

Machine
Learning
Systems applied
to health data
and systems

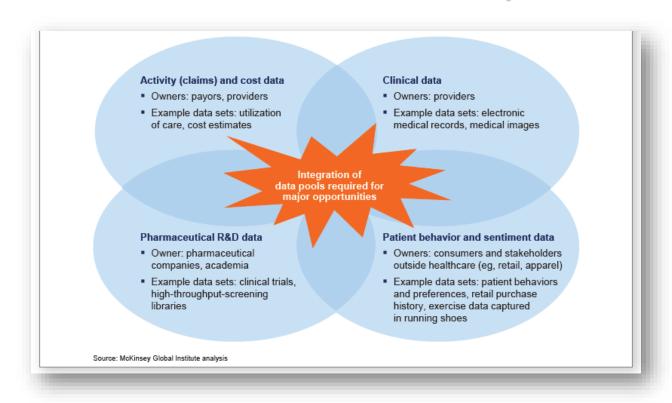
An overview on technological development, regulatory challenges and ethical issues

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The importance of **health data**, a potential **treasure** trove for medicine and pharmacology



It's a capital mistake to theorise before one has data

[Sherlock Holmes, A Study in Scarlett (A. C. Doyle) suggested by Leveraging Data in Healthcare, R.M.S. Busch CRC Press]



Sources for Health Data

Clinical Records

Imaging data

Claims data

Survey

Social Data

Internet of things (IoT)

Laboratory

Medical Device

Pharmacies

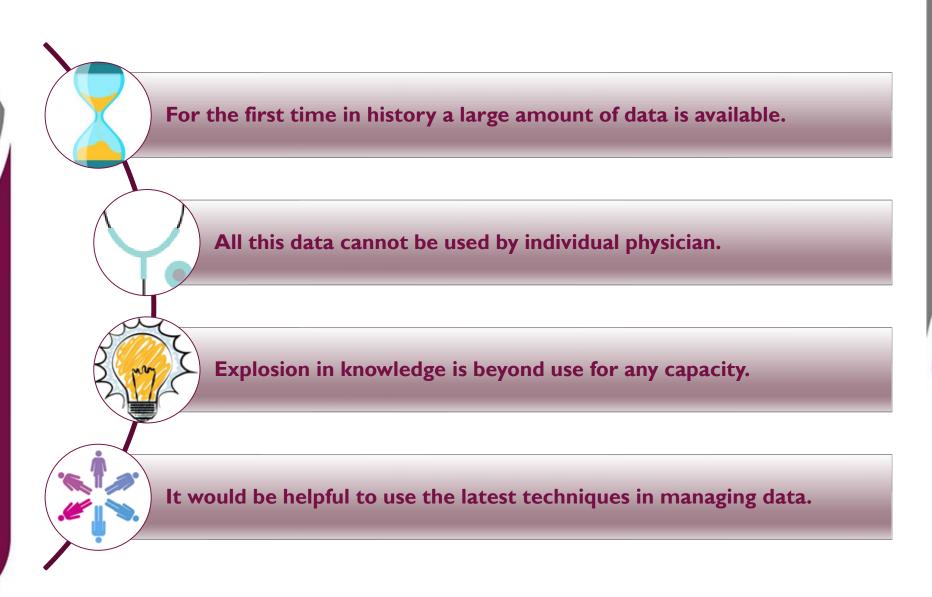
Clinical trials

"Omics" data (genome, transcriptome, metabolome, proteome, exposome, methylome and microbiome)

Social networks and apps



The Health Data Challenge







Processing all the available data to unlock their full potential in current clinical practice.



The amount and heterogeneity of Health data require new and powerful tools.

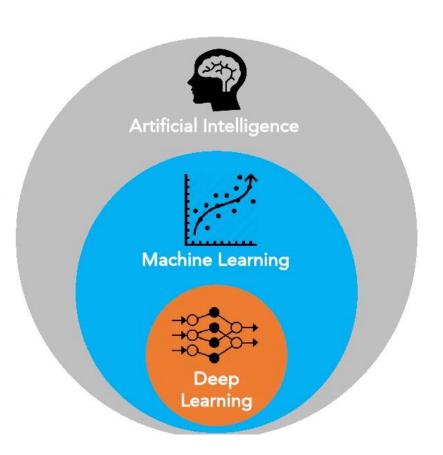


What tools does technology offer us to manipulate data and extract information?

Artificial Intelligence is a a branch of computer science dealing with the simulation of human intelligence processes. These processes include learning, recognizing and self-correction.

Machine Learning is a field of Artificial Intelligence, focused on getting machines to act without being programmed to do so.

Deep Learning is a class of machine learning algorithms that uses artificial neural networks (ANNs) with many layers of nonlinear processing units for learning data representations.

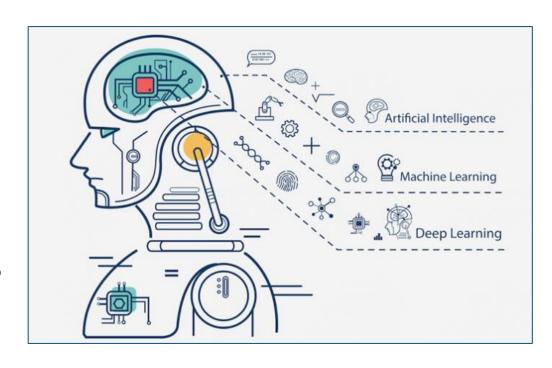




Why using AI techniques?

To **support** the decision-making process

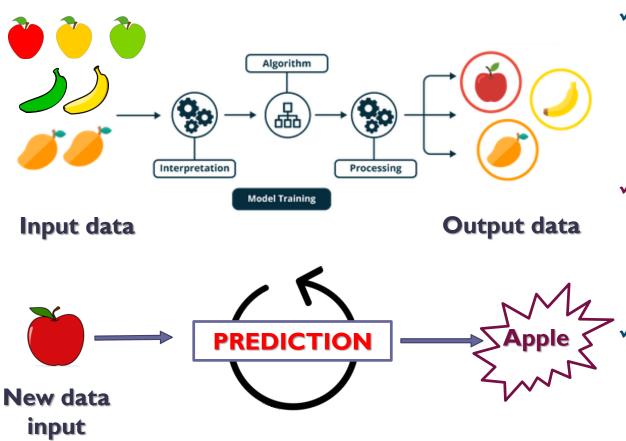
- ✓ better decisions
- ✓ in less time
- ✓ reducing human medical errors



...These processes include learning, recognizing and self-correction



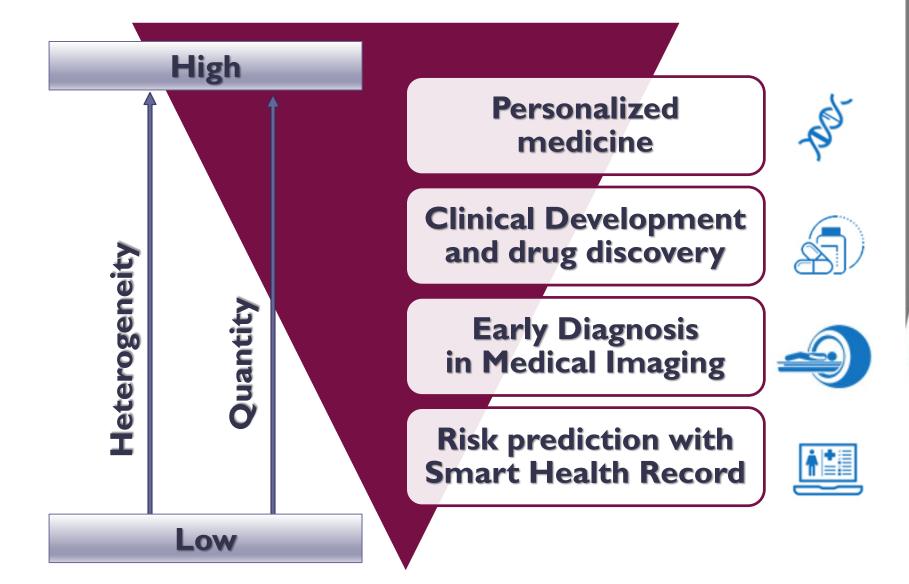
What is Machine Learning?



- ✓ Let's pretend to create a **system** that answers the question of whether a fruit is an apple, a banana or an apricot.
- This question is called "model" and the model is created via a process called training.
- ✓ After data training, the model is able to recognize a **new item** submitted to the system.

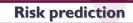


Current and future Machine Learning Health data applications



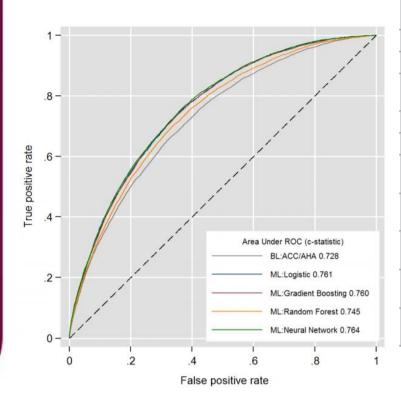






Machine Leaning offers opportunity to improve accuracy by exploiting complex interactions between risk factors.

Can it improve cardiovascular risk prediction?



RESEARCH ARTICLE

Can machine-learning improve cardiovascular risk prediction using routine clinical data?

Stephen F. Weng^{1,2 $\frac{1}{2}$ *, Jenna Reps^{3,4 $\frac{1}{2}$}, Joe Kai^{1,2 $\frac{1}{2}$}, Jonathan M. Garibaldi^{3,4 $\frac{1}{2}$}, Nadeem Qureshi^{1,2 $\frac{1}{2}$}}

| Machine-learning Algorithms | | | | |
|------------------------------------|------------------------------------|------------------------------------|------------------------------------|--|
| ML: Logistic Regression | ML: Random Forest | ML: Gradient Boosting Machines | ML: Neural Networks | |
| Ethnicity | Age | Age | Atrial Fibrillation | |
| Age | Gender | Gender | Ethnicity | |
| SES: Townsend Deprivation Index | | | Oral Corticosteroid Prescribed | |
| Gender | Smoking | Smoking | Age | |
| Smoking | HDL cholesterol | HDL cholesterol | Severe Mental Illness | |
| Atrial Fibrillation | HbA1c | Triglycerides | SES: Townsend Deprivation Index | |
| Chronic Kidney Disease | Triglycerides | Total Cholesterol | Chronic Kidney Disease | |
| Rheumatoid Arthritis | SES: Townsend Deprivation Index | HbA1c | BMI missing | |
| Family history of premature CHD | ВМІ | Systolic Blood Pressure | Smoking | |
| COPD | Total Cholesterol | SES: Townsend Deprivation Index | Gender | |









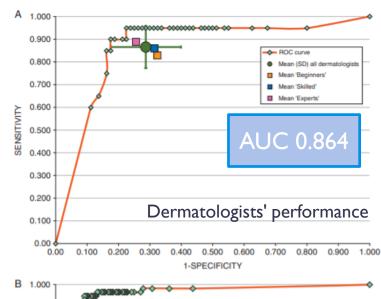


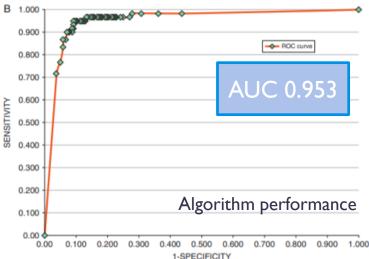
ORIGINAL ARTICLE

Man against machine: diagnostic performance of a deep learning convolutional neural network for dermoscopic melanoma recognition in comparison to 58 dermatologists

H. A. Haenssle^{1*,†}, C. Fink^{1†}, R. Schneiderbauer¹, F. Toberer¹, T. Buhl², A. Blum³, A. Kalloo⁴, A. Ben Hadi Hassen⁵, L. Thomas⁶, A. Enk¹ & L. Uhlmann⁷

- The aim was to train, validate, and test a deep learning CNN for the diagnostic classification of dermoscopic images of lesions of melanocytic origin (melanoma, benign nevi) and to compare the results to an international group of 58 dermatologists (beginners, skilled and expert).
- A dataset of 100 images was collected created by the International Skin Imaging Collaboration (ISIC) melanoma project.



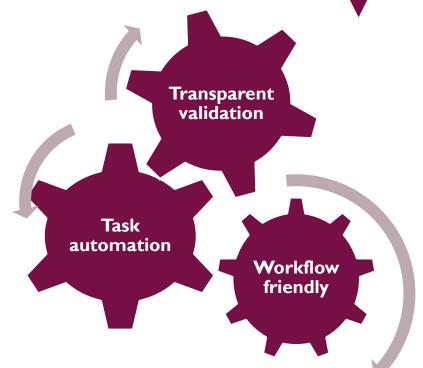


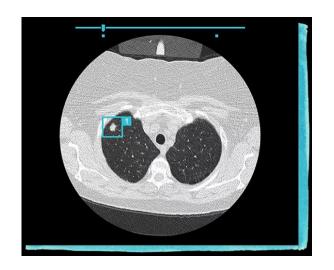




Early Diagnosis

- ✓ Aidance offers a software called Veye Chest which can help radiologists to detect, track and report on pulmonary nodules using machine learning.
- ✓ The pre-trained algorithm detects the presences or warning signs of pulmonary nodules within patient CT scans.





✓ Veye Chest has received the CE marking, allowing its use in clinical practice across the EU.





An overview of Machine Learning in the European Pharmaceutical Industry







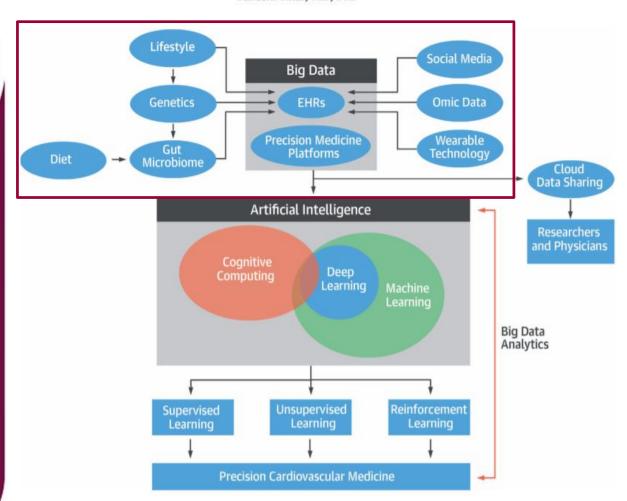
- ✓ Algorithms for predicting which molecules would best bind to form stable compounds that can be applied to new drugs
- ✓ Software for discovering small molecules and compounds that could treat single and bispecific target diseases
- ✓ Drug discovery platform for developing rare disease treatments



Artificial Intelligence in Precision Cardiovascular Medicine



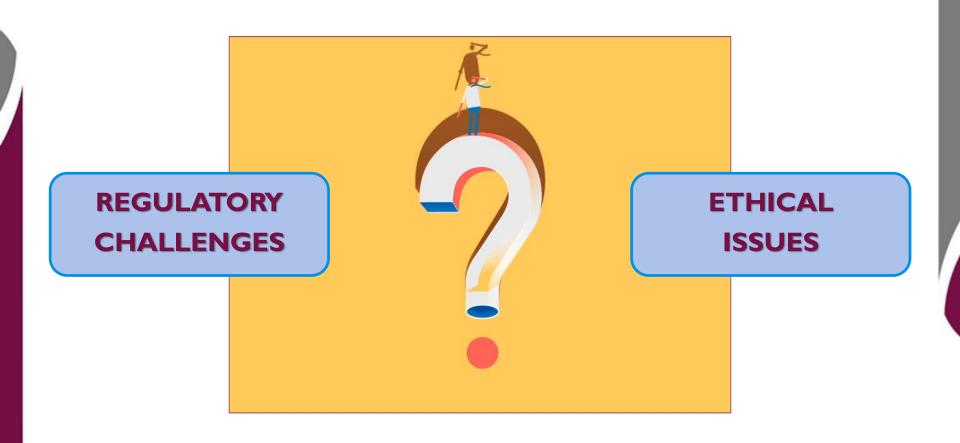
Chayakrit Krittanawong, MD, a,b HongJu Zhang, PhD,c Zhen Wang, PhD,de Mehmet Aydar, PhD,b,f Takeshi Kitai, MD, PhDb,g



Big data can be stored through EHRs or precision medicine platforms and can be shared for data analysis with other physicians or researchers through secure cloud systems. Big data analytics using artificial intelligence (machine learning, deep learning, or cognitive computing) and 3 main types of learning algorithms will enable precision cardiovascular medicine.



Questions for this new technological Era





Regulatory Challenges

Privacy

Some relevant EU provision, such as General Data Protection regulation, are applicable anyway even if GDPR could results in undesirable restrictions in the number and type of data to be accumulated.

Standardization

Standards can be developed to align with and support existing regulatory frameworks, whilst keeping pace with evolving technologies.

Safety

What would be the right protocol to work out a quality assurance plan, a risk identification and management approach or how to prove safety and efficacy?

Medical Device

Some of the previously mentioned concerns could be, at least partially, resolved in case the ML software is classified as a "Medical Device", a category covered by EU/national provisions.



REVIEW





Artificial intelligence as a medical device in radiology: ethical and regulatory issues in Europe and the United States

Filippo Pesapane 1 · Caterina Volonté 2 · Marina Codari 3 · Francesco Sardanelli 3,4

- instrument, apparatus, appliance, software, implant, reagent, material or other article intended by the manufacturer to be used, alone or in combination, for human beings for one or more of the following specific medical purposes:
 - diagnosis, prevention, monitoring, prediction, prognosis, treatment or alleviation of disease
 - ... injury or disability
 - investigation, replacement or modification of the anatomy or of a physiological or pathological process or state

The EU approach

| Table 1 | Regulatory | framework in the EU on medical devices |
|-----------|------------|--|
| Directive | 02/42/EEC | Directive on medical devices |

Directive 93/42/EEC Directive on medical devices

Will be replaced by MDR on 26 May 2020

MEDDEVS Non-binding guidelines on legislation related to

medical devices

MDR Regulation on medical devices

Applies from 26 May 2020 Repeals Directive 93/42/EEC

IVDR Regulation on in vitro diagnostic medical devices

Applies from 26 May 2022

MDR, Medical Device Regulation; IVDR, In Vitro Diagnostic Medical

Device Regulation; EEC, European Economic Community



Ethical Issues

Black box negligence

Medical malpractice and product liability that arise with the use of "black-box" algorithms: users cannot provide a logical explanation of how the algorithm arrived at its given output

Trust

Many patients may not feel comfortable with a machine making potentially life and death decisions about Healthcare.

Self correction and automatic decision

The AI choice may not always correspond to the most ethical choice.

Bias

The implicit racial, gender or other biases of the humans that code the algorithms or the data that is fed into the algorithms can skew the results. Data may not be representative of the population.



