

# AI and Healthcare

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## AI in healthcare - benefits and issues

# Why AI in healthcare?

- AI has the **potential to transform healthcare** since healthcare is producing **a large amount of clinical and administrative data**
- **This large amount of data can be used for analysis**
- Moreover, research studies have shown that **AI can carry out many key healthcare activities better than, or as well as, humans**, such as **diagnosing diseases**, for example by **analyzing radiology images**

# AI is sparsely implemented

- Today, AI solutions are **sparsely implemented in practical healthcare**
- Existing AI solutions are **mainly supporting the individual functions in healthcare, like radiology and pathology image analysis**

# Why is AI sparsely implemented? 1(2)

According to Davenport and Kalakota (2019), **two major reasons for AI being sparsely implemented in practical healthcare** are:

- **AI solutions are focusing on limited tasks** and are **rarely integrated into the clinical processes**
- Moreover, **AI is not implemented in electronic record systems (EHR)**. Therefore, AI is not part of the system that most healthcare personnel use for their day-to-day work

# Why is AI sparsely implemented? 2(2)

Panch et al. (2014) **add additional important reasons for AI being sparsely implemented in practical healthcare are:**

- **Healthcare systems are complex and fragmented, and will not easily change as a result of new technology**
- **Healthcare organisations lack the capacity to collect the necessary training data of sufficient quality - while also respecting ethical principles and legal constraints**

# AI technologies in healthcare

# AI technologies in healthcare

Note, according to Devenport and Kalakota (2019), AI is not one technology, but rather a collection of them.

Examples of AI technologies:

- Machine learning
- Natural language processing
- Rule based expert system
- Physical robots
- Robotic process automation



# Machine learning

- **Traditional machine learning** is the most common application in healthcare. This application is mostly **using supervised learning**, which **requires a training datasets** to be used to be able to do the work
- Supervised learning systems are **supporting the making of diagnosis**, and **predicting what treatment protocols are likely to be successful** for a patient, **based on various patient attributes and the treatment context**

# Neural network and deep learning

- A more complex form of supervised machine learning is the **neural network**. **Neural network** make use of a **network of variables** that **associate inputs with outputs** and create **weights on these associations, in order to predict outcome**
- A **neural network** can also have **variables on many different so called hidden layers**, called **deep neural network** or **deep learning**
- Deep learning has been very successful for identifying **clinically relevant features in imaging data** - beyond what can be perceived by the human eye
- **Deep learning** is also increasingly used for **speech recognition in NLP**, see next slide

# Natural language processing

- **Natural language processing (NLP) aims to make sense of human language. NLP includes application such as speech recognition, text analysis, translation.**
- In healthcare, NLP can, for example, be used for **analyzing unstructured clinical notes** and supporting the **transformation from speech to text**

# Rule based expert system 1(2)

- **Rule based expert system** - require human experts and knowledge engineers to **construct a series of rules in a particular knowledge domain, which will be the base for the expert system**
- **Rule based expert systems in healthcare - are the base for many clinical decision support system**
- **Rule based expert systems - are also be part of many medical record systems (i.e. EHR systems), for example, they provide **functionality to warn for drug-to-drug interactions**, and **support the physician of making diagnoses****

# Rule based expert system 2(2)

## The limitation of rule based expert systems:

- Rule based expert systems - **work well if the rules are not so many.**
- However, **if number of rules is over several thousand, it is hard to maintain the rules**, for example, **the rules soon start to conflict with each other.**
- Moreover, **if the knowledge domain changes, rules need to change, which may be time-consuming**, especially if the rules are many, and related on each other
- Therefore, **due to this limitations, rule based expert systems are being replaced by systems based on ML algorithms**

# Physical robots

- **Physical robots** - perform pre-defined tasks in factories and warehouses, like lifting and assembling objects
- Applied in healthcare are **surgical robots** – which can **improve the surgeons ability to see and make tasks more precise**
- **Moreover, physical robots are also becoming more intelligent**, as other **AI capabilities are being embedded in their operating systems.**

# Robotic process automation (RPA)

- **Robotic process automation (RPA) – record the keyboard and mouse actions of a human being, and repeat these actions automatically**
- **RPA does not involve physical robots – instead RPA is a form of software**
- **RPA act like a semi-intelligent user of the systems, following a script or a set of rules based on actions done by human beings**
- **RPA can be used in healthcare for updating patient records, billing or other administrative tasks**
- **Moreover, RPA can be used in combination with other technologies, for example combining image recognition and RPA, where RPA can be used for extract data from the recognitions of images and update EHR system with this data**

# AI technologies can be combined

- **AI technologies are being more and more combined and integrated**, for example:
  - physical robots are getting AI-based features
  - image recognition is being integrated with RPA.



## AI application areas in healthcare

# AI application areas

## Example of AI application areas in healthcare:

- Diagnosis and treatment
- Patient engagement and adherence
- Administrative activities

# Diagnosis and treatment 1(3)

- **IBM's Watson has received a lot of attention** for its **application in diagnosis and treatment area**, particularly **cancer diagnosis and treatment**
- Watson consisted **of a set of 'cognitive services'**, employing a combination of **machine learning and NLP technologies**
- However, **IBM's Watson's application in healthcare has not been a success:**
  - Watson has **not been able to handle different types of cancer**
  - Watson has also **been hard to integrate into care processes and systems**

# Diagnosis and treatment 2(3)

## Other examples of the use of AI for diagnosis and treatment:

- Several organizations **work on ML based solutions** to better **understand the how different genetic variants of humans will response to different treatments, such as drugs and protocols.**
- Organizations are also **working on ML based solution to predict populations at risk of particular diseases, high-risk conditions or to predict hospital readmission**

# Diagnosis and treatment 3(3)

## Drawbacks of using AI in the application of diagnosis and treatment:

- **To embed AI-based diagnosis and treatment recommendations into clinical workflows and EHR systems has not been successful**
- According to Davenport and Kalakota (2019), **“such integration issues have probably been a greater barrier to broad implementation of AI than any inability to provide accurate and effective recommendations”**



# Patient engagement and adherence 1(2) Stockholms universitet

- **Patients engagement in their own well-being and care are important for receiving better outcome** in healthcare
- **The major problem is that the patient may not make necessary behavioral adjustment, that is, does not follow a course of treatment or take the prescribed drugs**

# Patient engagement and adherence 2(2)

- Therefore, ML and business rules engines can be used to support patient engagement and adherence, by:
  - sending message alert to patients,
  - providing targeted content given the patients' status and characteristics,
  - tailoring recommendations by comparing patient data to other effective treatment pathways for similar cohorts
  - nudging patient behavior in a more anticipatory way

# Administrative activities

Different AI technologies can be used for administrative tasks:

- RPA can be used for a variety of applications in healthcare, like managing medical records
- NLP can be applied in chatbots for patient interaction
- ML could be used to verify whether millions of insurance claims are correct, for example, by applying probabilistic matching of data across different databases



## Healthcare workers

# Implication for healthcare workforce 1(2)

- According to Davenport and Kalakota (2019) estimate that **it will take 20 years before will see any substantial change in healthcare employment due to AI**
- **Instead, there is also the possibility that new jobs for working with AI technologies are created**

# Implication for healthcare workforce 2(2)

- The area where **most healthcare jobs will be automated** are those dealing with **digital information, radiology and pathology**
- **However, for example, not even radiologist jobs will not disappear in the near future, and maybe not in the long term either – see next slides**

# Implication for radiology 1(2)

- Today, **radiology AI systems can only perform single tasks.**
- **Radiology AI systems cannot fully identify all potential findings in medical images.**  
**Radiologist are still needed for that**
- **Radiologists also do a lot of other thing than just read and interpret images:**
  - radiologists **relate findings from images to other medical records and test results**
  - radiologists **consult with other physicians** regarding diagnosis and treatment
  - radiologists **discuss procedures and results with patients**
  - radiologists **define the technical parameters of imaging examinations.** The parameters need to be tailored to the patient's condition

# Implication for radiology 2(2)

- **Moreover, for employing full scale AI-based image work:**
  - **clinical processes need to be changed**, which will take time
  - **an aggregated repository of radiology images is required for training the AI system**, but such **an aggregated repository is lacking today**
  - **changes in medical regulation and health insurance contracts for automated image analysis are needed**

## A brief summary

# To summarize 1(2)

- The **greatest challenge to AI** is to **ensure its adoption in daily clinical practice.**
- **There are a number of challenges to overcome to achieve this.**
- Therefore, Davenport and Kalakota (2019) **estimate that we will see a limited use of AI in clinical practice within 5 years and more extensive use within 10**

## To summarize 2(2)

- Moreover, **“AI systems will not replace human clinicians on a large scale, but rather will augment their efforts to care for patients”**.
- According to Davenport and Kalakota (2019) **it might take 20 years before will see any substantial change in healthcare employment**