference between estimated age and chronological age (Mitsuyama, 2023)

## And now, generative AI and large language models (LLM)

- Introduction of ChatGPT in November, 2022 brought a new type of AI into focus: generative AI
- LLMs based on transformer models trained with large amounts of text and achieving unprecedented ability to chat and answer questions (Thirunavukarasu, 2023)
- Although imperfect and sometimes prone to confabulation (Lee, 2023), impressive performance in medicine and beyond, e.g.,
  - US Medical Licensing Exam (USMLE) (Nori, 2023)
  - Board exams in e.g., radiology (Bhayana, 2023) and clinical informatics (Kumah-Crystal, 2023)
  - New England Journal of Medicine clinical cases (Kanjee, 2023)
  - Answering questions in social media forums (Ayers, 2023)
  - Drafting letters to patients (Ali, 2023)

## Will AI replace physicians?

- Real-world use and evidence base still modest
  - Systematic review of clinical trials of predictive AI systems showed small number of trials (relative to predictive modeling papers), mediocre methodologies, and mixed results (Plana, 2022)
- “AI won’t replace radiologists, but radiologists who use AI will replace radiologists who don’t,” (Langlotz, 2019)
- Must address bias in data and algorithms
  - AI may compromise care if not used properly (DeCamp, 2023)
  - Implementing responsible (Dorr, 2023) and fair (Chen, 2023) AI



## We must also learn to practice medicine by alternative modalities

- Telehealth/telemedicine – clinical care separated by time and/or distance (Daniel, 2015)
  - Synchronous – real-time
  - Asynchronous – sending images, video, etc.
- Usage exploded at onset of pandemic, aided by relaxation of rules (Verma, 2020)
  - Has reduced from peak but well above pre-pandemic baseline (Anderson, 2022)

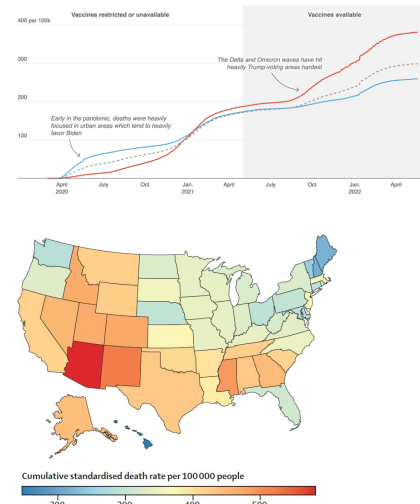
## Informatics and medical education

- “Informatics training for clinicians is more important than hardware and software” (Safran, 2009)
- Health informatics is a “required skill for 21<sup>st</sup> century clinicians” (Fridsma, 2018)
- OHSU is a leader: competencies (Hersh, 2014; Hersh, 2020) and curricula (Hersh, 2017)
- Need to prepare physicians for the “clinical algorithm era” (Goodman, 2023)

1. Find, search, and apply knowledge-based information to patient care and other clinical tasks
2. Effectively read from, and write to, the electronic health record for patient care and other clinical activities
3. Use and guide implementation of clinical decision support (CDS)
4. Provide care using population health management approaches
5. Protect patient privacy and security
6. Use information technology to improve patient safety
7. Engage in quality measurement selection and improvement
8. Use health information exchange (HIE) to identify and access patient information across clinical settings
9. Engage patients to improve their health and care delivery through personal health records and patient portals
10. Maintain professionalism through use of information technology tools
11. Provide clinical care via telemedicine and refer patients as indicated
12. Apply personalized/precision medicine
13. Participate in practice-based clinical and translational research
14. Apply machine learning applications in clinical care

# Informatics lessons from COVID-19

- Exposed both health disparities and technology disparities
  - Broadband Internet is a social determinant of health (Sieck, 2021)
  - Public health information systems not up to task (Gottlieb, 2021)
- Disinformation can be deadly
  - Spread widely by small number of people (Disinformation Dozen, 2021), including about 50 physicians (Sule, 2023), augmented via large following and automated means on social media, e.g., Facebook (Ayers, 2021)
  - Impact related to vaccine uptake and (indirectly) political leanings (Wood, 2022; Bollyky, 2023)
  - Leading to assaults on science and scientists (Hotez, 2021; Hotez, 2023)



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## Clinical informatics

- Part of larger *biomedical and health informatics*, the field concerned with the optimal use of information, often aided by technology, to improve
  - Individual health
  - Healthcare
  - Public health
  - Biomedical research
- (Detmer, 2014; Hersh, 2020; Hersh 2022)



William Hersh, MD

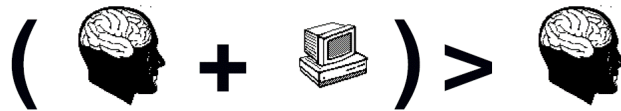
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# Fundamental theorem of informatics

Goal of informatics is



Goal is not



(Friedman, 2009)

## Clinical informatics

- Competence required of all; career opportunities available for some
- Growing number of physicians work in roles such as *Chief Medical Informatics Officer* (CMIO) or others in academia or industry
  - Clinical informatics is now a subspecialty of all medical specialties
- OHSU is a national leader in research and education; MD informatics curriculum probably more advanced than any other medical school
  - <http://www.ohsu.edu/informatics>

# What can you do in clinical informatics?

- Informatics skills are essential to the practice of the 21<sup>st</sup> century physician
  - You should master informatics just as you master any other clinical skill
- For those interested as a career, plenty of opportunities in medical school and beyond
  - Scholarly projects, electives, and more
  - Advanced study – e.g., graduate degree and/or fellowship
  - Clinical informatics subspecialty – of all specialties; requires fellowship after primary specialty training

## Discussion

- What are the most important ways that clinical informatics can benefit healthcare, public health, and individual health?
- How can and should we engage patients in the use of informatics tools we have discussed today?
- How can we make the best use of AI while minimizing its risks?



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