
Multimedia applications and artificial intelligence for digital health -- some standardization perspective--

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Some history...

- Healthcare and medicine were one of the first applications, along with natural language understanding, of Artificial Intelligence in the 1960s and 1970s when the so-called expert system became popular (e.g., MYCIN)
- But they did not achieve widespread use, due to:
 - Lack of sufficient computing power, for processing and storage
 - Lack of sufficient large volume of data
 - Lack of digitization (e.g., machine readable health record, MRI)
 - Lack of sufficient supporting technologies (e.g., ultra-high definition video, computer vision, natural language processing)



But now..

- Improvements in computing power resulting in faster data collection and data processing
- Increased volume and availability of health-related data (including text)
- Implementation of electronic health record (EHR) systems
- Development of natural language processing and understanding technology, for large volume
- Improvements in high-definition video and computer vision technology, enabling machines to have better “eyes”
- Growth of genomic sequencing databases

Importance of Sensor, Data and Analysis

- Central to the current trend in AI are:
 - Sensor devices such as camera, microphone, LiDAR, etc.
 - Data, derived from sensor devices, but also from other sources like web-pages, catalogues, EHR, etc. May be structured as well as unstructured
 - Analysis: inspecting, cleansing, transforming and modeling data in order to discover useful information, esp. for decision-making. Machine-learning is an important part of data analysis to predict relations, attributes, etc., of the data.
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Types of Data

- Sensor data
 - Video and graphic (images of skin cancer, eye scan, brain scan, etc.)
 - Audio/sound/speech
 - Heat/temperature
 - Textual data
 - Typing mistakes, grammatical error
 - documents,
 - EHR, EMR, PHR, etc.
 - Questionnaire
 - Numerical data
 - Age, blood pressure, heart rate, etc.
 - Time-series data
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- IQ

Importance of Imaging in medical/health

- “Some statistics say that up to 80 percent of all medical diagnoses are made or confirmed through **imaging studies**” (Shinjini Kundu)
- Techniques in medical imaging include:
 - Magnetic resonance imaging (MRI)
 - Radiography
 - (Functional) near-infrared spectroscopy (fNIRS)
 - Tomography (e.g., CT)
 - Ultrasound
 - Angiography
 - Photoacoustic imaging
 - ...
- It is clear that imaging data will be one of the key areas where AI can be applied.

Brain Healthcare and AI

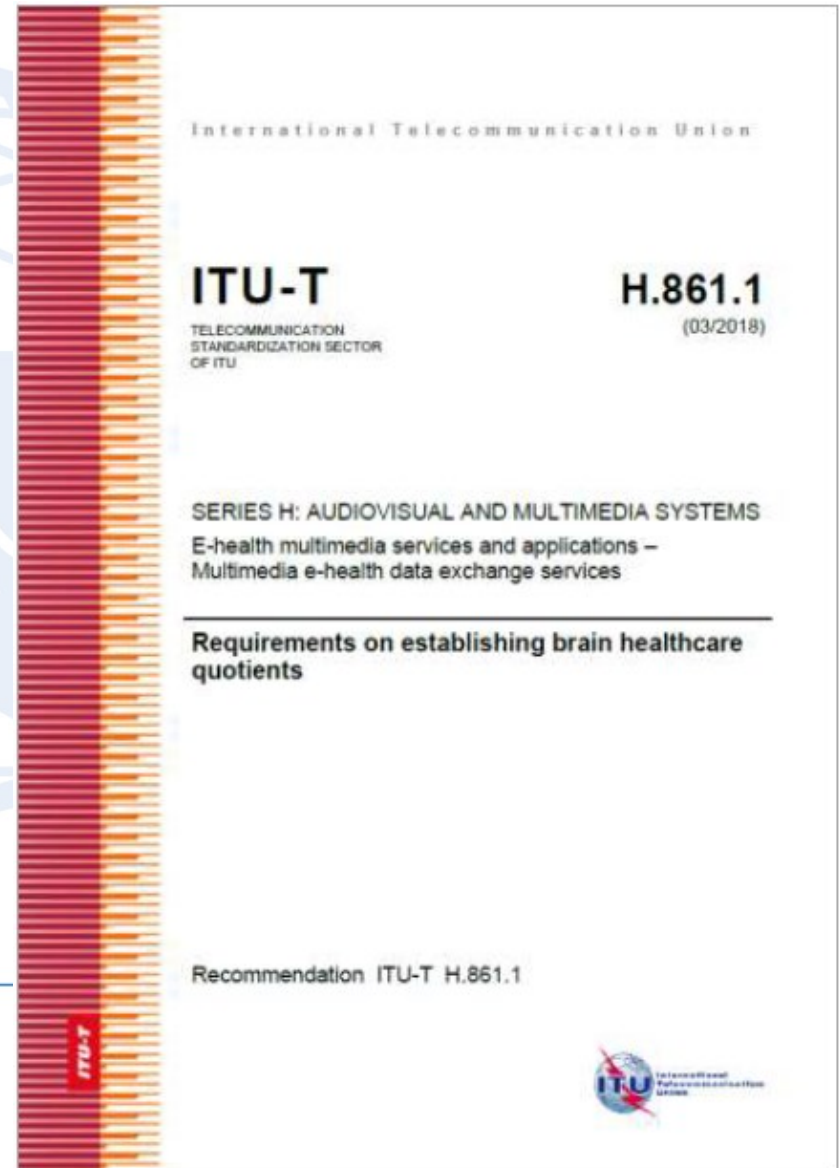
- As an example of using imaging data for healthcare, we introduce here the case of Brain Healthcare Quotient (BHQ) and its use with AI for help improve the QoL (quality of life) of an aging population.

ITU Multimedia Standards for Digital Health



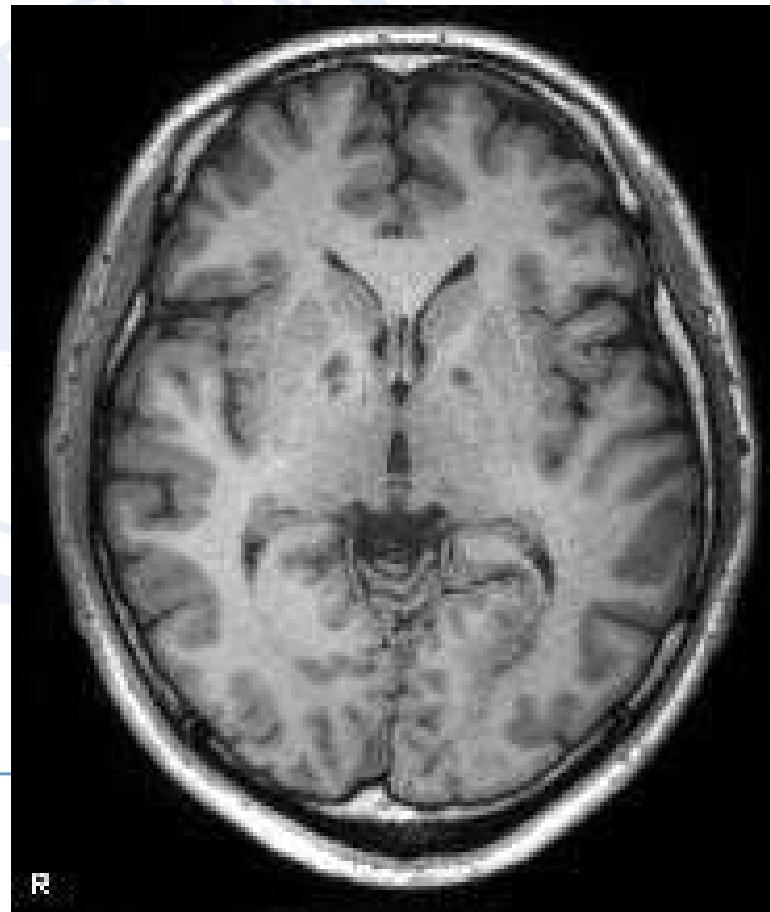
Brain Healthcare Quotients (BHQ)

- ITU-T SG16 has standardized ITU-T Rec. 861.1“, defining quotient for brain health, called “Brain Healthcare Quotient” based on the calculation on MRI scanning



Brain Image

- Brain scanning is the most popular, if not most effective, means to detect abnormality in the brain, including cognitive diseases



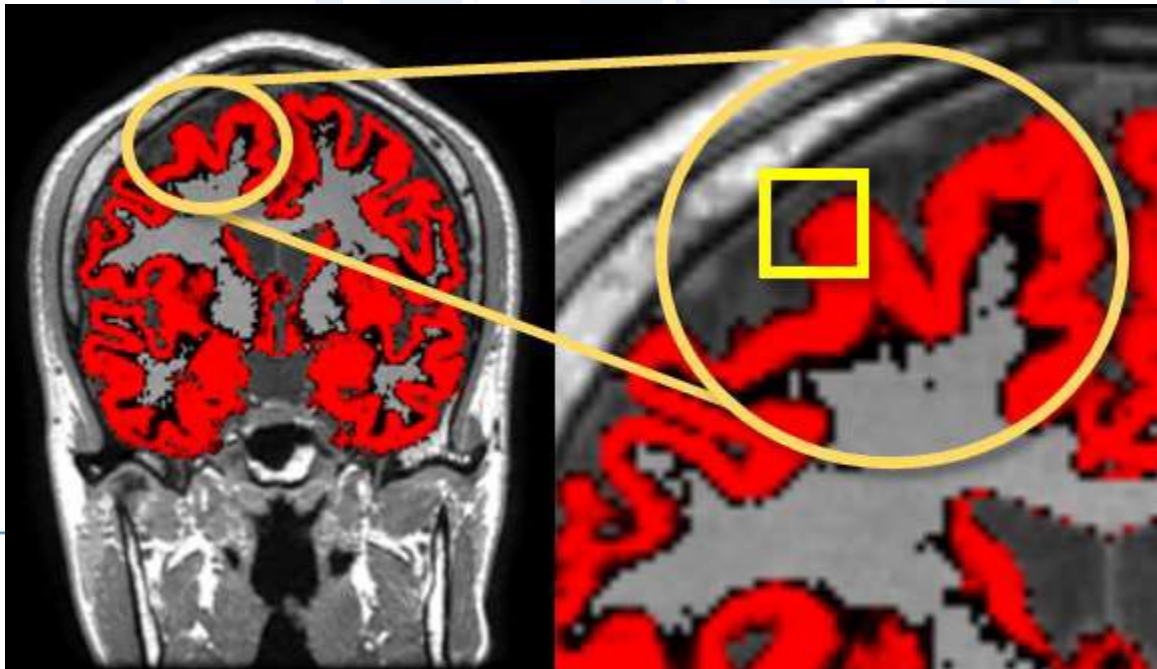
Brain Healthcare Quotients (BHQ)

- A numerical indicator representing physical characteristics of the brain that are purported to be indicative of some state of a health related condition
- Currently primarily used to calculate the volume of brain gray matter (GM)
- It has the following characteristics:
 - detect minute changes in the brain of a person in the course of time
 - the exchange of information about the brain in a standardized way
 - on the basis of neuroscience.
 - derived from neuroimaging analysis.
 - built upon an established method to assure its validity
 - employ a standardized score (e.g., mean of population is 100 and 1 standard deviation is 15, so that 95% of population falls within the range of 70-130.).



GM-BHQ

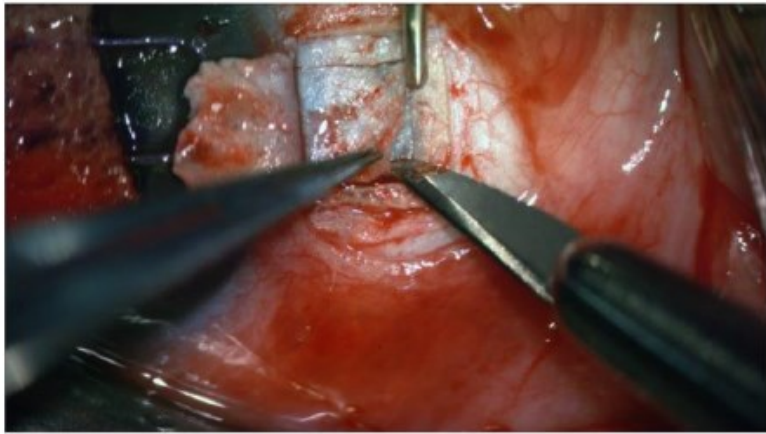
- Evaluates volume of gray matter where a high density of neuronal cells can be found.
- • Gray matter is assessed by VBM (Voxel Based Morphometry)



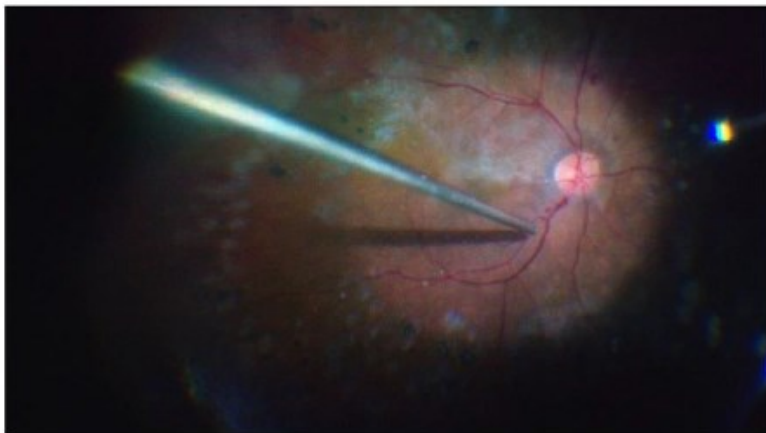
Using AI to estimate BHQ

- These correlations mean that there is a possibility of estimating BHQ by analyzing the life style of a patient.
- With the big database of “brain dry dock” examinees, and associated questionnaires, an AI algorithm can estimate the rough range of BHQ.
- As MRI scanning is still expensive and time-consuming, such an estimate BHQ will make the usefulness of such a process even relevant in the prevention of cognitive diseases, as well as healthy life style
- This project is being undertaken by NTT Data and Kyoto University

Use of Ultra High Definition (UHD) Video in Endoscopy and Microscopy



a) Glaucoma surgery by 8K microscope



b) Retinal surgery by 8K microscope

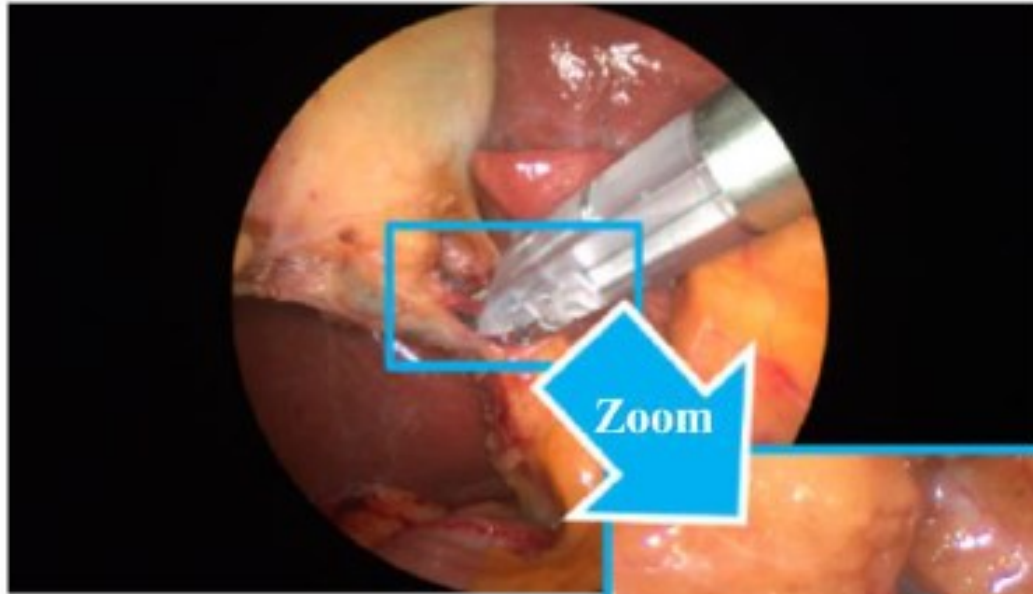
- UHD Endoscope and Microscope, as defined in ITU-T F.780 allows doctors see objects 3.8 larger than they normally can, leading to detection of formerly e.g., “unobservable” tumors or making operation easier.

Improvement of Endoscopy using UHD and AI

- With the advent of ultra high definition broadcasting and video technology, digital imaging in medicine is experiencing a new era of digital medicine
- ITU-T SG16 has defined ITU-T F.780 for UHD imaging for telemedicine
- 8K UHD imaging allows video with a 16-times higher resolution (7680×4320 pixels, about 33 million pixels) than the current high-definition (HD) (1920×1080 pixels)
- Endoscopic surgery, where surgeons operate by watching a lesion on a video screen with a camera in the body, requires video imaging
- Aided by AI, such UHD technology will enhance the accuracy of detection of as well as operation on critical conditions such as cancer



Use of Ultra High Definition (UHD) Video in Endoscopy and Microsurgery (2)



Digital zoom
× 1.0 ~ 4.0

UHD Microsopy
allows zooming up
to 4.0 times the
original size



×4.0





Real-time endoscopic diagnosis system using AI






- The system can discover colorectal cancer and precancerous lesions (colon neoplastic polyps) in real time during endoscopic examination. The PC-based system uses AI technology utilizing deep learning for image analysis, high speed processing algorithms, and an advanced image processing device.
- 5,000 endoscopic scan images were provided by the National Cancer Research Center Central to NEC for AI learning.
- Using this AI technique resulted in a cancer detection rate of 98%.

Demonstration

ファイル名 経過時間/トータル時間

 1秒あたりの処理フレーム数



| | |
|---|---------------------------|
|  | 検知時間: 00:00:06 確度: 0.2 % |
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Further Points for Standardization in AI and Multimedia for Digital Health

- Common data formats, platform for interoperability and assessment
- Reference metrics
- Guidelines on good practice, ethics and security
 - WHO Guideline on AI
 - HSCA (Healthcare Supply Chain Association) Guideline on Cybersecurity
- Common knowledge database for wide usage

Conclusion

- In (application of AI in) healthcare and medicine, imaging is important as up to 80 percent of all medical diagnoses are made or confirmed through imaging
- MRI, Endoscopy and microscopy are examples of imaging in medicine that can be used with AI to enhance services
- ITU-T has standards for them
- Additional common issues for standardization should be explored, such as commonalities, interoperability, security and ethics