

Essentials of Artificial Intelligence for Nursing

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Learning Outcome

- After completing this learning activity, the participant will be able to examine technology and the role of nurses in incorporating artificial intelligence into the healthcare setting.

Objectives

Describe how Artificial Intelligence is impacting healthcare

- Introduce the use of artificial intelligence (AI) in healthcare and Nursing
- Briefly describe AI
- Introduce the DIKW framework. Introduce professional responsibility to understand the key concepts and relevant questions to ask
- Introduce the DIKW framework and the quintuple aim as a framework for value-based care

Describe 2 basic Artificial Intelligence principles

- Review the DIKW (data-information- knowledge-wisdom) framework
- Discuss how data can be used by giving examples of real-world applications within the DIKW framework

Describe 2 uses of Artificial Intelligence technology and their impact on nursing practice and patient care

- Description of two AI decision support tools that can help nurses optimize their practice
- Discuss important questions regarding AI/CDS tools that can help nurse leaders identify the optimal technology for improving patient care and nursing practice in their unique environment



“That it will ever come into general use, notwithstanding its value, I am extremely doubtful; because its beneficial application requires **much time**, and gives a good deal of **trouble** both to the patient and the practitioner; and because its whole hue and character is **foreign**, and opposed to all our habits and associations.”

John Forbes MD, 1821

"Artificial intelligence represents one of technology's most important priorities, and healthcare is perhaps AI's most urgent application"

Satya Nadella
CEO, Microsoft



Digital Transformation in Health

Technology enabled care, health promotion and disease prevention that advances the Quintuple Aim



Better Health

Improve population health



Better Care

Improve the experience of care



Lower Cost

Reduce the per capita cost of care



Clinician Experience

Improve clinician workflow efficiencies



Health Equity

Understanding Social Determinants of Health

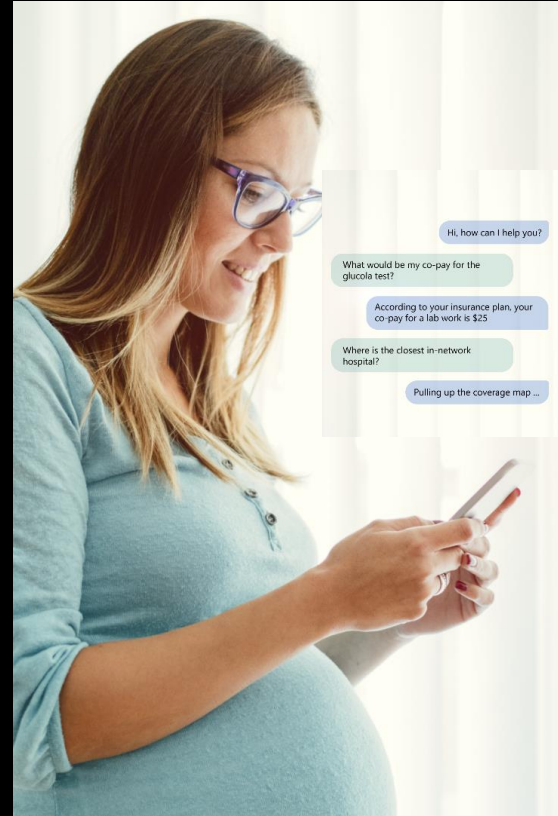
Successful digital transformation is focused on four areas to unlock value



Vision
& strategy



Culture



Unique
potential



Capabilities

Why now?



Why now?

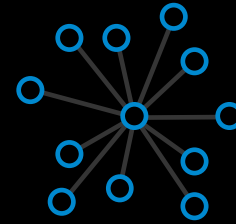


101010
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More data



More Compute

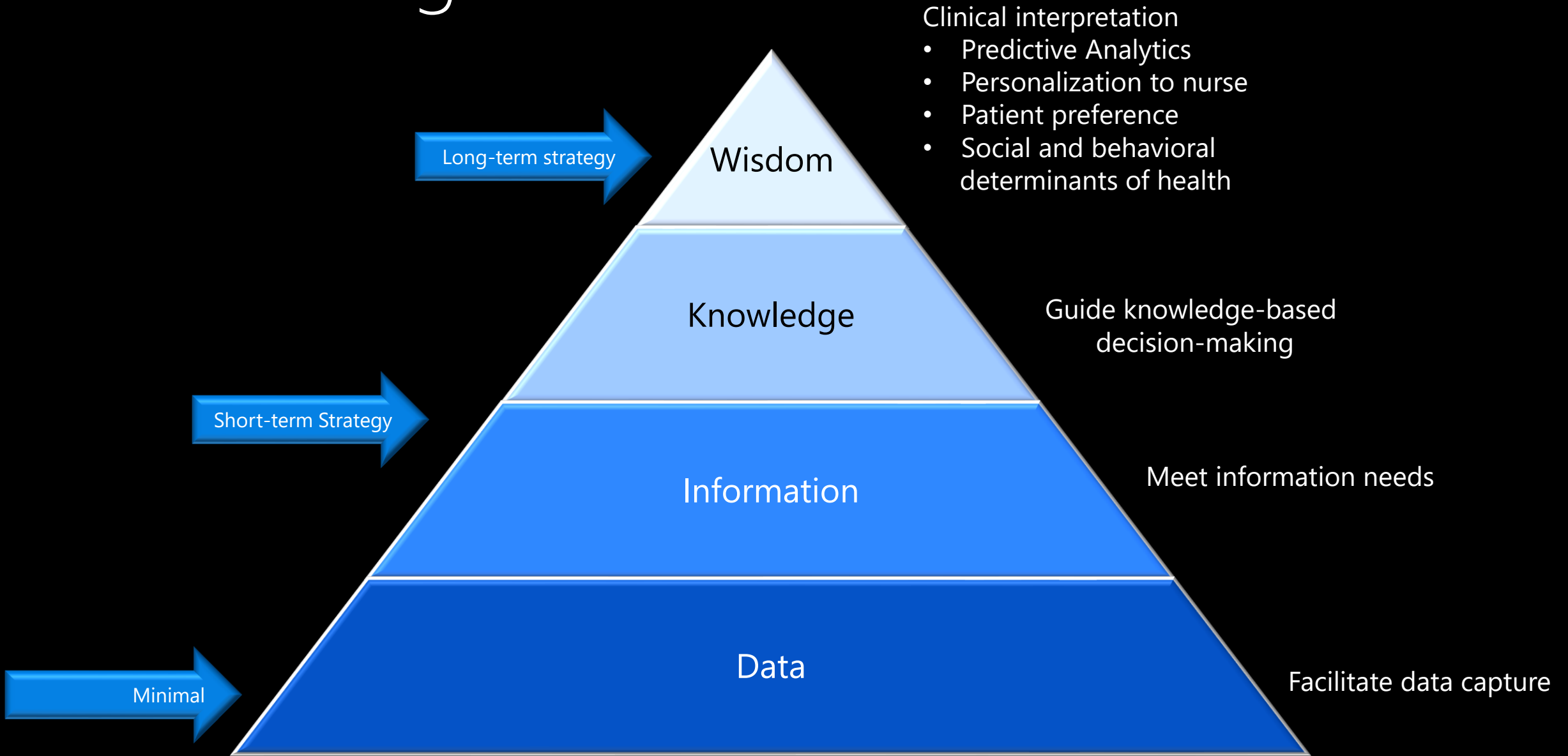


Innovative
algorithms,
tools &
frameworks

Data Information Knowledge Wisdom

Concept	Definition
Wisdom	Understanding and internalization of knowledge patterns and relationships
Knowledge	Derived by discovering patterns and relationships between types of information
Information	Data plus meaning
Data	Little or no meaning in isolation

DIKW Strategic Plan





What is Artificial Intelligence?

"AI is an area of computer science that emphasizes the creation of machines that work and react like humans. This means system that have the ability to depict or mimic human brain functions including learning, speech (recognition and generation); problem-solving, vision and knowledge generation."

Amplifying human ingenuity with intelligent technology



Learning

Learn over time without
direct intervention



Reasoning

Interpret meaning of data
including text, voice, images



Cognition

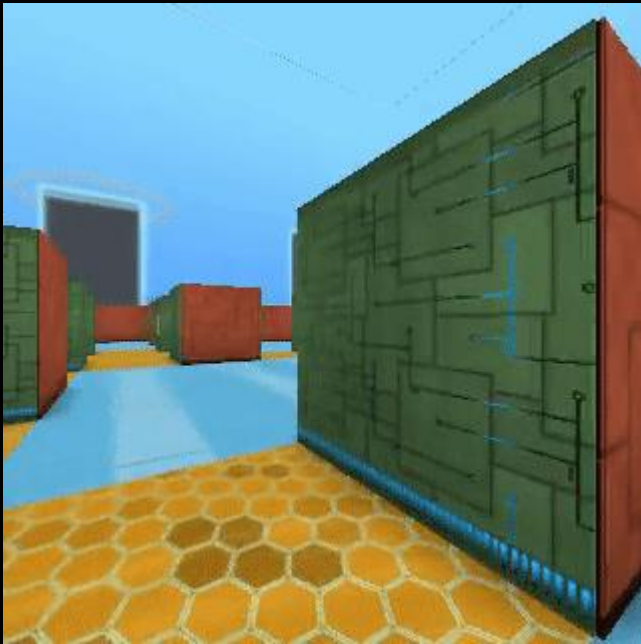
Form conclusions with
imperfect data



Interacting

Interact with people
in natural ways

Artificial Intelligence (AI)



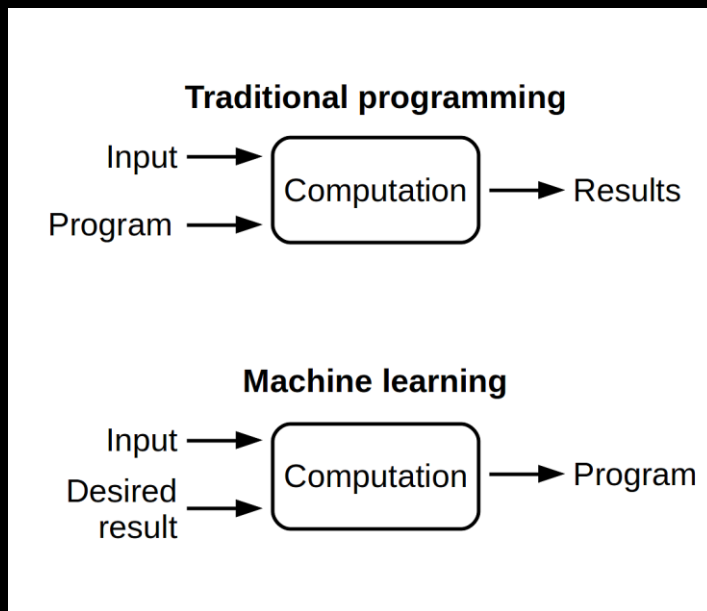
source: <https://www.youtube.com/watch?v=M40rN7afngY>

The study of “intelligent agents”. Any device that perceives its environment and takes actions that maximize its chance of successfully achieving its goals.



source: <https://www.youtube.com/watch?v=imOt8ST4Ejc>

Machine learning (ML)



The study of algorithms to perform a specific task without using explicit instructions. Instead build a model based on sample data, known as "training data", in order to make predictions without being explicitly programmed to perform the task.

Typically seen as a subset of AI.

source: <https://www.futurice.com/blog/differences-between-machine-learning-and-software-engineering/>

Building models

What is a model?



A model is a function, with its parameters learned from data

How is it created?



Machine learning is using a variety of algorithms and techniques to learn the right parameters

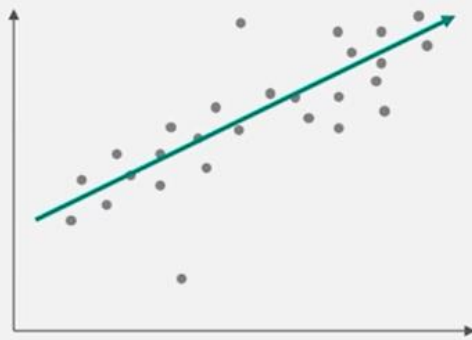
What languages are used?



Majority of ML is done in Python and R, using frameworks like scikit-learn

What does a model do?

Regression



$$F(x) = mx + b$$

Learn m & b from the data

Classification



Dog
.78

Cat
.04

Tiger
.001

Clustering

- Segmenting customers
- Arranging articles into categories
- Discovering similar items

What is Deep Learning?

- Deep Learning describes techniques to build models which use neural networks
- Deep networks (many layers) enable you to learn very complex functions
- Generally, uses frameworks like Tensorflow, PyTorch, Chainer & more
- The math used in training a deep learning model can be accelerated on specialized hardware like GPUs and FPGAs

AI for Health:

When we refer to AI in the context of health we typically mean ML

The AI/ML components are individual pieces in bigger health systems, that combine big data, cloud computing, and emerging devices and user interfaces



Big
Data

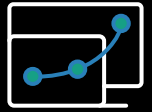


Big Computing
in the Cloud

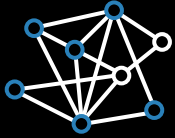


Powerful
Algorithms,
APIs, Bots & More

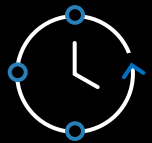
World view



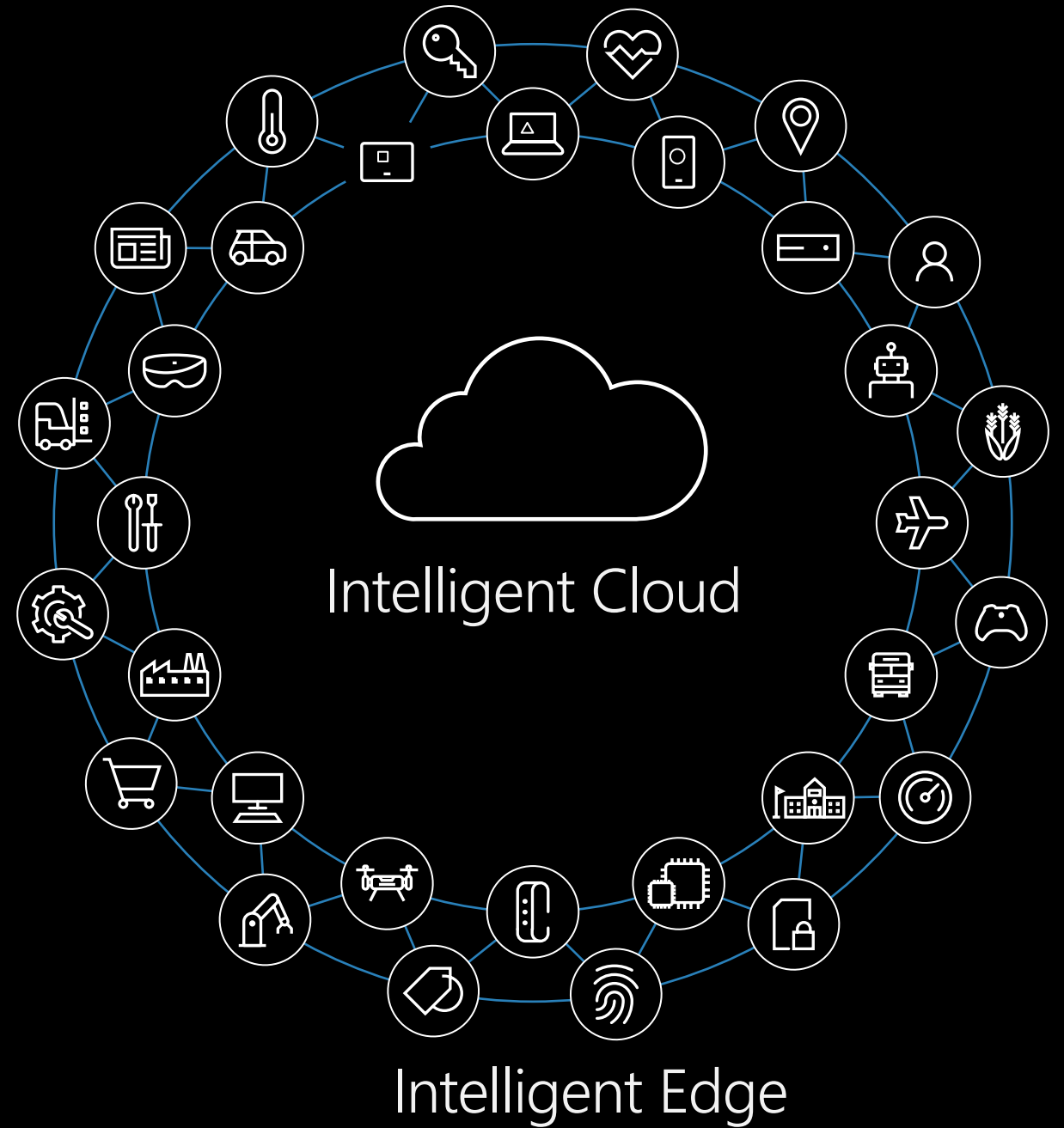
Multi-device,
Multi-sense



Artificial
Intelligence



Serverless



The Bigger Picture: Health Use Case Taxonomy

Clinical Analytics:

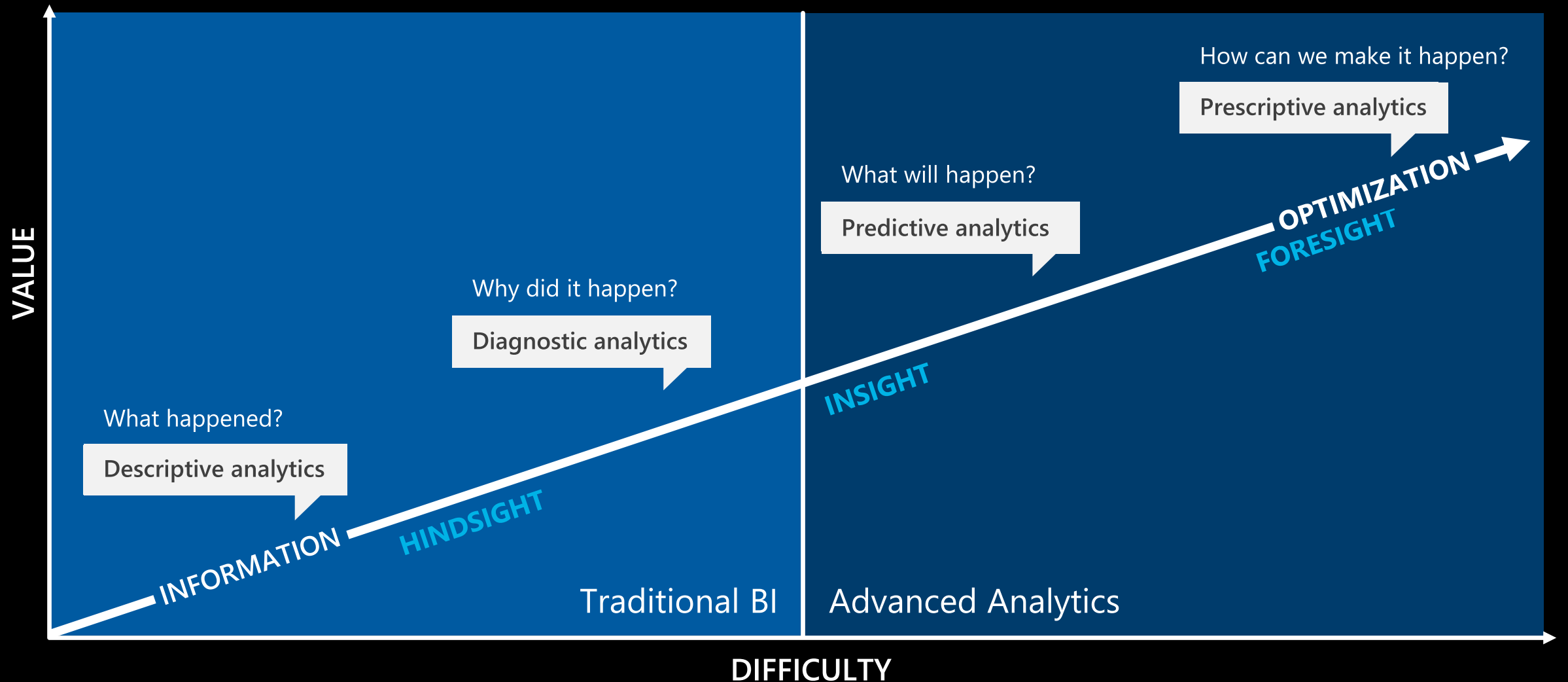
Transform data into prescriptive insights

Operational Analytics

Actionable insights to optimize performance



Beyond business intelligence



Advanced Analytics in Health: New Paradigms



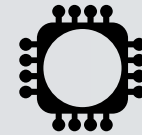
REAL-TIME & PREDICTIVE ANALYTICS

to shape outcomes



SELF-SERVICE ANALYTICS

to provide research on-demand



INTERNET OF THINGS

to connect devices to data bases



CLOUD ANALYTICS

to erase historical boundaries for data

Geo/Social/Environmental Data
Weather patterns, economics, social services data etc.

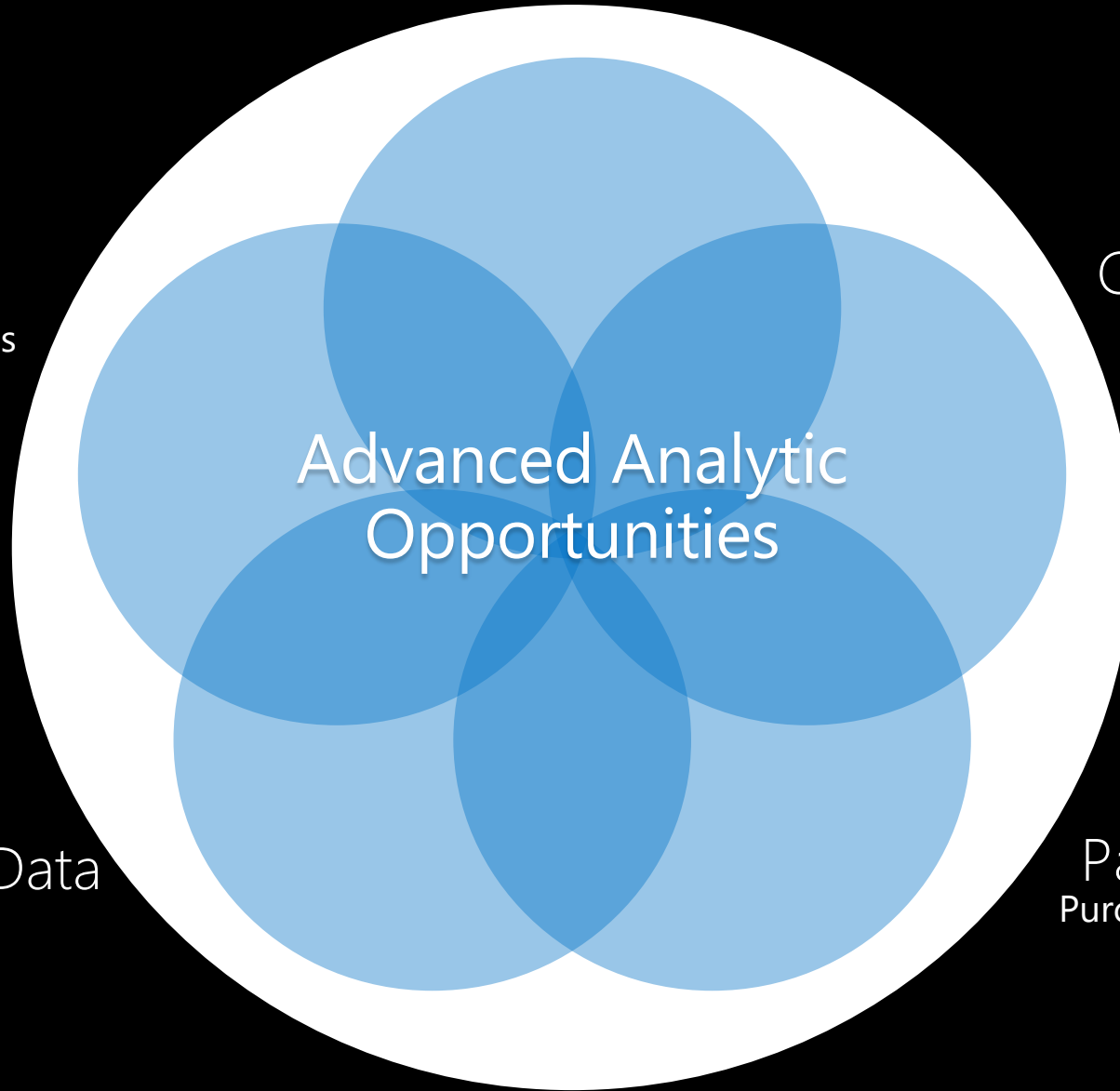
Clinical Data
EMRs, diagnostic images

Claims & Cost Data
Claims, revenue cycle

Advanced Analytic
Opportunities

Pharma & Life Science Data
Clinical trials, genomics

Patient & Citizen Data
Purchasing patterns, open gov data,
social media,





Vision

Image tagging, thumbnails
 OCR, handwriting recognition
 Customized image recognition
 Face detection
 Emotion recognition
 Video insights
 Image and video moderation
 Log detection



Speech

Speech to text (speech transcription)
 Customer speech transcription (complex word, noisy environment)
 Text to speech
 Speaker ID and authentication
 Real-time speech translation



Language

Contextual language understanding – customize intent analysis
 Sentiment analysis, key phrase detection
 Text translation up to 60+ languages
 Spell checking



Knowledge

Academic Knowledge
 Entity Linking
 Knowledge Exploration
 Recommendations
 QnA Maker



Search

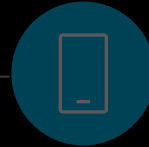
Automatic search suggestions
 Comprehensive news, image, video results
 Entity Information augmentation
 Tailored and customized search experiences
 Local business search

Microsoft Cognitive Services

Systems of Engagement



CRM



Mobile



Chat

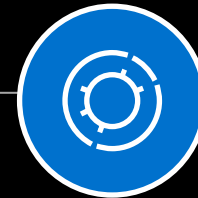
Systems of Intelligence



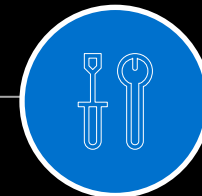
Transform clinical services



Engage patients & consumers



Optimize operations



Empower clinicians & employees

Systems of Record



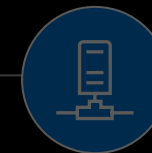
EMR



PAS



ERP



Imaging

AI Maturity Model

Approaching



Hopeful on AI and its promise

- Digitization under way
- Looking to increase or optimize processes
- Cautious about disruption

Aspirational



Experimented and applied AI

- High digitalization
- Desires new business models
- Achieved a data culture

Mature



Emerging data science and operational capability

- Understands model lifecycle and management
- Building a foundational data architecture

Foundational



Questioning what AI is and how to apply it

- Wrong expectations or disappointment
- Low digitization
- Basic analytical

 Culture Shift

 Ethical Shift

 Ownership Shift

AI challenges and risks

Challenges

- Interdisciplinary effort between ML and medical professionals
- Data that is siloed, unstructured and manual entry
- Data privacy considerations
- Lack of interoperability

Risks

- Exacerbating societal biases
- Loss of jobs requiring training for new skills
- Lack of regulatory oversight

Six principles to guide AI development and use



Fairness



Reliability
& Safety



Security &
Privacy



Inclusiveness



Transparency



Accountability

Concerns of ML

ML models reflect (and often amplify) biases in training data

81% of participants in genome-mapping studies were of European descent.

RESEARCH ARTICLE

Dissecting racial bias in an algorithm used to manage the health of populations

Ziad Obermeyer^{1,2,*}, Brian Powers³, Christine Vogeli⁴, Sendhil Mullainathan^{5,*†}

+ See all authors and affiliations

Science 25 Oct 2019:
Vol. 366, Issue 6464, pp. 447-453
DOI: 10.1126/science.aax2342

A person's body-hair type can skew an AI's assessment of whether or not he or she has skin cancer.

<https://science.sciencemag.org/content/366/6464/447/tab-pdf>

<https://qz.com/1367177/if-ai-is-going-to-be-the-worlds-doctor-it-needs-better-textbooks/>

Concerns of ML

ML models reflect (and often amplify) biases in training data.

Any discussion about bias in AI will be confusing, difficult, and uncomfortable, because bias is hidden and tricky, until it's obvious and dangerous. That is to say, biased outcomes from a biased algorithm are easier to spot than the biased data fed into the machine.

87% of participants in economic modeling

body-hair type can skew an AI's assessment of whether or not he or she has skin cancer.

Bias example

Case study: predicting pneumonia risk

- Predict probability of death for patients with pneumonia so that high-risk patients can be admitted to the hospital while low-risk patients are treated as outpatients
- ML models perform well on this task, BUT closer inspection reveal that on one dataset the models learn $\text{HasAsthma}(x) \Rightarrow \text{LowerRisk}(x)$
- Why does the model learn this (obviously) wrong relationship?

Why $\text{HasAsthma}(x) \Rightarrow \text{LowerRisk}(x)$?

BECAUSE it reflects a true pattern in the training data:

- Patients with history of asthma who presented with pneumonia were admitted directly to ICU (Intensive Care Unit)
- The aggressive care received in ICU lowered their risk of dying from pneumonia compared to the general population

Models trained on the data incorrectly learned that asthma lowers risk, when in fact asthmatics have much higher risk (if not hospitalized).

So what can we do?

- More research on AI and ML fairness on existing models
- Understand and identify problems prior to deploying
- Involve diverse stakeholders & multiple perspectives
- Research on new models appropriate for health applications
 - Interpretable machine learning models
 - Causal models

Questions for Clinical Leaders to ask

Concept	Question
Data	<ul style="list-style-type: none">• What data are used in the AI/CDS tool?• How are data captured?• Does the data capture fit into the existing clinical workflow?• Is there an appropriate life cycle plan for the CDS?
Information	<ul style="list-style-type: none">• Does the AI/CDS information take into account the clinical context?• The CDS should be agile enough to adapt to changes in clinical settings.• Does the information produced make clinical sense and have clinical relevance?
Knowledge	<ul style="list-style-type: none">• Does the AI/CDS help solve a clinical problem? What were the examples used to teach the model?• Does the CDS fit nursing processes?
Wisdom	<ul style="list-style-type: none">• Is the AI/CDS augmenting or taking over decision-making?• Is the AI explainable to the clinician?• Is the required short- and long-term training in place?

Ethical questions?



- What is the plan for thoroughly stress testing AI for unintended biases?
- Are we effectively identifying and managing the ethical implications of technology?
- Can we explain how the AI makes decisions using data?
- Is there a clear breakdown of how the adoption of AI will adhere to the fundamental principles of managing AI in an ethical way?

Responsible advancement of AI

- Demand non-discrimination
- Protect the patient
- Investigate the AI algorithm
- Check the data

AI is about:

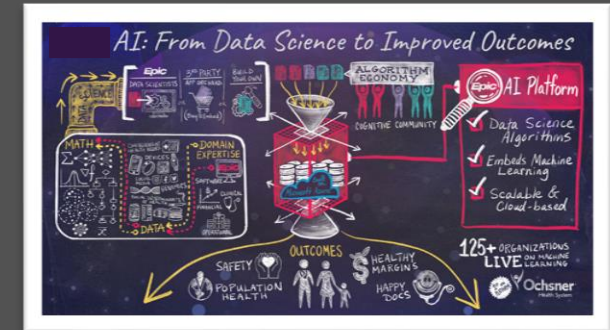
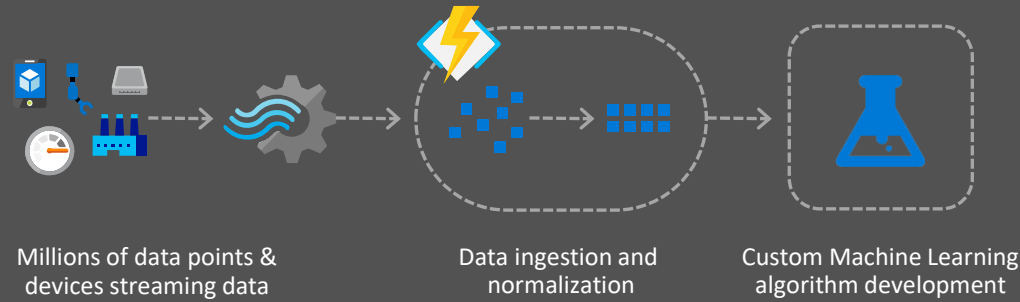
- Tasks, not jobs
- Helping employees and clinicians do their jobs, not replacing their jobs
- Eliminating burdensome tasks and producing better outcomes
- A better work life for employees and clinicians
- Creating better experiences for patients

Critical success factors for AI Integration

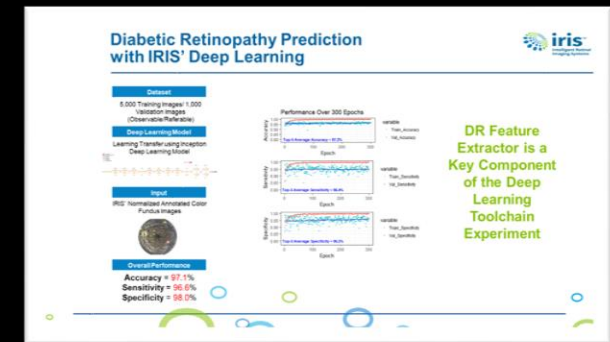
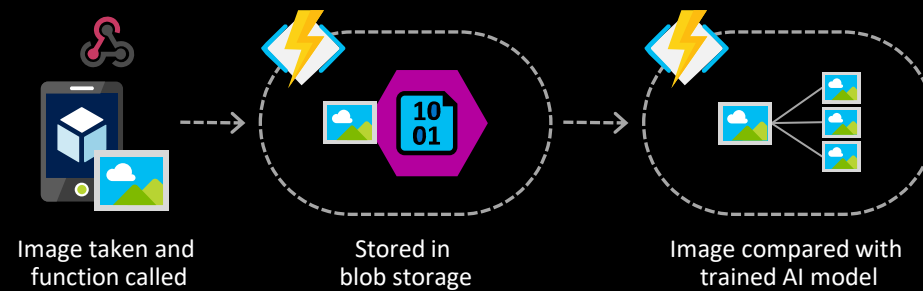
- Multidisciplinary team management
- Change management approach
- Culture
- Partnering with AI experts
- Workflow
- Project identification

Emerging cloud & AI application patterns

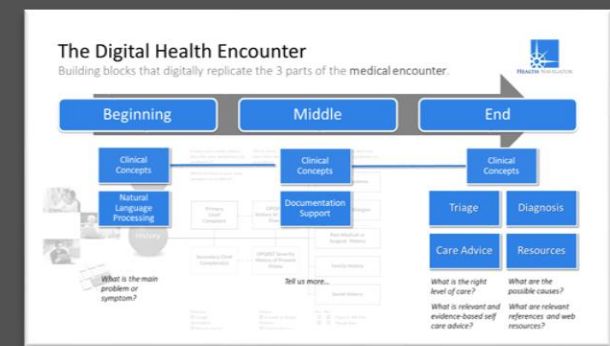
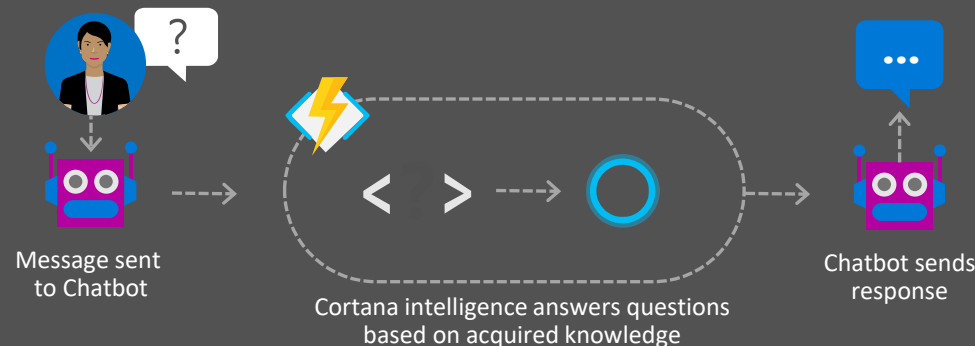
Continuous patient monitoring



AI infused clinical workflows



Interactive health agents



Reinventing our health system for the digital age



Accelerating time to diagnosis



Prevent the sick getting sicker



Keeping the sick out of the hospital



Detecting disease before it reaches us

As “smart” as AI is at certain things, no one has figured out how to imbue or mimic those uniquely human qualities that are essential to the care process.

Wisdom

Reasoning

Judgement

Imagination

Critical thinking

Common sense

Empathy

Transforming healthcare

- AI/ML is disrupting the health industry
- More research is necessary to understand differential implications to different demographics
- Policies and regulations need to catch up with these new models and their applications



Artificial Intelligence is not about quantifying the obvious...

It's about opening our minds and processes to discover new things we have yet to consider.

Learn more:

<https://sponsors.aha.org/HFC-Gen-Microsoft-AI-in-Healthcare2021.html>



AI in Health Care: Leading Through Change

Join us to learn how AI-powered solutions are transforming health care and earn continuing education credit.

[Register for this Complimentary Course ›](#)



Health care is evolving faster than we ever could have imagined. We are in the midst of a health care revolution, driven by Artificial Intelligence (AI). AI is creating a space for innovation and can enable you to work more efficiently and effectively in the face of unprecedented change.

To help you understand how AI is meeting the today's challenges and is shaping your work, the American Hospital Association and Microsoft have teamed up to offer a unique course.

- **Convenient:** Learn on-demand, at your convenience.
- **Fast-paced:** Quickly complete the course in just one hour.
- **Relevant:** Understand the major role AI will play in the future of care delivery. Participants may earn continuing education contact hours.
- **Transformative:** Challenge the way you think about health care from renown instructors.

Thank you

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→ <https://www.linkedin.com/in/kathleenmcgrow/>



References

- Frost and Sullivan. Artificial intelligence in healthcare takes precision medicine to the next level. Cision: PR Newswire. 2018. <http://www.prnewswire.com/news-releases/artificial-intelligence-in-healthcare-takes-precision-medicine-to-the-next-level-300712098.html>.
- Pan Y. Heading toward Artificial Intelligence 2.0. *Engineering*. 2016;2(4):409–413.
- Obermeyer Z, Emanuel EJ. Predicting the future—big data, machine learning, and clinical medicine. *N Engl J Med*. 2016;375(13):1216–1219.
- Starr D. Current use cases for machine learning in healthcare. Microsoft Azure. 2018. <https://azure.microsoft.com/en-us/blog/current-use-cases-for-machine-learning-in-healthcare>.
- Panetta K. 5 trends emerge in the Gartner Hype Cycle for emerging technologies, 2018. Gartner. 2018. <http://www.gartner.com/smarterwithgartner/5-trends-emerge-in-gartner-hype-cycle-for-emerging-technologies-2018>, [Cited Here](#)
- Fu K, Swain PH. On syntactic pattern recognition. *SEN Report Series Software Engineering*. 1971;(2):155–182.
- Shetty B. Natural language processing (NLP) for machine learning. Towards Data Science. 2018. <https://towardsdatascience.com/natural-language-processing-nlp-for-machine-learning-d44498845d5b>.
- College of Southern Nevada: College Library Services. Databases vs. search engines: what's the difference? <http://www.csn.edu/sites/default/files/legacy/PDFFiles/Library/dbasesearch3.pdf>.
- Brownlee J. A gentle introduction to computer vision. Machine Learning Mastery. 2019. <https://machinelearningmastery.com/what-is-computer-vision>.
- Lustig T. Fueling science and research through machine perception. Mellanox Technologies. 2017. <http://www.mellanox.com/blog/2017/09/science-research-machine-perception-ai>.
- Chinnakali K. 9 key benefits of data lake. Data Science Central. 2016. <http://www.datasciencecentral.com/profiles/blogs/9-key-benefits-of-data-lake>.
- Kumar R. SQL server blog: Microsoft for the modern data estate. Microsoft. 2017. <https://cloudblogs.microsoft.com/sqlserver/2017/09/25/microsoft-for-the-modern-data-estate>.

