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AI IN HEALTH CARE- MIT SLOAN SCHOOL OF MANAGEMENT

DIGITAL TRANSFORMATION AND INNOVATION IN HEALTHCARE-IMPERIAL COLLEGE LONDON

AI FOR HEALTHCARE: EQUIPPING THE WORKFORCE FOR DIGITAL TRANSFORMATION- UNIVERSITY OF MANCHESTER AND NHSE

CLINICAL SAFETY OFFICER

Al definition



Deep Learning



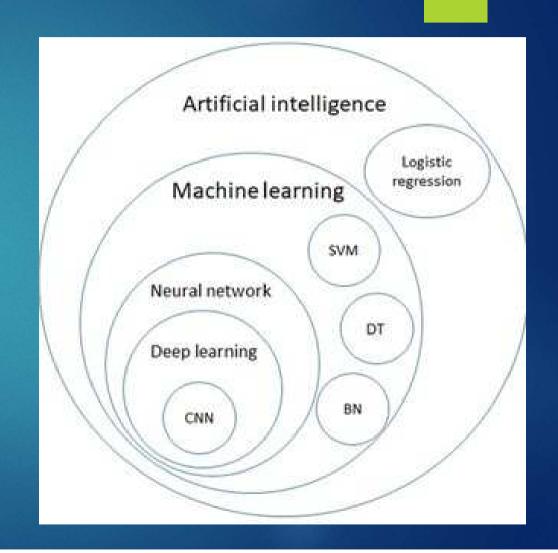
Machine Learning



Artificial Intelligence



Predictive Al/Generative Al



Main issues with AI in health

- Political Is is politically acceptable?
- ▶ Economic Is it commercially viable- short, medium or long term?
- Social Cultural beliefs, health and safety, bias, equity, patient and clinician acceptance
- ► Technological Is it technically feasible now?
- ▶ Legal Medico-legal regulations are way behind innovation
- **Environmental** Effect on environment

Why AI?

- Al predictions could reduce pressure on A&E by triaging patients via online consultations more consistently
- It can be useful in radiology to rule in or out abnormalities, to help radiologists concentrate on potentially abnormal scans
- ▶ It can be useful in predicting progression of disease
- It can help advise on real world outcomes in multi-morbidity where there is lack of clinical trial data
- It can personalize surgical and cancer treatment
- It can help optimize patient flow in hospitals and primary care
- Can help extremely customized treatment for individual patients

Limitations of AI in health care?

- ► Explainability and Interpretability: Medicolegally, AI models needs to be explainable and results interpretable
- ▶ **Differences in machines:** Algorithms trained with data from one type of imaging machine may not perform in the same way when assessing instances from other types
- ▶ Reporting accuracy: The nature of your training data's composition means that the ostensible accuracy of your model might not be as accurate as purported.
- ▶ Biased data: Data used to train, or test may not be representative of the actual patient
- ▶ **Distributional shift:** Model may be unable to accurately apply what it has learned from training data to a novel set of data if there is a significant difference
- Adversarial interventions: Tricked to produce a certain result
- Democratization: It is not always possible to take a preconstructed algorithm, apply it in a new situation, and expect perfect performance

Potential challenges to implementing AI solutions:

- ▶ Overenthusiasm: Enthusiasm should be approached with caution when you aim to implement AI in health care. Overemphasized and glorified claims, such as computers replacing all physicians, can create unrealistic predictions and a false sense of what AI will be able to do.
- ► Fear: The inverse of overenthusiasm, one should not let fearfulness unnecessarily block AI implementation. Overly restrictive guardrails may impede the potential benefits of AI in health care.
- ▶ Managing change: Change management for AI in health care is a complex task. Implementing AI in health care should be managed carefully and involve all relevant stakeholders. Fearfulness and overenthusiasm can be addressed with careful change management

Mind the Gap please



News > Health News

DOCTOR WON'T SEE YOU SOON Computerswill replace doctors in just ten years, says Health Secretary Jeremy Hunt

Jeremy Hunt predicts patients will be diagnosed by machines by the time the NHS is 80 in 2028

Shaun Wooller

Published: 2:45, 13 Sep 2017





Babylon did not live up to expectations. One former employee, Hugh Harvey, a doctor, said that the company's "Al algorithm" was little more than a standard medical if/then "decision tree" set out in an Excel spreadsheet.

In 2017, the Care Quality Commission raised concerns.

In 2018, The Lancet concluded that there was no evidence that Babylon's chatbot worked better than a doctor, and there was "a possibility that it might perform significantly worse".

Assistive versus Autonomous AI in health care

	Assistive AI algorithms		Autonomous Al algorithms		
	Level 1	Level 2	Level 3	Level 4	Level 5
	Data presentation	Clinical decision-support	Conditional automation	High automation	Full automation
Event monitoring	Al	Al	Al	Al	Al
Response execution	Clinician	Clinician and Al	Al	AI	AL
Fallback	Not applicable	Clinician	At, with a backup clinician available at Al request	Al	Al
Domain, system, and population specificity	Low	Low	Low	Low	High
Liability	Clinician	Clinician	Case dependent	Al developer	Al developer
Example	Al analyses mammogram and highlights high-risk regions	Al analyses mammogram and provides risk score that is interpreted by clinician	At analyses mammogram and makes recommendation for biopsy, with a clinician always available as backup	Al analyses mammogram and makes biopsy recommendation, without a clinician available as backup	Same as level 4, but intended for use in all populations and systems



Thank You! Any Questions?