

An Introduction to Medical Imaging and Biophotonics

& an intro to selected works from Prof. AUDREY K. BOWDEN

3D reconstruction of cystoscopy videos for comprehensive
bladder records, *Biomedical optics express 2017*

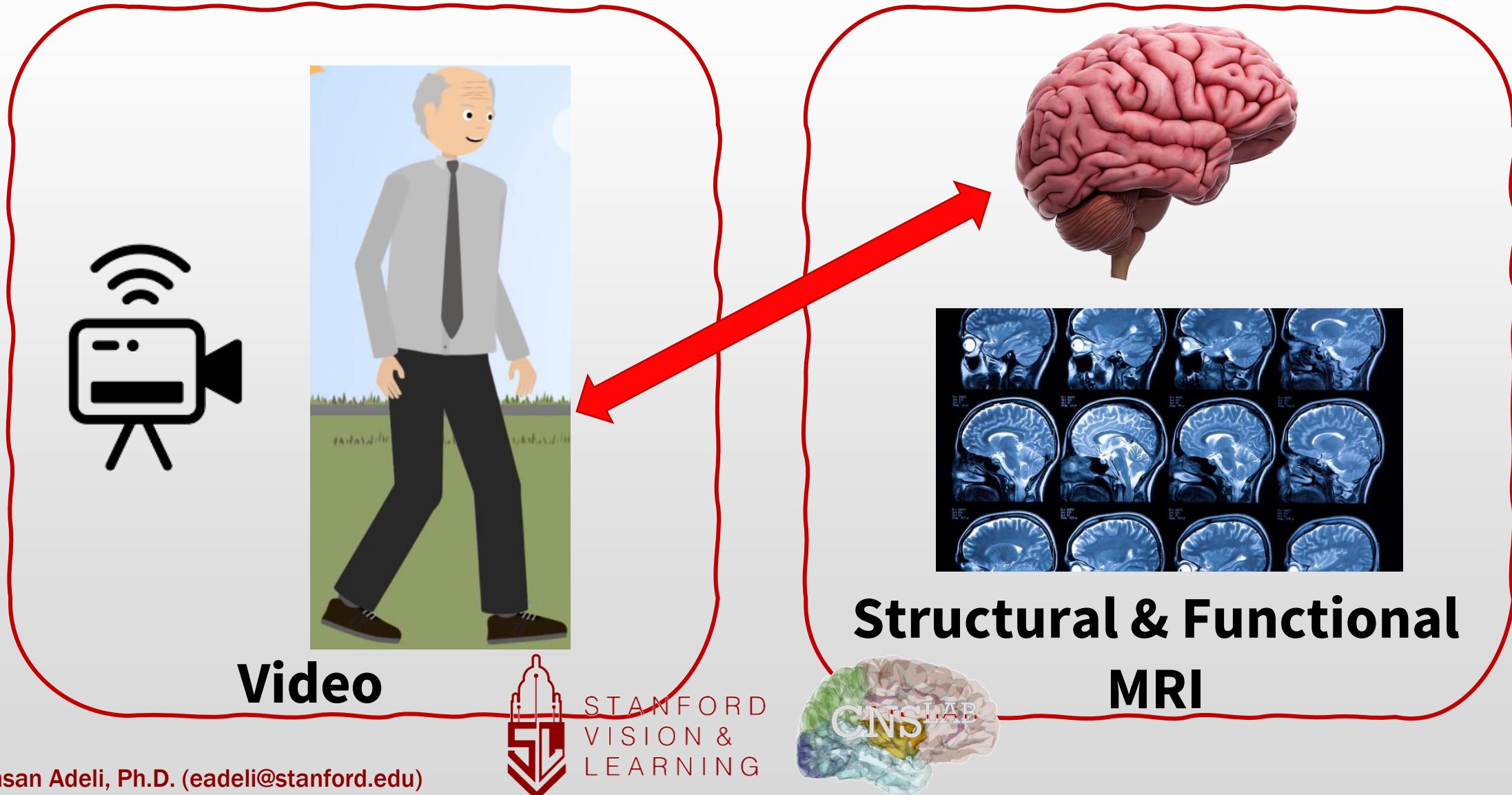
Ehsan Adeli, Ph.D.

CS114/CS214

May 18, 2022



Motion + Brain Circuitry





3D reconstruction of cystoscopy videos for comprehensive bladder records

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Prof. Audrey K Bowden

- Dorothy J. Wingfield Phillips Chancellor Faculty Fellow and Associate Professor of Biomedical Engineering (BME) and of Electrical Engineering (EE) at Vanderbilt University
- Prior to this, she served as Assistant and later Associate Professor of Electrical Engineering and Bioengineering at Stanford University.
- Research Focus
 - Biophotonics (light-based) tools for applications to medicine and biology, such as for early detection, diagnosis and therapy for cancer;
 - Development and deployment of low-cost, high-performing point-of-care technologies for rural and global health applications





IMAGING IN MEDICINE



Imaging

- Everyone knows about cameras...



- What else might you be interested in “imaging”?

Medical Imaging ca. 1895



Need to find a way to see inside without “light”

X-Ray

- November 8th, 1895, German scientist **Wilhelm Roentgen** was conducting experiments in his laboratory on the effects of cathode rays.
- the effect of passing an electrical discharge through gases at a low pressure.
- when passing current through the cathode ray, rays are given off that passed through different materials.



he name it X-RAYS
X meaning unknown

electromagnetic radiation of short wavelength produced when high-speed electrons strike a solid target

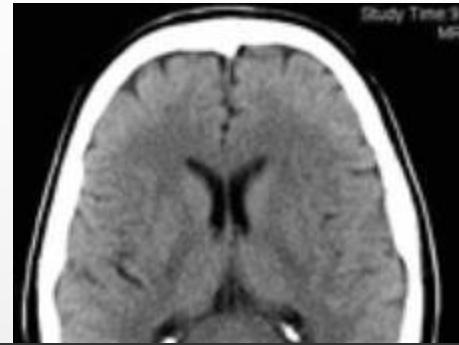


Medical Imaging Today

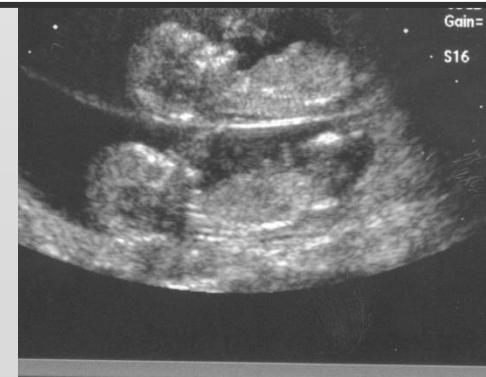
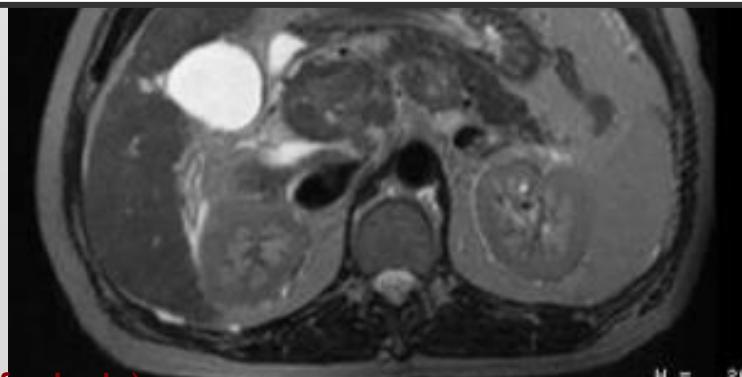
X-Ray



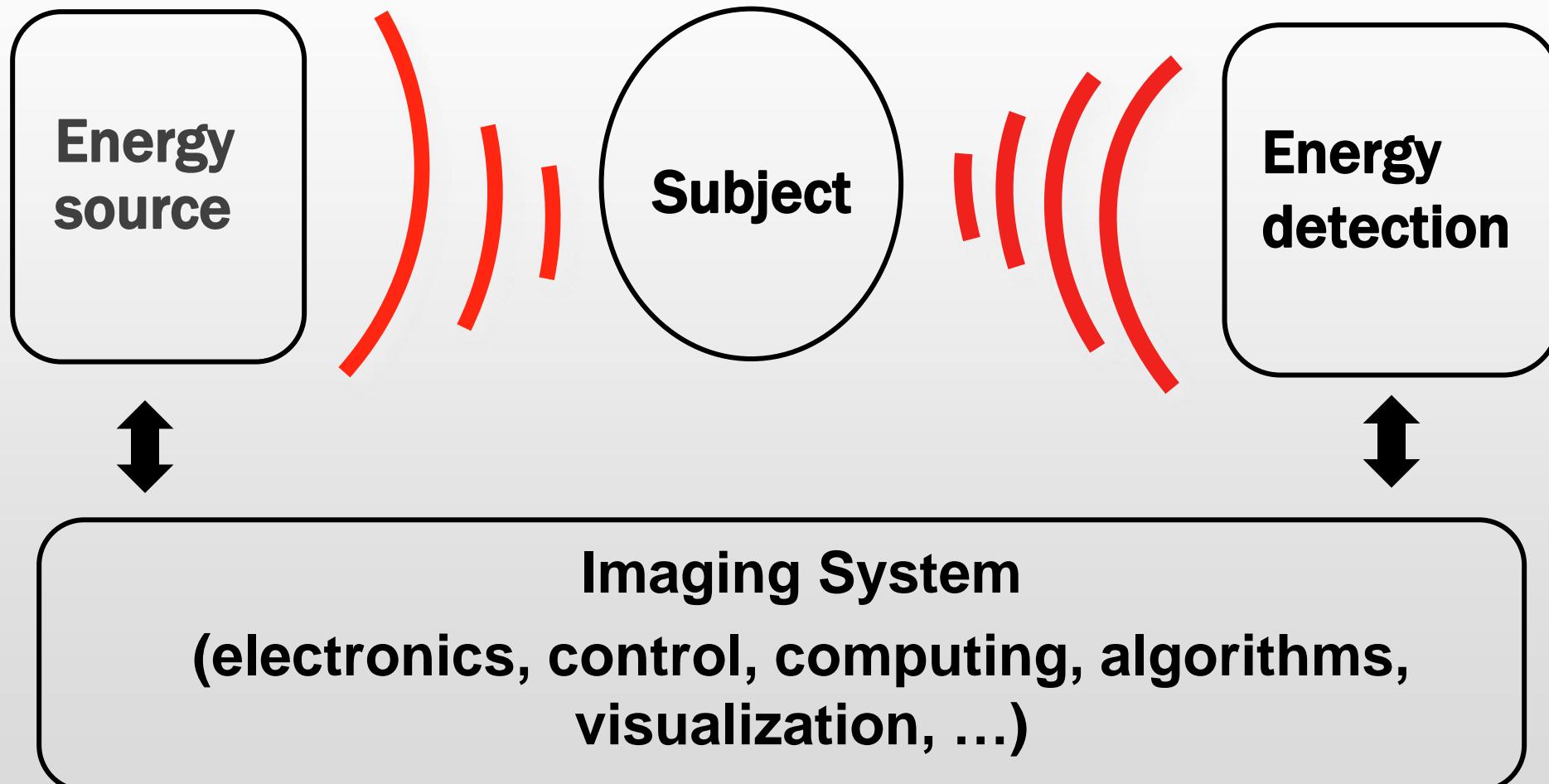
CT



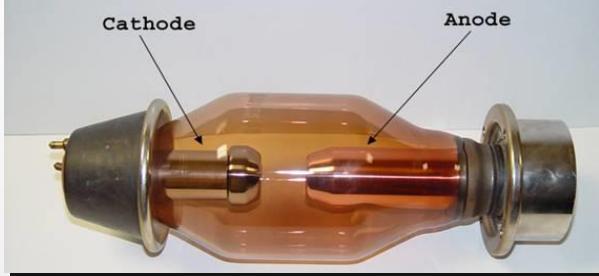
**All of these were enabled/dramatically advanced
by the mathematical and hardware design
techniques you will learn here!**



Imaging In General

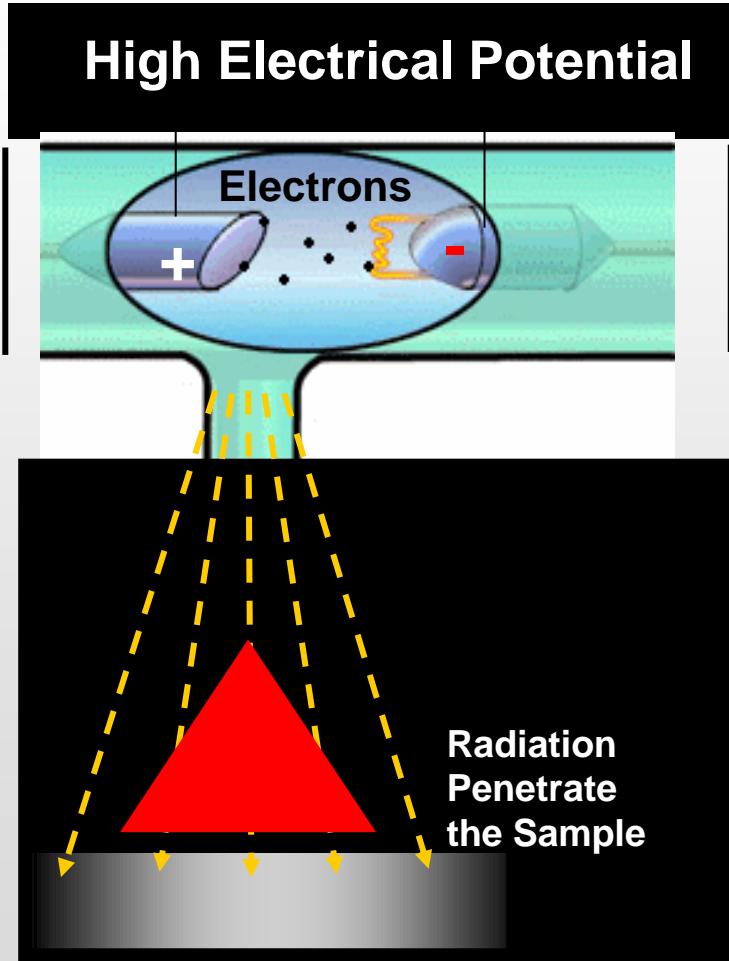


Conventional X-ray Imaging. X-ray Production.

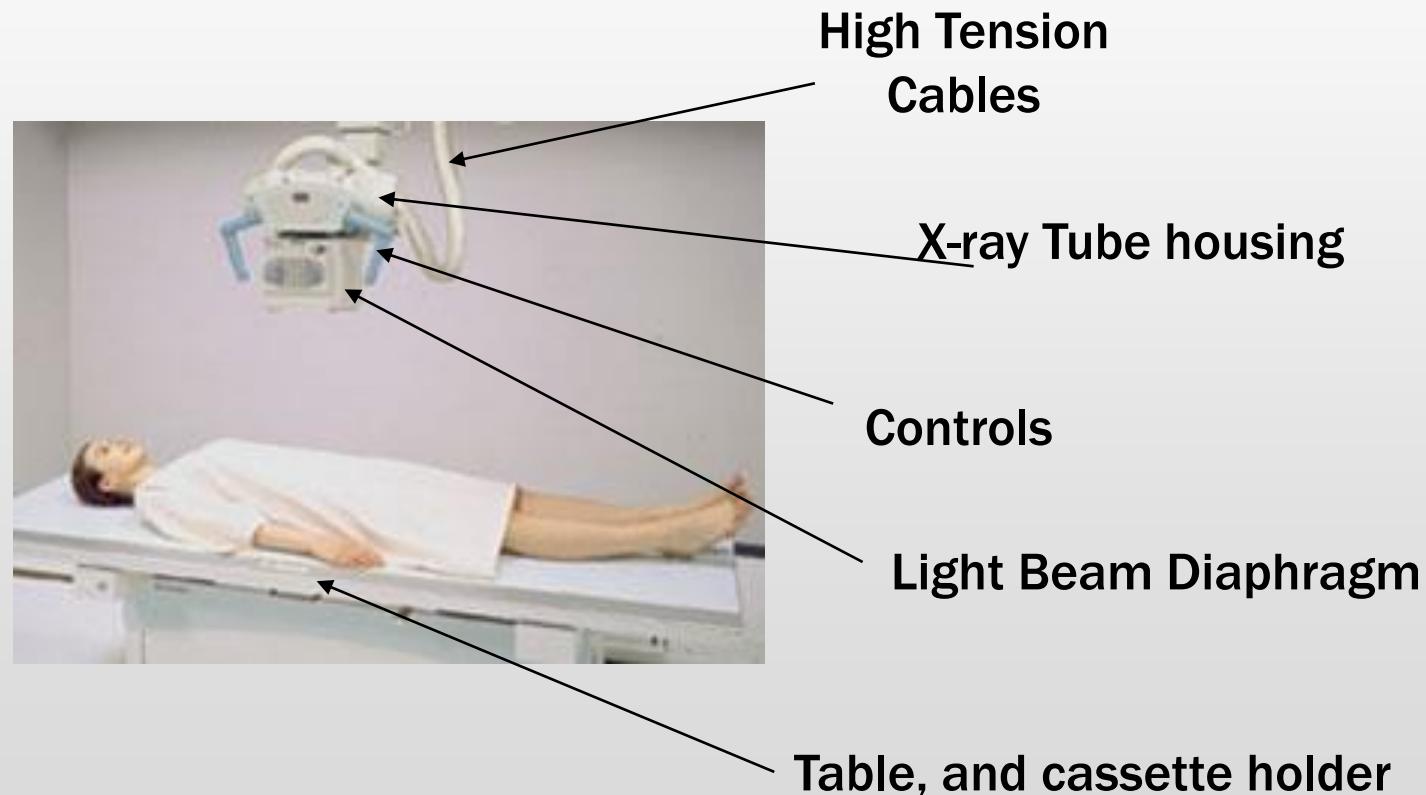


Electrons from cathode filament are accelerated towards and impact the rotating anode.

Rapid deceleration produces heat (~ 98%) and x-rays (~2%)



Over couch X-ray Tube and Table



Conventional X-ray Image of a Hand



Normal



Arthritic



Chest X-ray



Computerized Tomography (CT)

combines special x-ray equipment with sophisticated computers to produce multiple images of the inside of the body.



CT Scanner

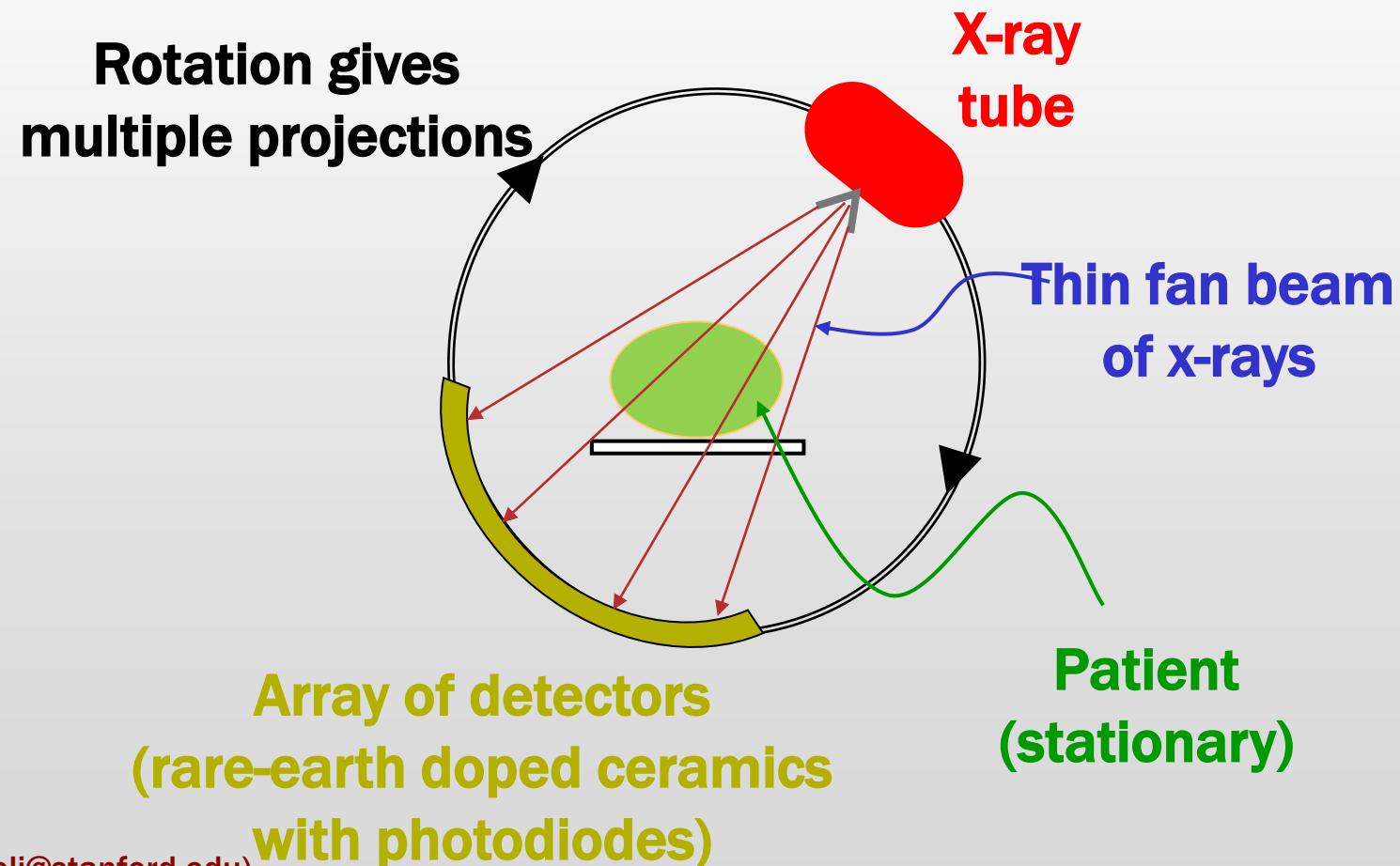


Brain Axial Image





Computed Tomography (CT)

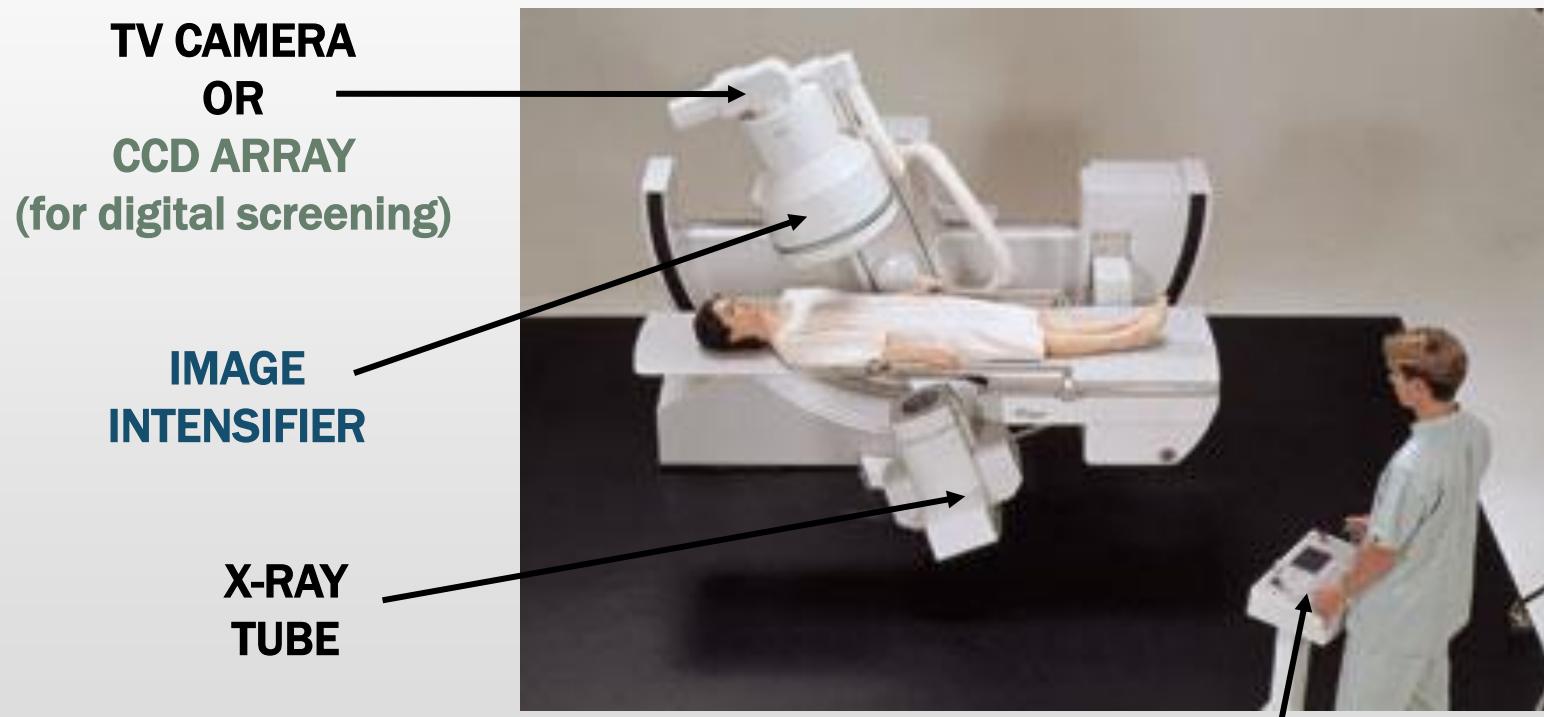


Fluoroscopy

- moving body structures
- A continuous x-ray beam is passed through the body part and is transmitted to a monitor



Fluoroscopy

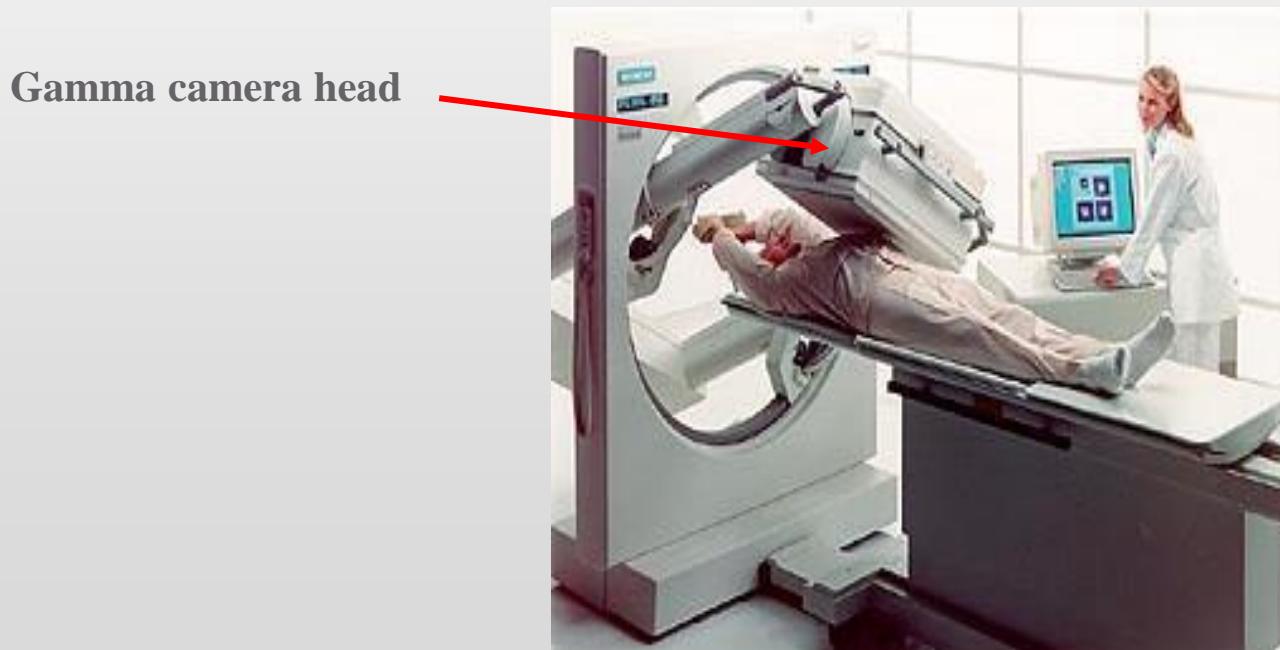


Allows dynamic imaging of
blood vessels (angiography)
and 'interventional' procedures

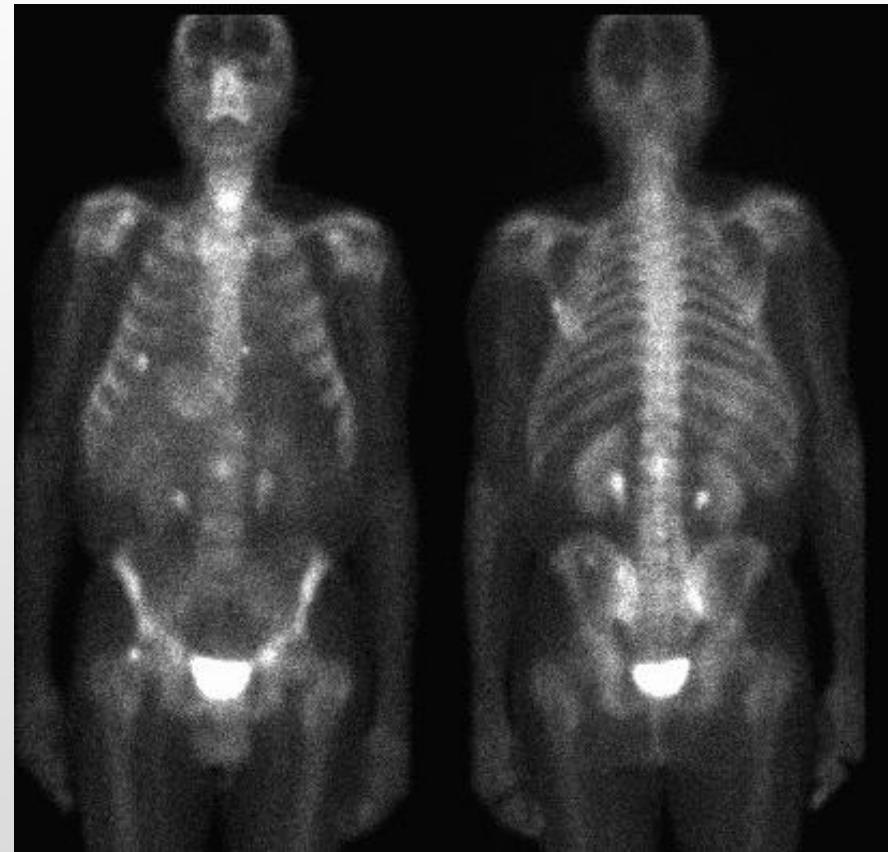


Radioisotope Imaging, Nuclear medicine (NM)

- the formation of images provides information about the function of various organs in the body,
- using internally administered radioisotopes as a radiation source.
- locate tumors or cancers and to examine the flow patterns of body fluids.



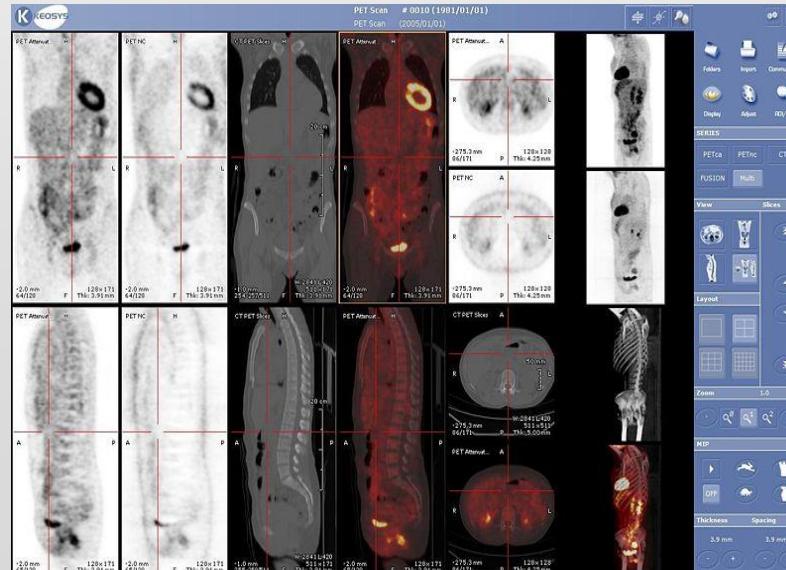
Gamma Camera Scan



Radioisotope Imaging

Positron Emission Tomography (PET)

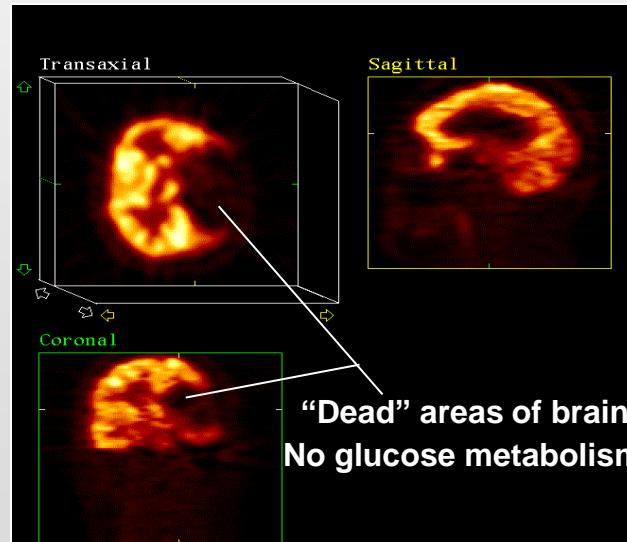
- a nuclear medicine imaging technique
- produces a 3D images of functional processes in the body



Radioisotope Imaging

Single Photon Emission Computed Tomography (SPECT)

- a nuclear medicine tomographic imaging technique using gamma rays
- similar to conventional nuclear medicine planar imaging using a gamma camera.
- Able to provide true 3D information.



Human Brain - Stroke



Ultrasound Imaging (US)

- common diagnostic medical procedure that uses high-frequency sound waves to produce images (sonograms) of organs, tissues, or blood flow inside the body.
- involves using a transducer, which sends
 - a stream of high-frequency sound waves into the body
 - and detects their echoes as they bounce off internal structures.
- The sound waves are then converted to electric impulses.



Ultrasound Transducer



Ultrasound Image of 19 Week Old Foetus



3D Ultrasound



Magnetic Resonance Imaging (MRI)

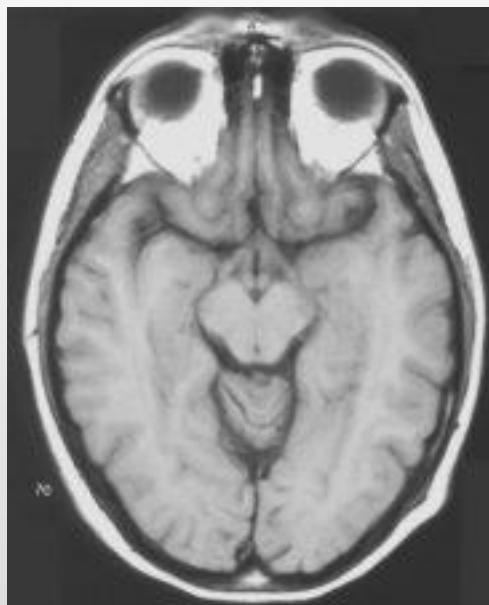
- uses a powerful magnetic field, radio frequency pulses, and a computer
- produce detailed pictures of organs, soft tissues, and all other internal body structures.



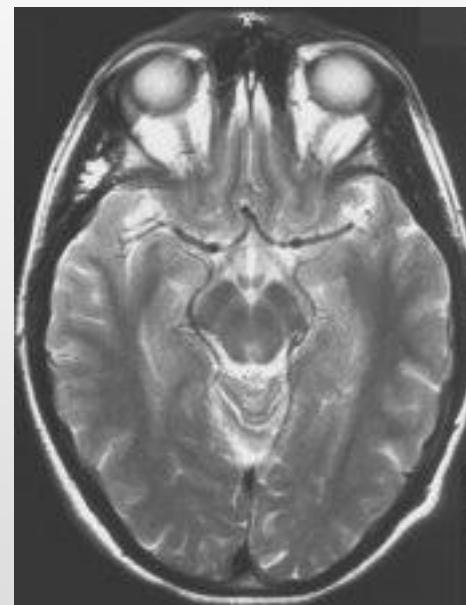
- ★ Big superconducting magnet (~ 1.5 tesla).
- ★ Gradient coils.
- ★ Radiofrequency coils.

Magnetic Resonance Imaging (MRI)

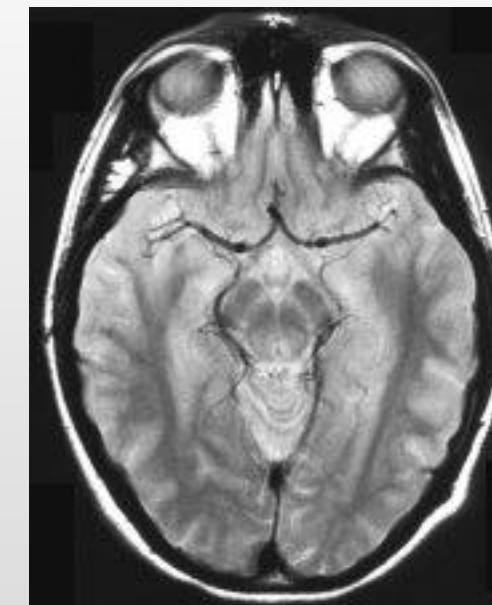
Axial Brain Images



T_1 -weighted



T_2 -weighted



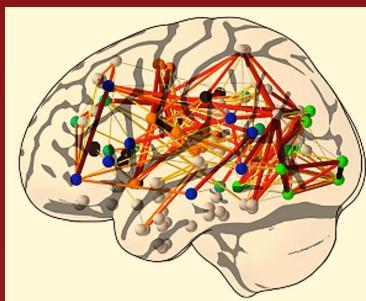
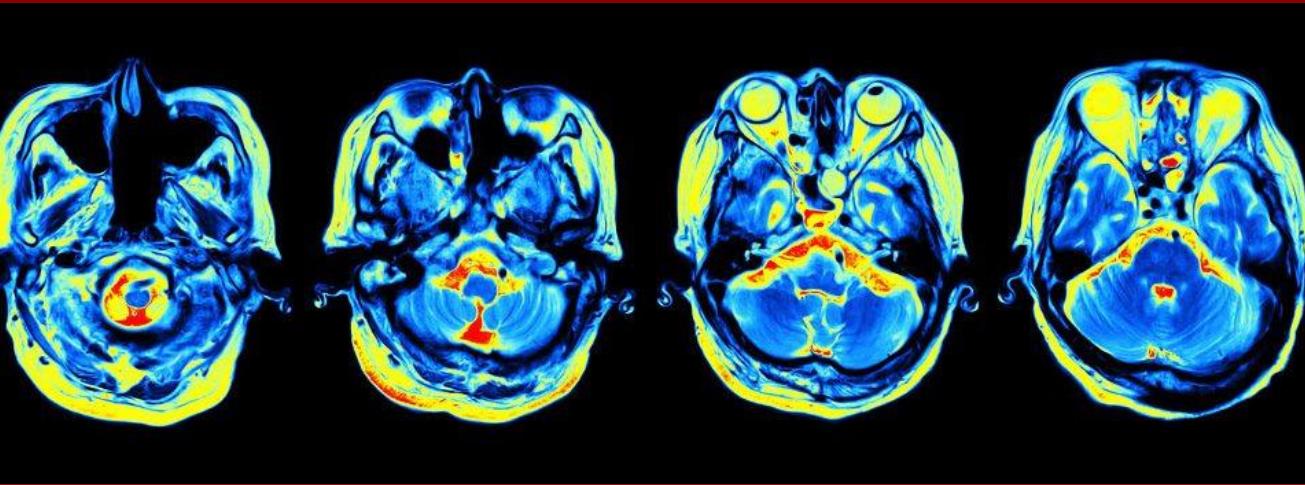
Proton density
weighted



Safety

Modality	Radiation Type	Comments
X-ray imaging	Ionising Radiation	Biological effect need protection against unnecessary doses
Radioisotope scanning		
Ultrasound Imaging	Non-ionising Radiation	Less harmful effects. Better for the foetus.
MRI		

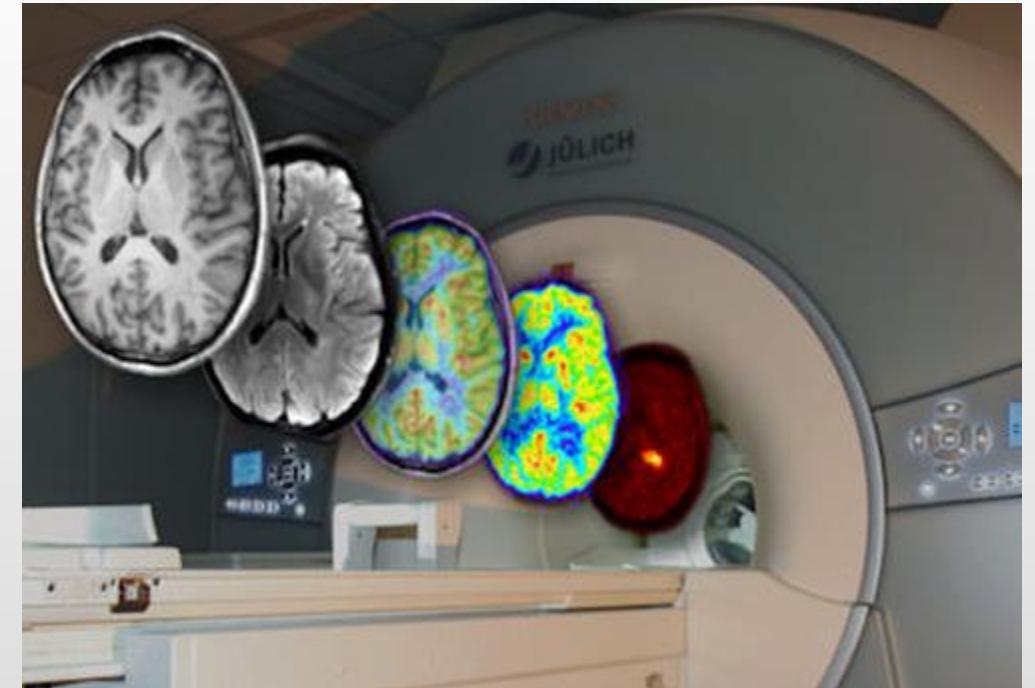




NEUROIMAGING

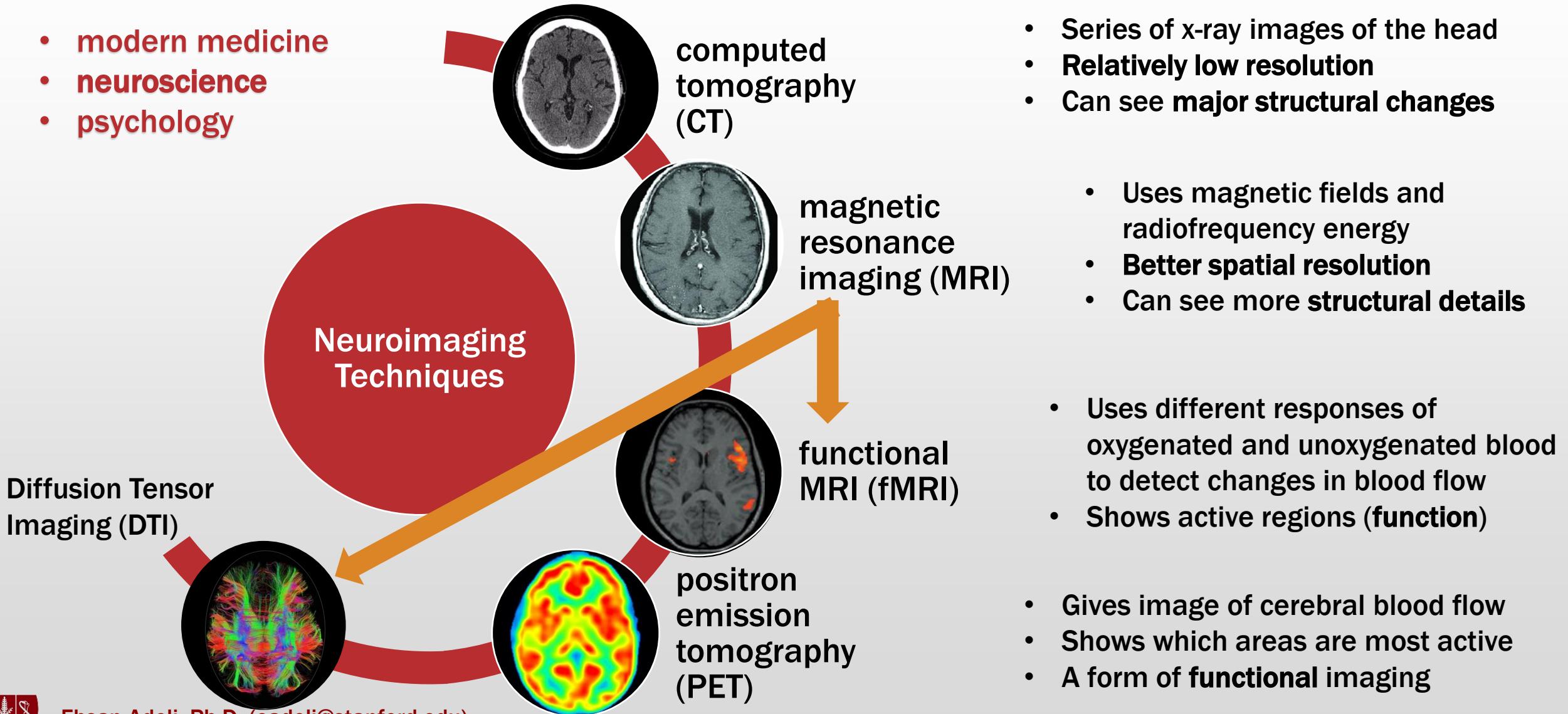
What is considered Neuroimaging?

Neuroimaging, or brain scanning, is a process of producing images of the brain or other parts of the nervous system. Current neuroimaging techniques are typically able to show both the structure and the functions of the brain.



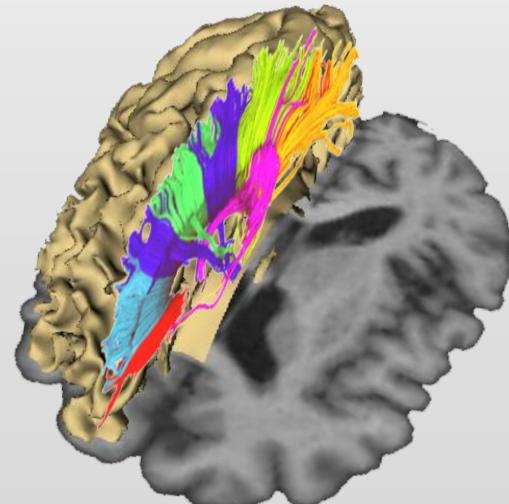
Neuroimaging types/modalities

- modern medicine
- neuroscience
- psychology

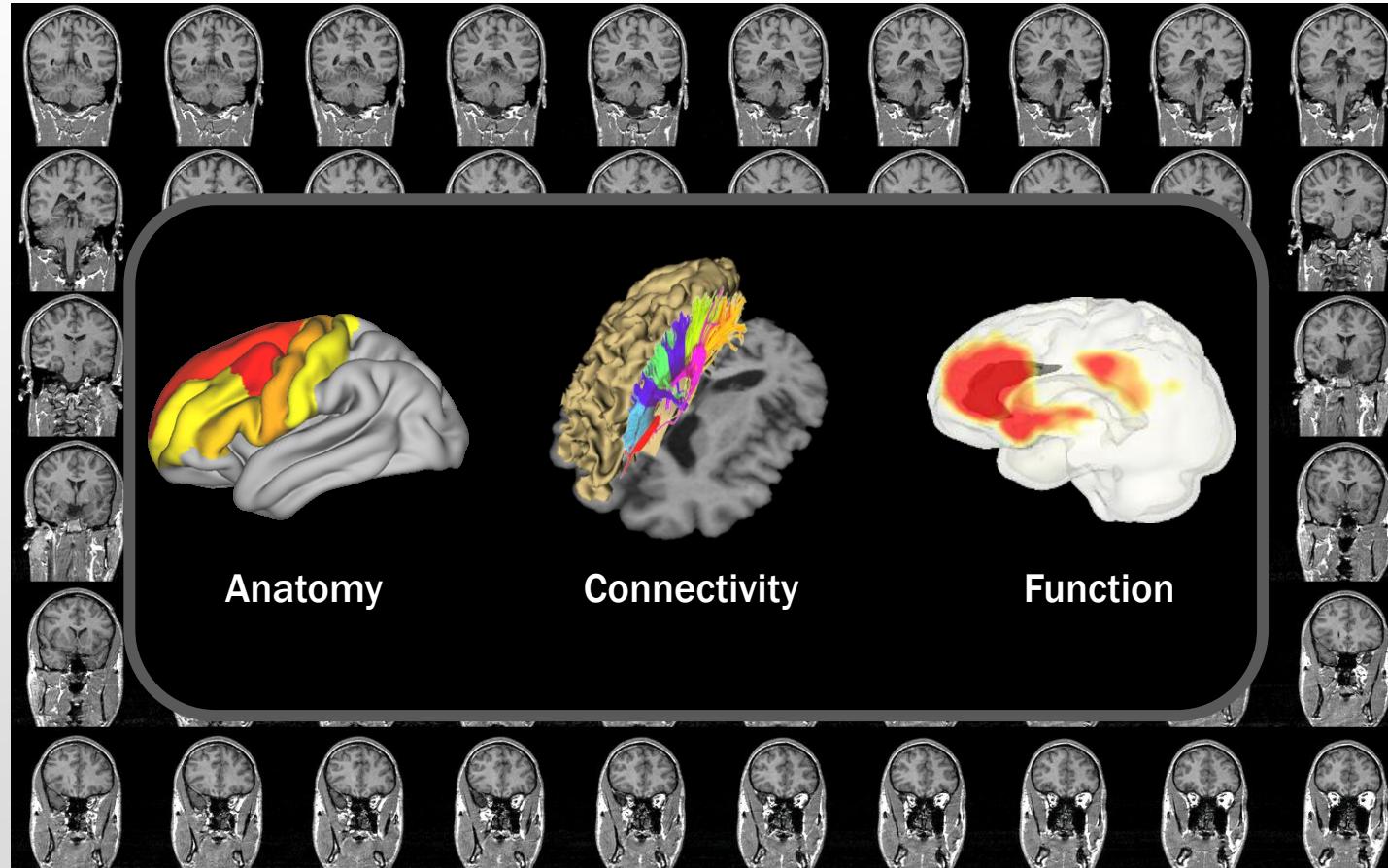


Goal

**Identify biomedical phenotypes
improving the mechanistic understanding,
diagnosis, and treatment
of neuropsychiatric disorders.**



Acquire Brain MRIs

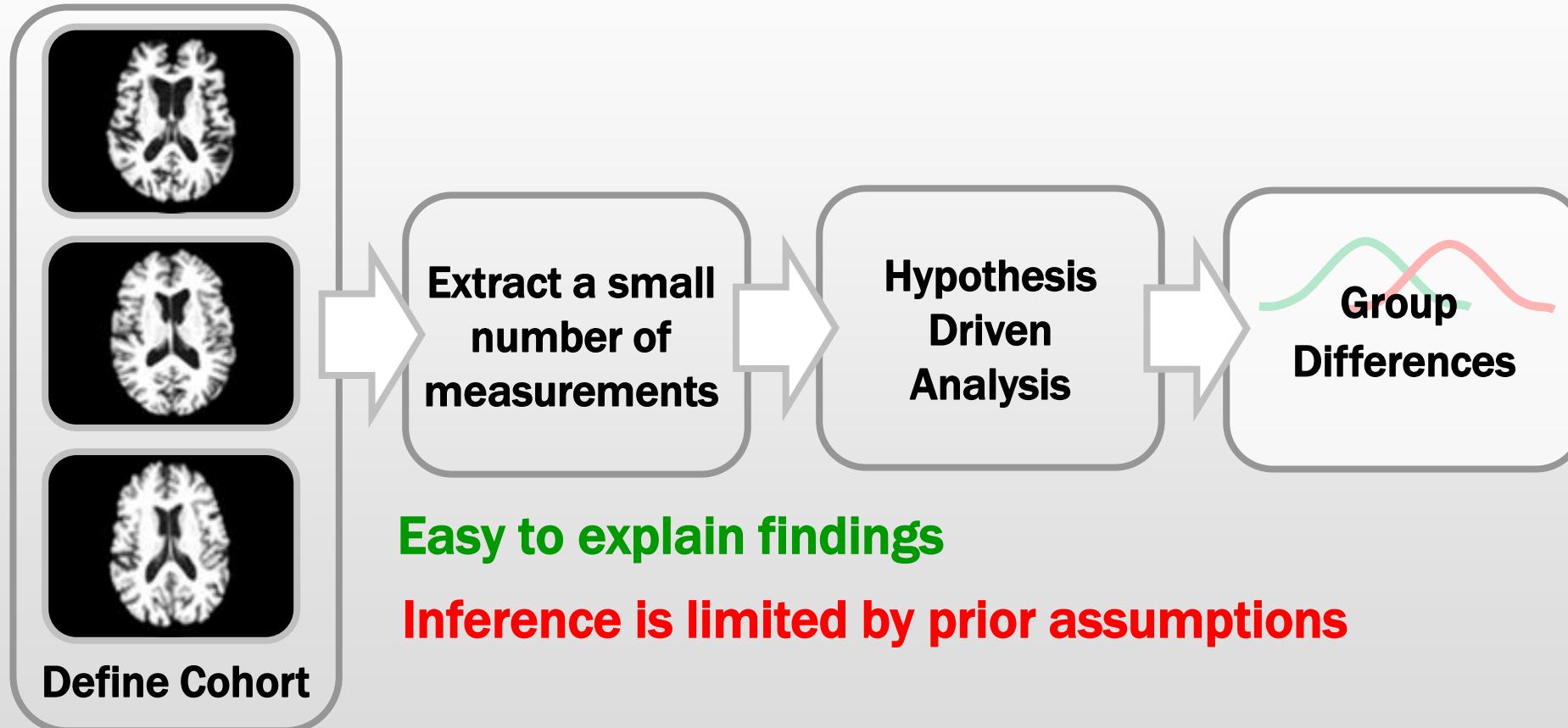


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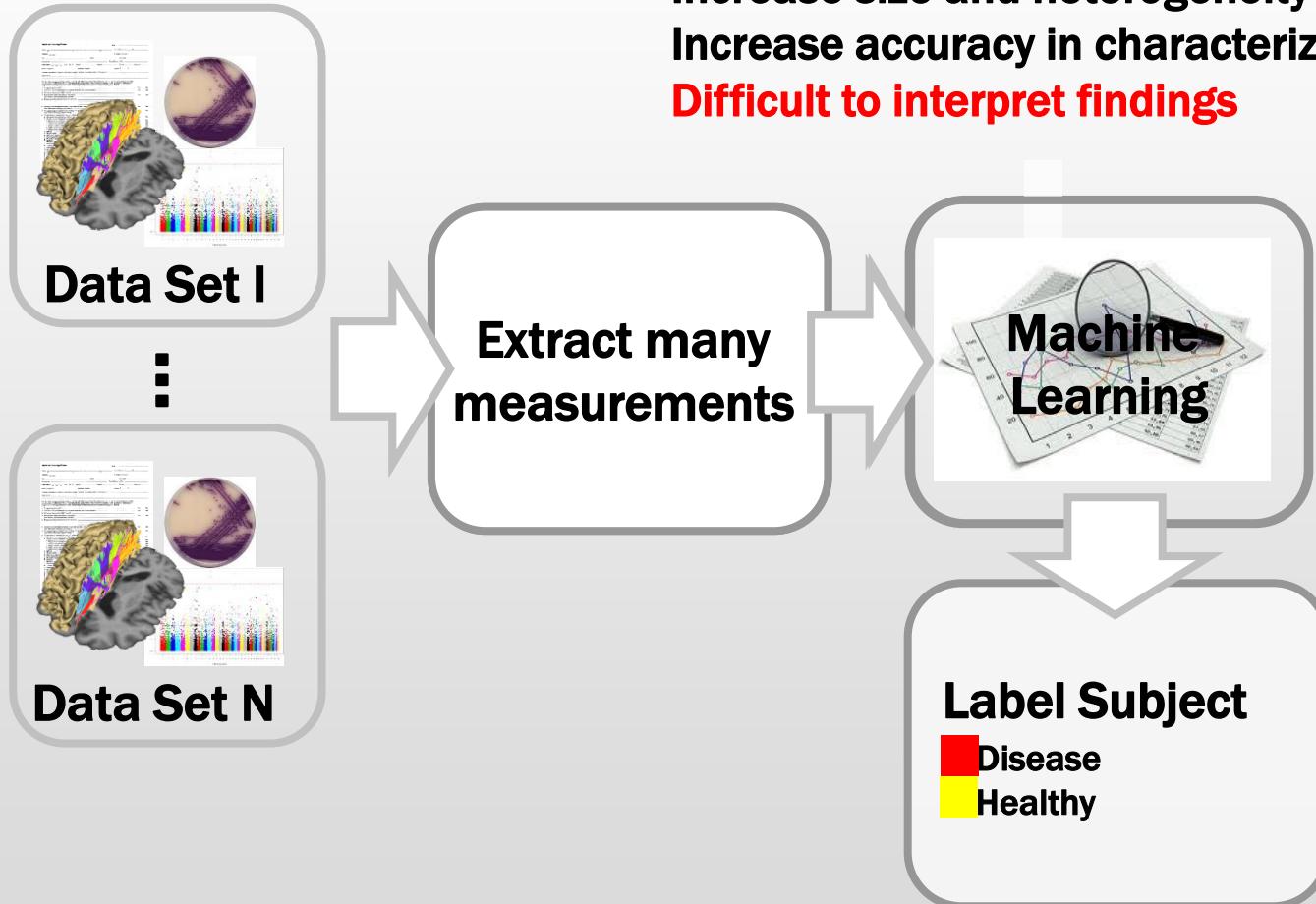
Slide courtesy of Kilian Pohl

Group Analysis: A Neuroscientific View

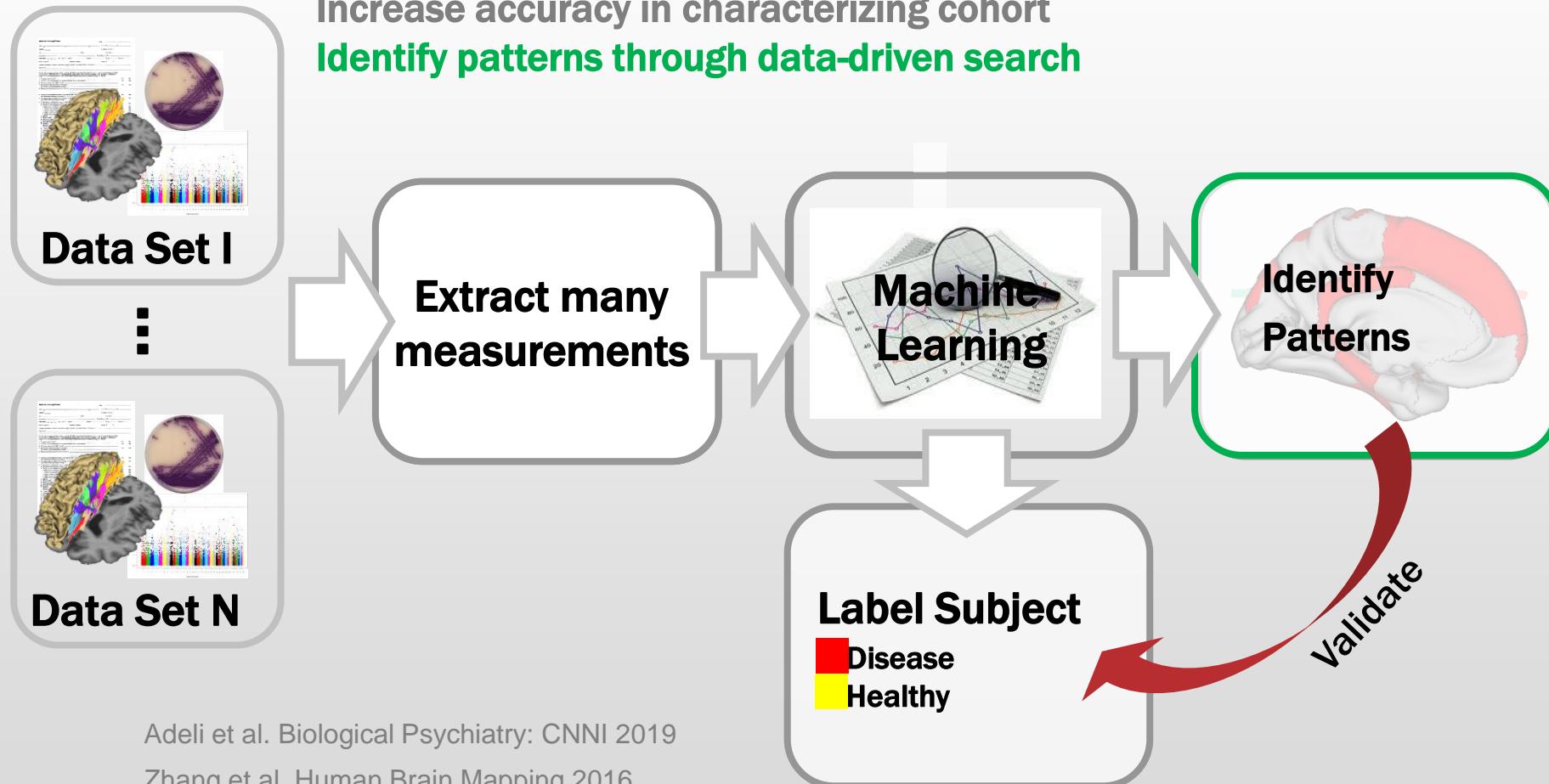
Zhao et al. Addiction Biology 2020



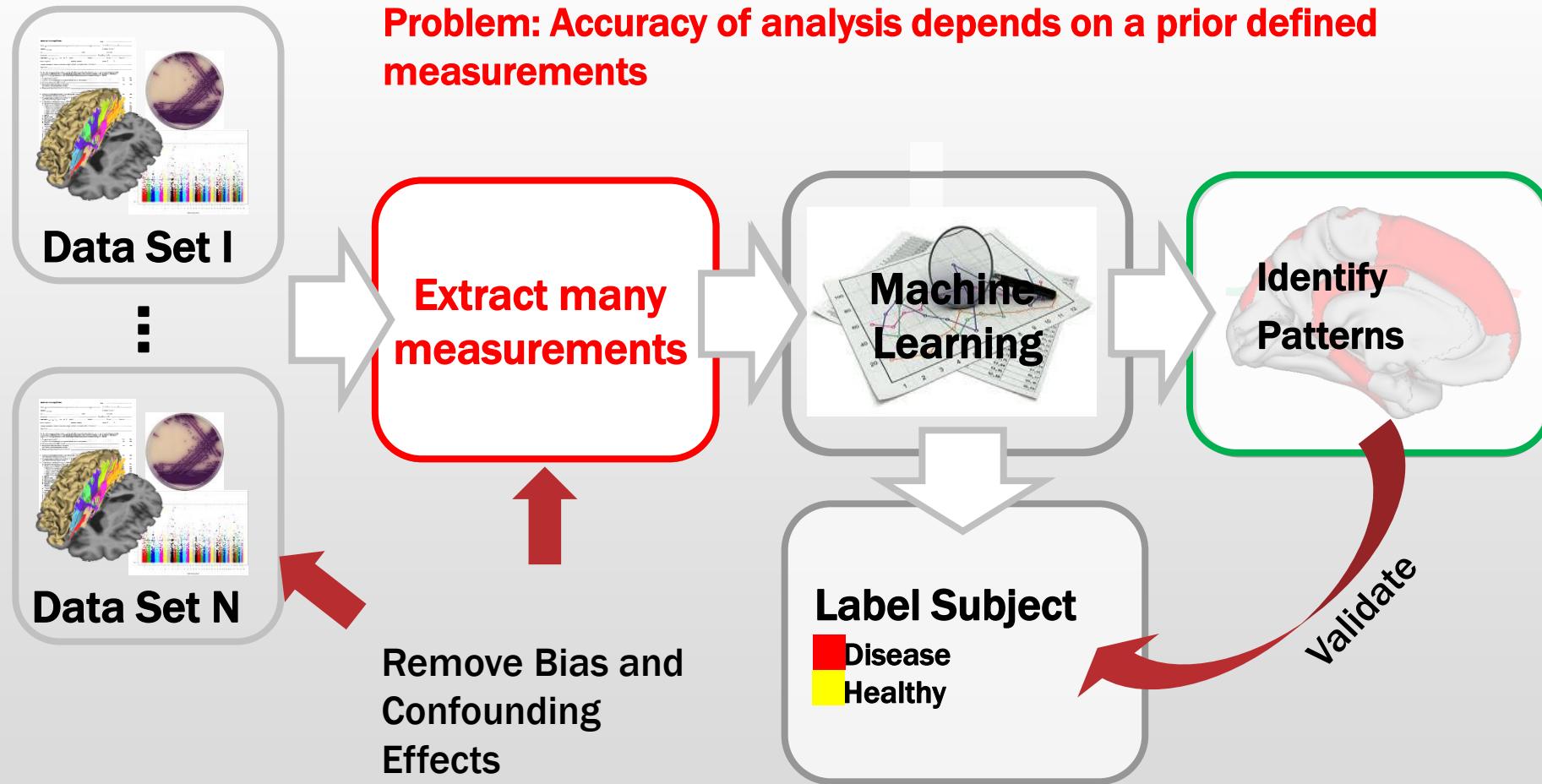
Identifying Disease Specific Markers



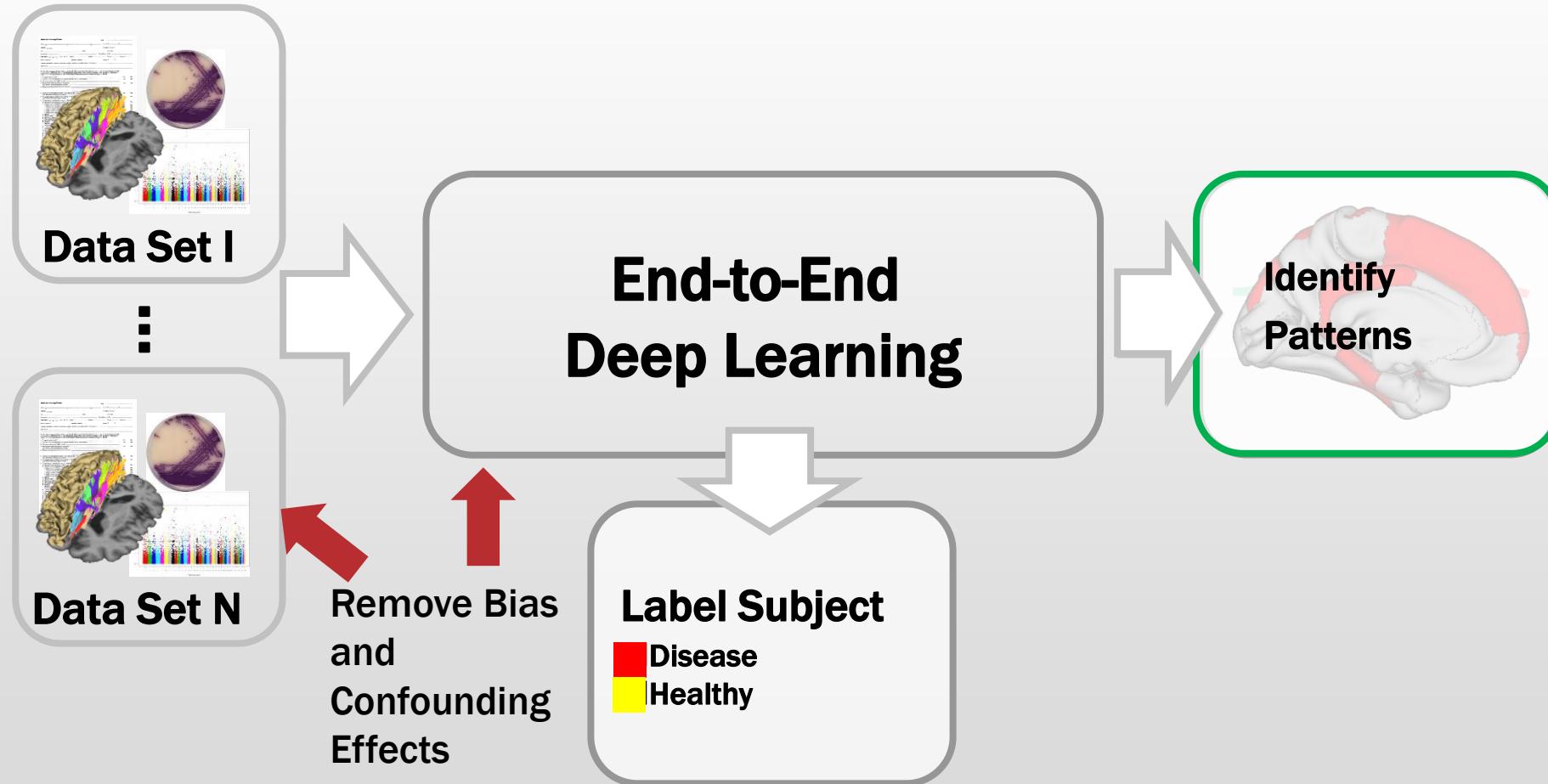
Identifying Disease Specific Markers



Identifying Disease Specific Markers



Identifying Disease Specific Markers





VIDEO

Rating the motor impairments severity



MDS-UPDRS

Gait Score:

0
1
2
3
4



*Movement Disorder Society Unified Parkinson's
Disease Rating Scale (MDS-UPDRS) [1]*

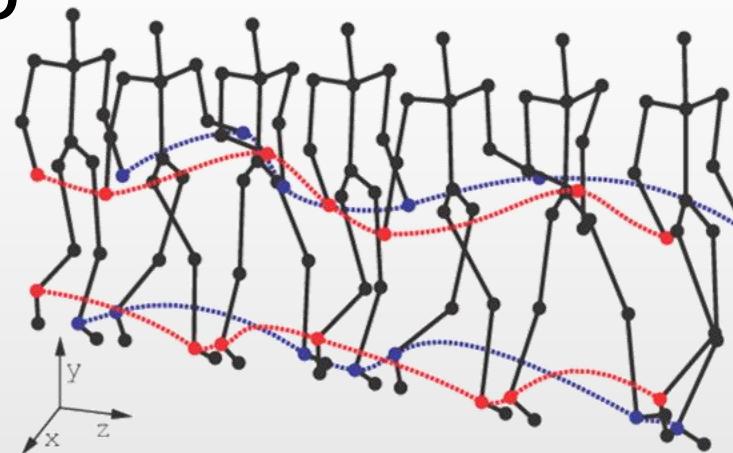
[1] Goetz et al. Movement disorders 2008



Rating the motor impairments severity



Sequence of 3D
skeletons
(pose)



Estimate



MDS-UPDRS
Gait Score

- (+) Low-dimensionality data
- (+) Efficient in required number of samples
- (+) Anonymized data

- (+) Entirely vision-based method
- (+) Non-intrusive video recordings
- (+) Scalable



Lu et al., MICCAI 2020



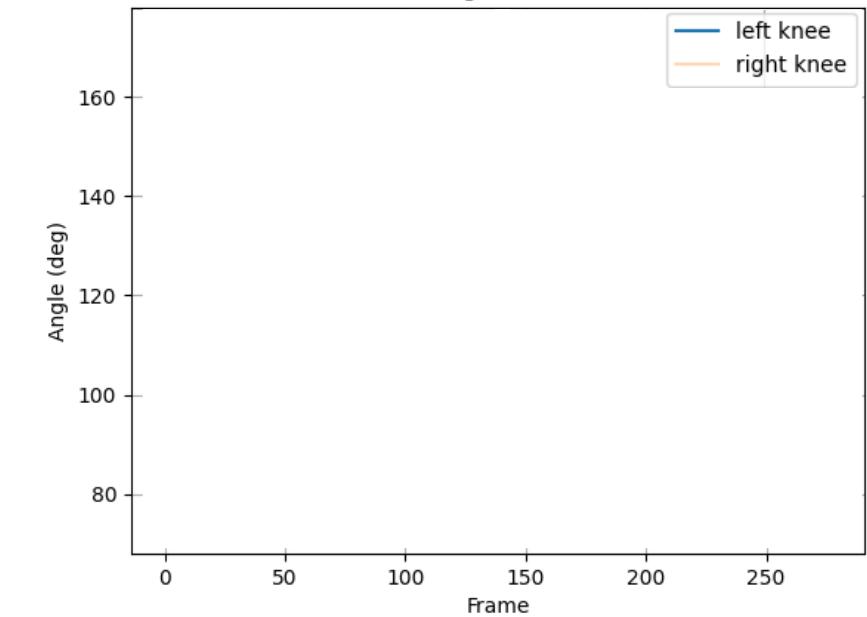
Original Video
Overlaid 3D Mesh/Pose



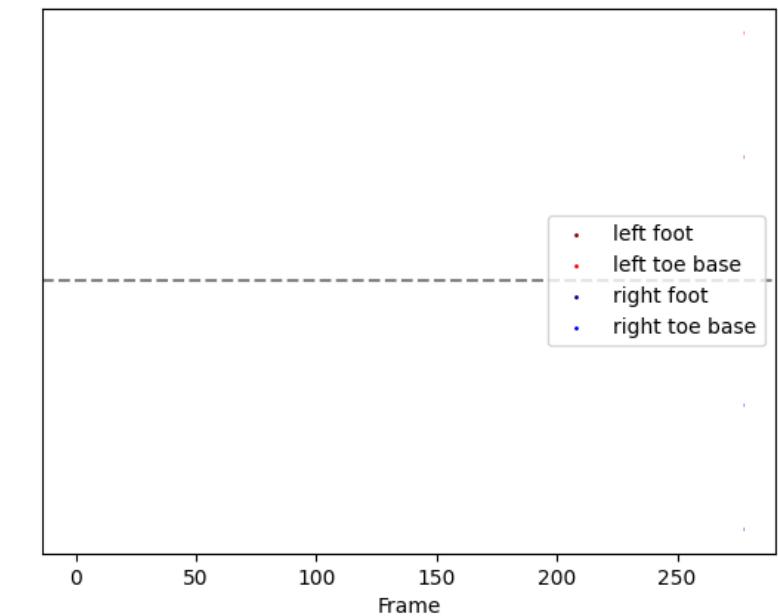
Side View



Knee Angle Over Time



Foot Contacts



Endowing healthcare spaces with ambient intelligence



Stanford Partnership in
AI-Assisted Care (PAC)



Hospital spaces



Daily living spaces

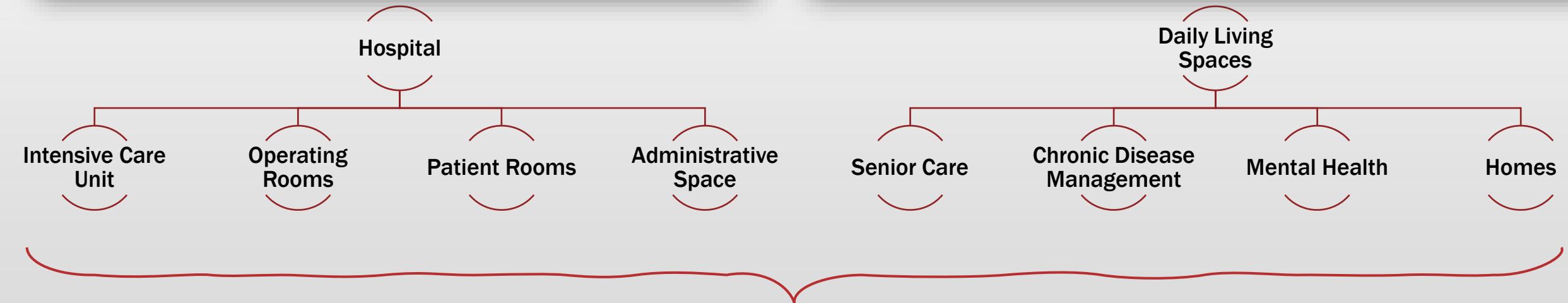
Haque et al., Nature 2020, Illuminating the Dark Spaces of Healthcare with Ambient Intelligence

Yeung et al., New England Journal of Medicine 2018, Bedside Computer Vision — Moving Artificial Intelligence from Driver Assistance to Patient Safety

Luo and Hsieh et al., Machine Learning for Healthcare 2018, Computer Vision-based Descriptive Analytics of Seniors' Daily Activities for Long-term Health Monitoring



Ehsan Adeli, Ph.D. (eadeli@stanford.edu)



Recognizing Activities in Videos from Contactless Sensors using Computer Vision Technology



Alan Luo

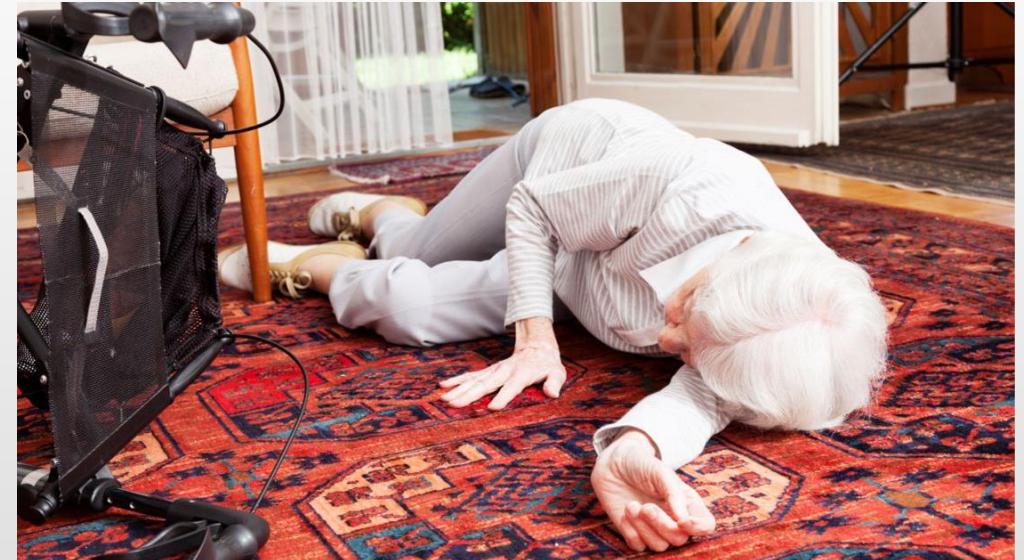


Ehsan Adeli, Ph.D. (eadeli@stanford.edu)

Intervene to improve care and reduce costs

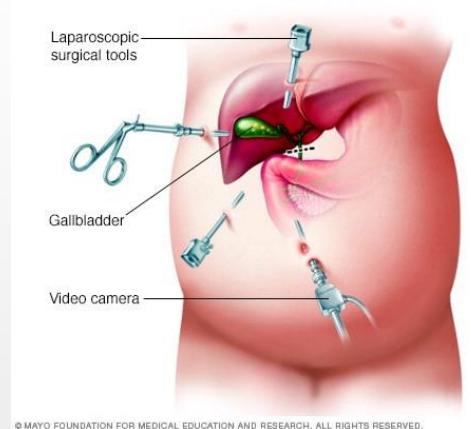
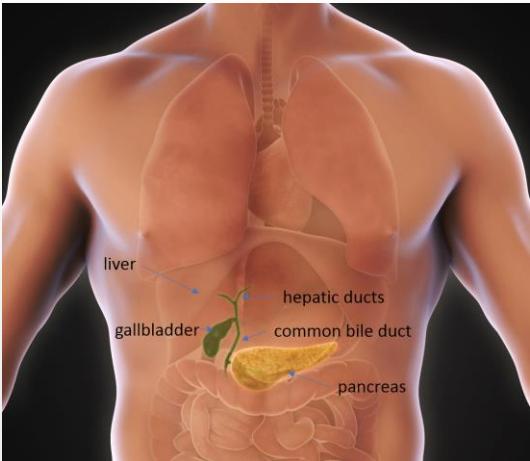


Early mobilization



Fall detection

Surgical Phase Recognition - Cholecystectomy



Surgical Workflow



Preparation

Clipping/Cutting

Gallbladder Packaging

Gallbladder Retraction

Calot Triangle Dissection

Gallbladder Dissection

Cleaning/Coagulation

Tobias, et al. "Opera: Attention-regularized transformers for surgical phase recognition." *MICCAI 2021*.

Slides Courtesy of
Tobias Czempiel, Magda Paschali

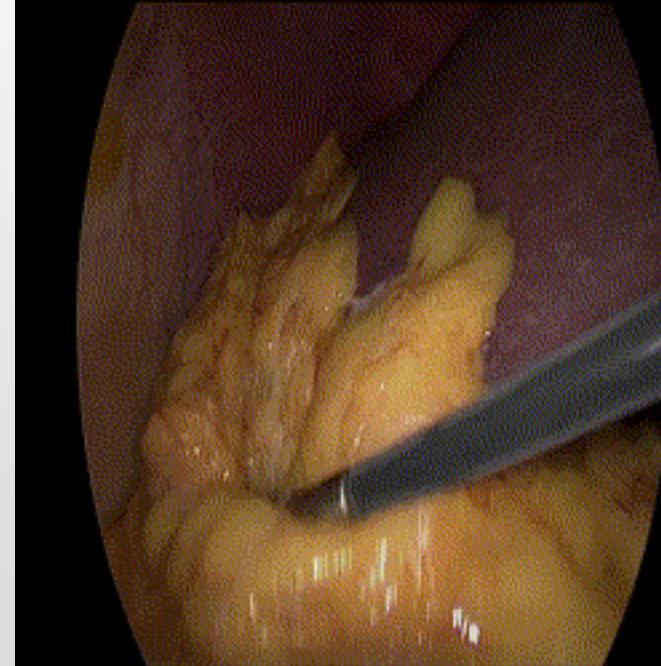
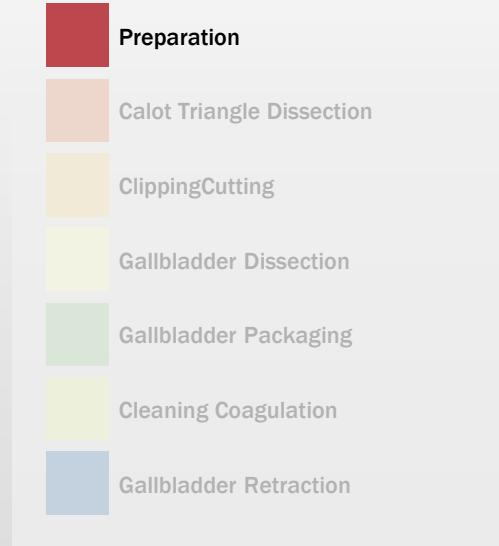


Surgical Phase Recognition - Cholecystectomy

Cholec80 dataset



Surgical Workflow

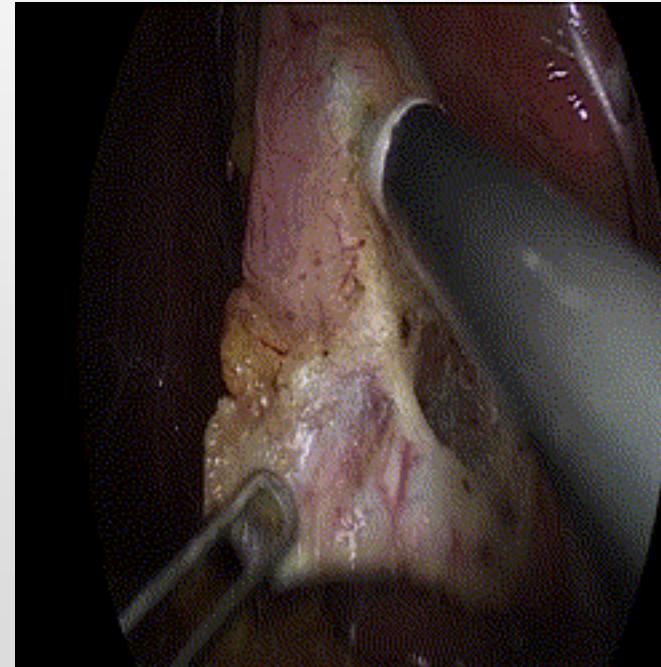
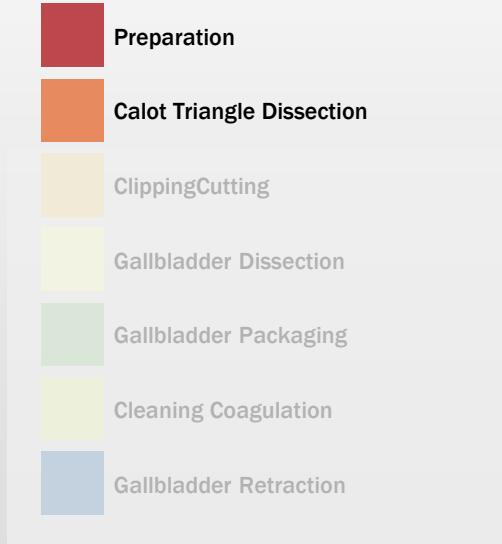


Surgical Phase Recognition - Cholecystectomy

Cholec80 dataset



Surgical Workflow



Surgical Phase Recognition - Cholecystectomy

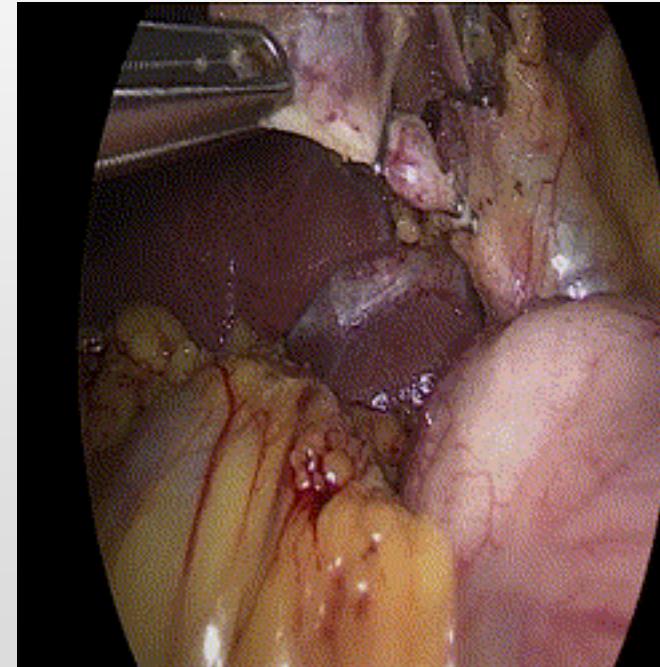
Cholec80 dataset



Surgical Workflow



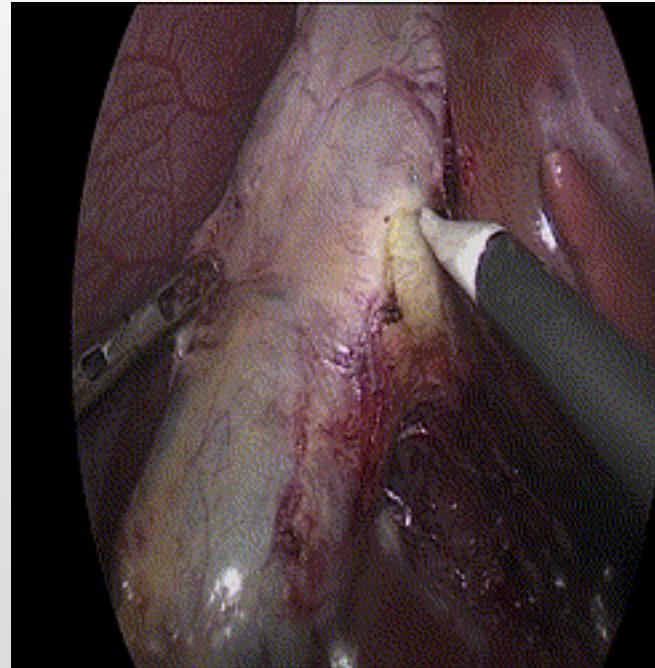
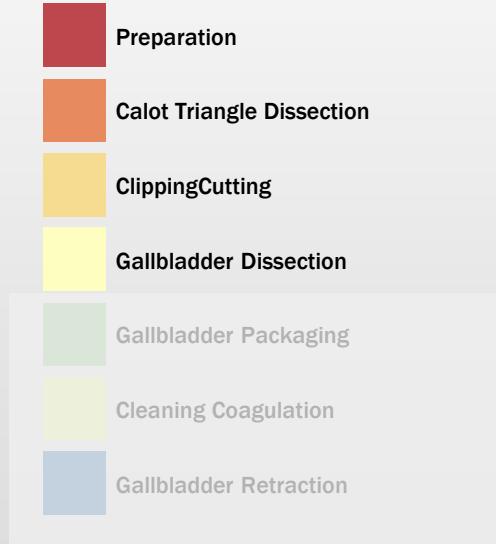
- Preparation
- Calot Triangle Dissection
- Clipping/Cutting
- Gallbladder Dissection
- Gallbladder Packaging
- Cleaning/Coagulation
- Gallbladder Retraction



Surgical Phase Recognition - Cholecystectomy

Cholec80 dataset

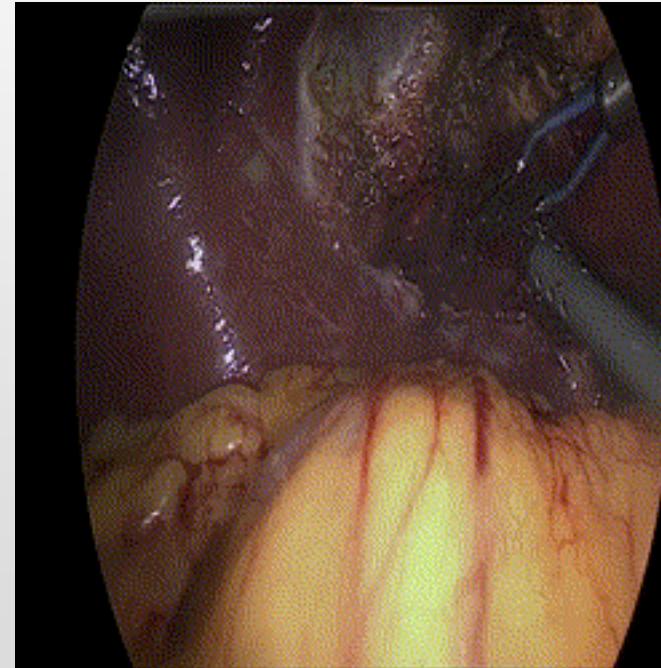
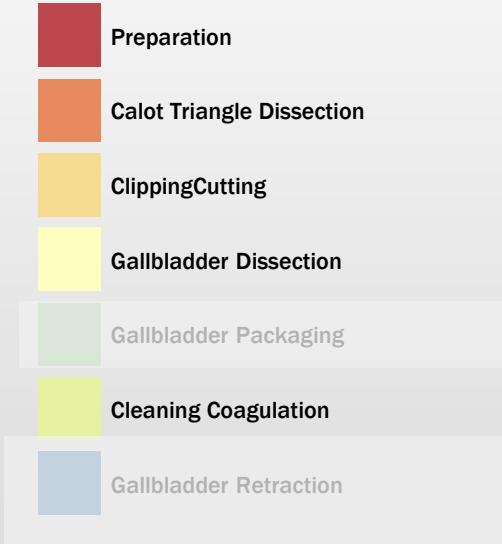
Surgical Workflow



Surgical Phase Recognition - Cholecystectomy

Cholec80 dataset

Surgical Workflow



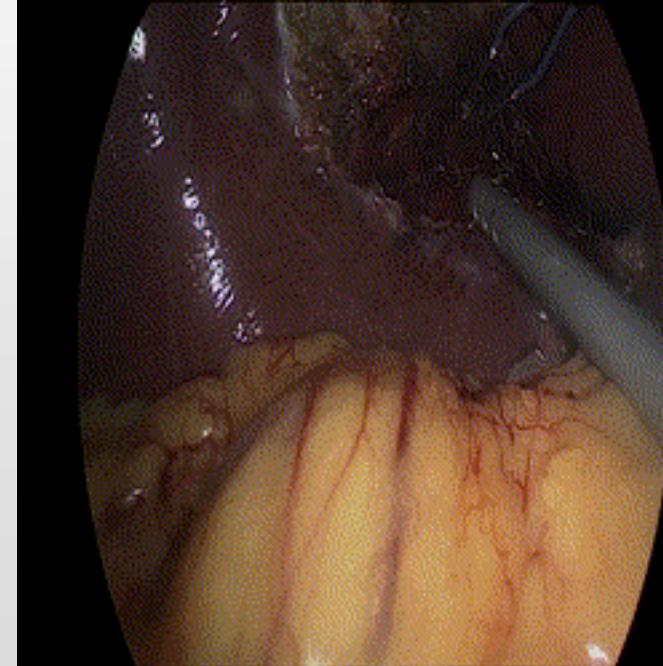
Surgical Phase Recognition - Cholecystectomy

Cholec80 dataset

Surgical Workflow



- Preparation
- Calot Triangle Dissection
- Clipping Cutting
- Gallbladder Dissection
- Gallbladder Packaging
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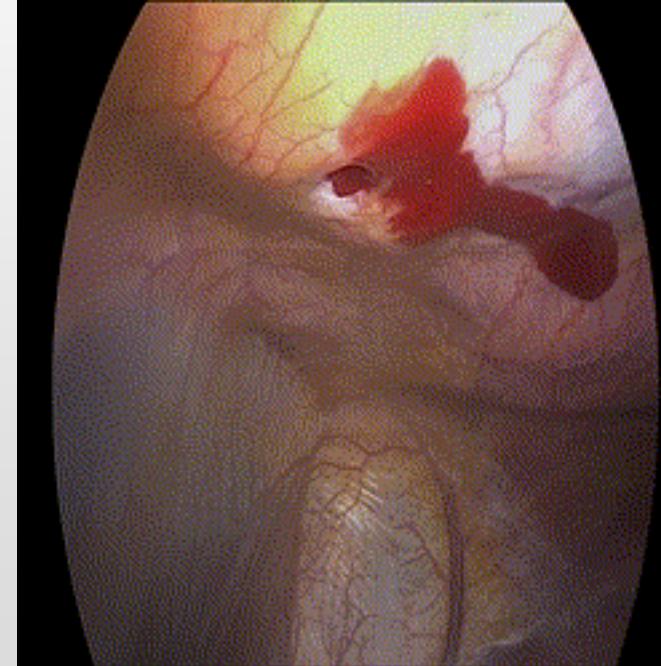
Surgical Phase Recognition - Cholecystectomy

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Surgical Workflow

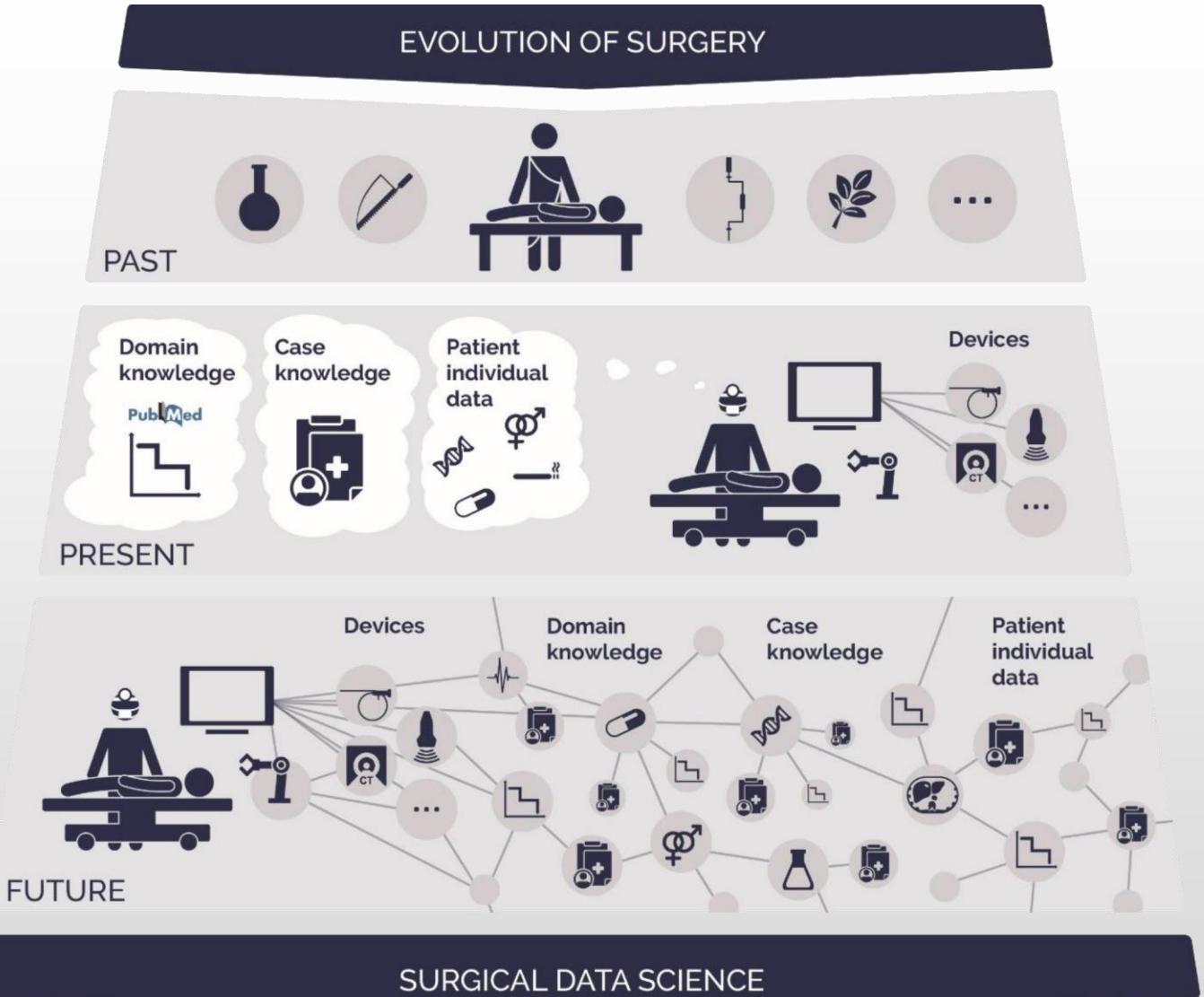


- Preparation
- Calot Triangle Dissection
- Clipping/Cutting
- Gallbladder Dissection
- Gallbladder Packaging
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- Gallbladder Retraction



Potentials & Challenges

- + Early warnings when deviating from surgical plan
- + Context aware systems
- + Automatic archiving, surgical protocol



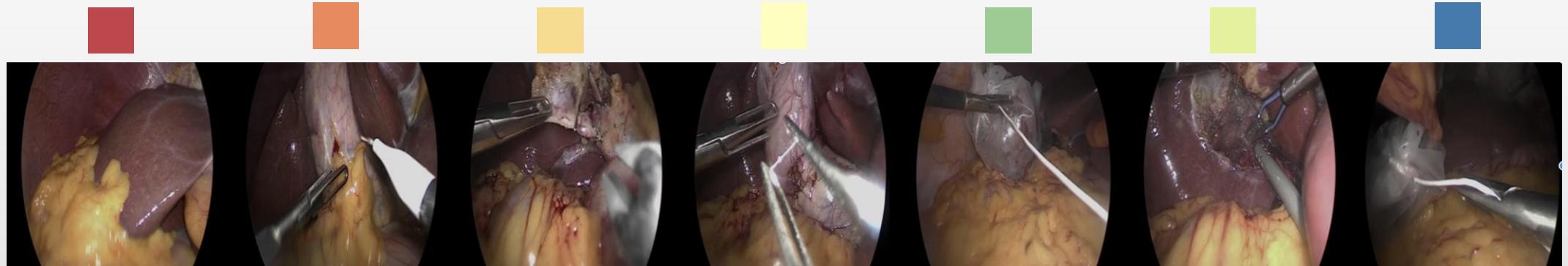
Maier-Hein, L., Vedula, S. S., Speidel, S., Navab, N., Kikinis, R., Park, A., Eisenmann, M., Feussner, H., Forestier, G., Giannarou, S., Hashizume, M., Katic, D., Kenngott, H., Kranzfelder, M., Malpani, A., März, K., Neumuth, T., Padoy, N., Pugh, C., ... Jannin, P. (2017). Surgical data science for next-generation interventions. *Nature Biomedical Engineering*, 1(9), 691–696. <https://doi.org/10.1038/s41551-017-0132-7>



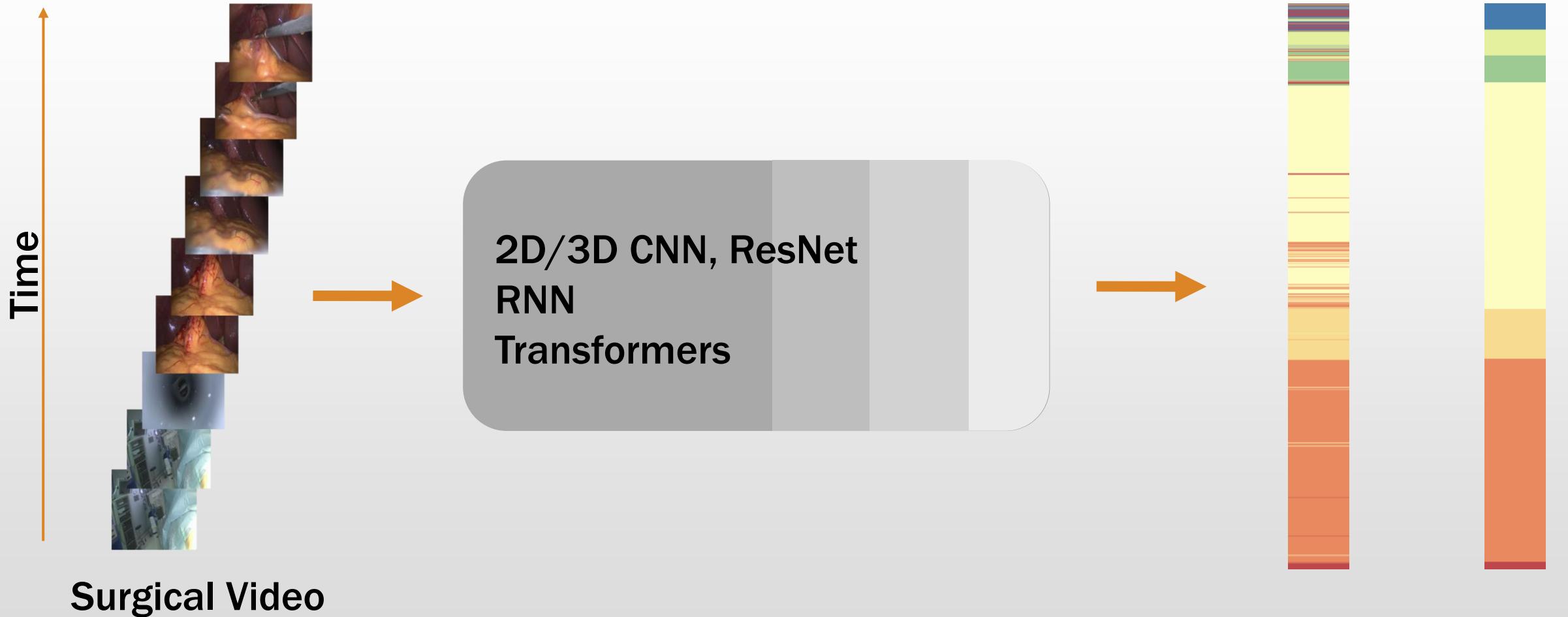
Potentials & Challenges

- + Early warnings when deviating from surgical plan
- + Context aware systems
- + Automatic archiving, surgical protocol

- Variability of patient anatomy and surgeon style
- Limited available training data
- Strong similarities among different phases



Methods





3D reconstruction of cystoscopy videos for comprehensive bladder records

KRISTEN L. LURIE,^{1,2} ROLAND ANGST,³ DIMITAR V. ZLATEV,²
JOSEPH C. LIAO,^{2,4} AND AUDREY K. ELLERBEE BOWDEN^{1,5}

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²*Dept. of Urology, Stanford University, Stanford, CA, USA*

³*Max Planck Institute, Saarbrücken, Germany*

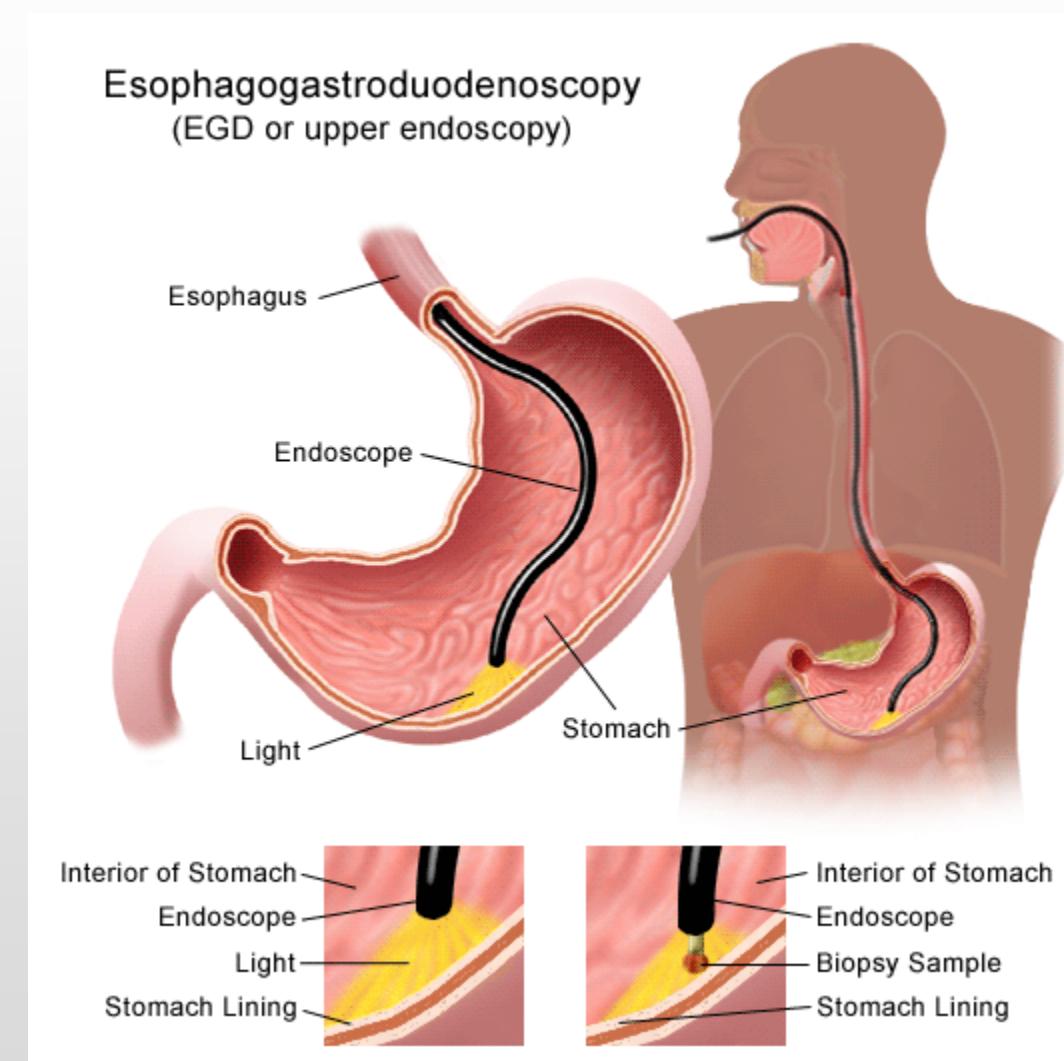
⁴*Corresponding author: jliao@stanford.edu*

⁵*Corresponding author: audrey@ee.stanford.edu*



Endoscopy Videos

- Endoscopy and its organ-specific derivatives (e.g., laparoscopy, colonoscopy, cystoscopy) play a powerful role in **diagnostic imaging, surgical guidance, and cancer surveillance**.
- Rich information in endoscopy videos
- Cumbersome nature of post-session video review
- Results in condensing lengthy video data into a few still images and brief notes or drawings about the locations and appearance of suspicious lesions and scars.





Motivation

- The availability of complete organ reconstructions for other medical imaging modalities has been of powerful effect, leading to clinical advances in areas such as intravenous injection, sleep apnea evaluation, and cervical and gastric cancer
- a comprehensive representation of the endoscopy data that enables straightforward and rapid review of a single endoscopy session or comparisons across several could better support the clinical decision-making process and enable new directions for cancer research.



Framework

- Develop comprehensive representations of cystoscopies, an important and clinically significant application,
- bladder cancer has the highest recurrence rate of all cancers and demands at least annual surveillance through cystoscopy to monitor recurrence (most expensive cancer to treat in patient's lifetime).
- Three-dimensional reconstructions that can capture both the 3D organ shape and appearance can enable depiction of full organs and localization of individual regions to anatomical locations in the organ.



Challenges

1. The shape of the bladder is not known *a priori*; further, the bladder may change shape depending on its level of distension between imaging sessions.
2. While the shape of the bladder is important for orienting the physician, the surface appearance will be more carefully scrutinized by the physician than the exact shape.
3. The endoscopic light source is part of the endoscope, which means the surface illumination is constantly changing.
4. Finally, a cystoscopic video has a duration of several minutes and covers a large surface area relative to the area viewed in a single frame. As a result, a feature does not remain in the field of view for a large percentage of the video, which is problematic given that the key to reconstruction is efficiently detecting when a feature re-enters the field of view in order to reconstruct an accurate model.



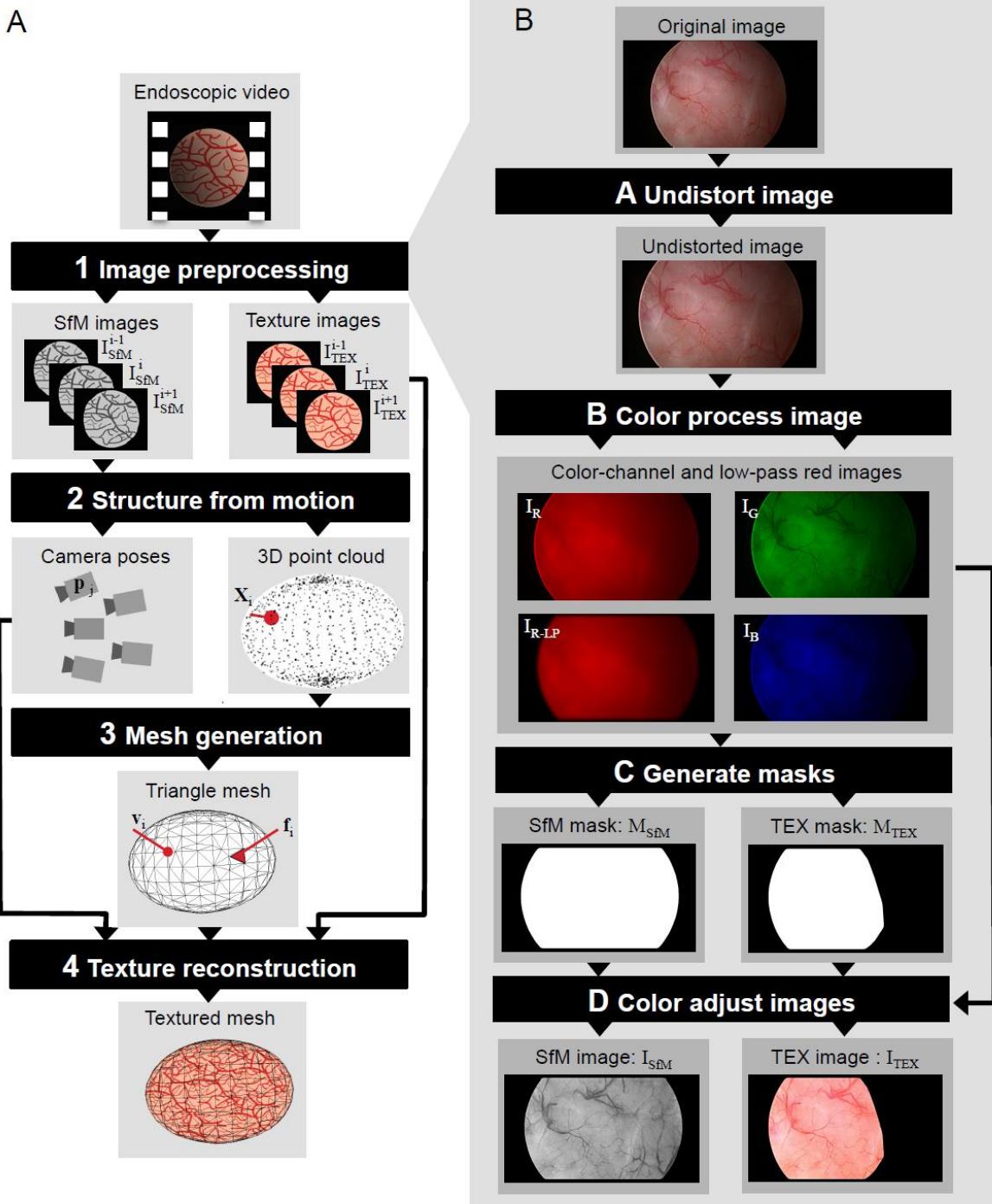
This paper

- First method for dense 3D reconstruction of the bladder from white light cystoscopy (WLC) videos that uses standard clinical hardware and introduces only a minor modification to the standard clinical scan pattern.
- The four challenges are resolved by
 - (1) not utilizing any ground truth data to reconstruct the bladder surface,
 - (2) presenting a complete pipeline to convert the cystoscopic images into a 3D textured model of the bladder,
 - (3) developing an image preprocessing technique to remove lighting artifacts and thus strengthen feature matching, and
 - (4) utilizing a structure-from-motion algorithm that efficiently can detect loops.



Overview of the four-step 3D reconstruction algorithm

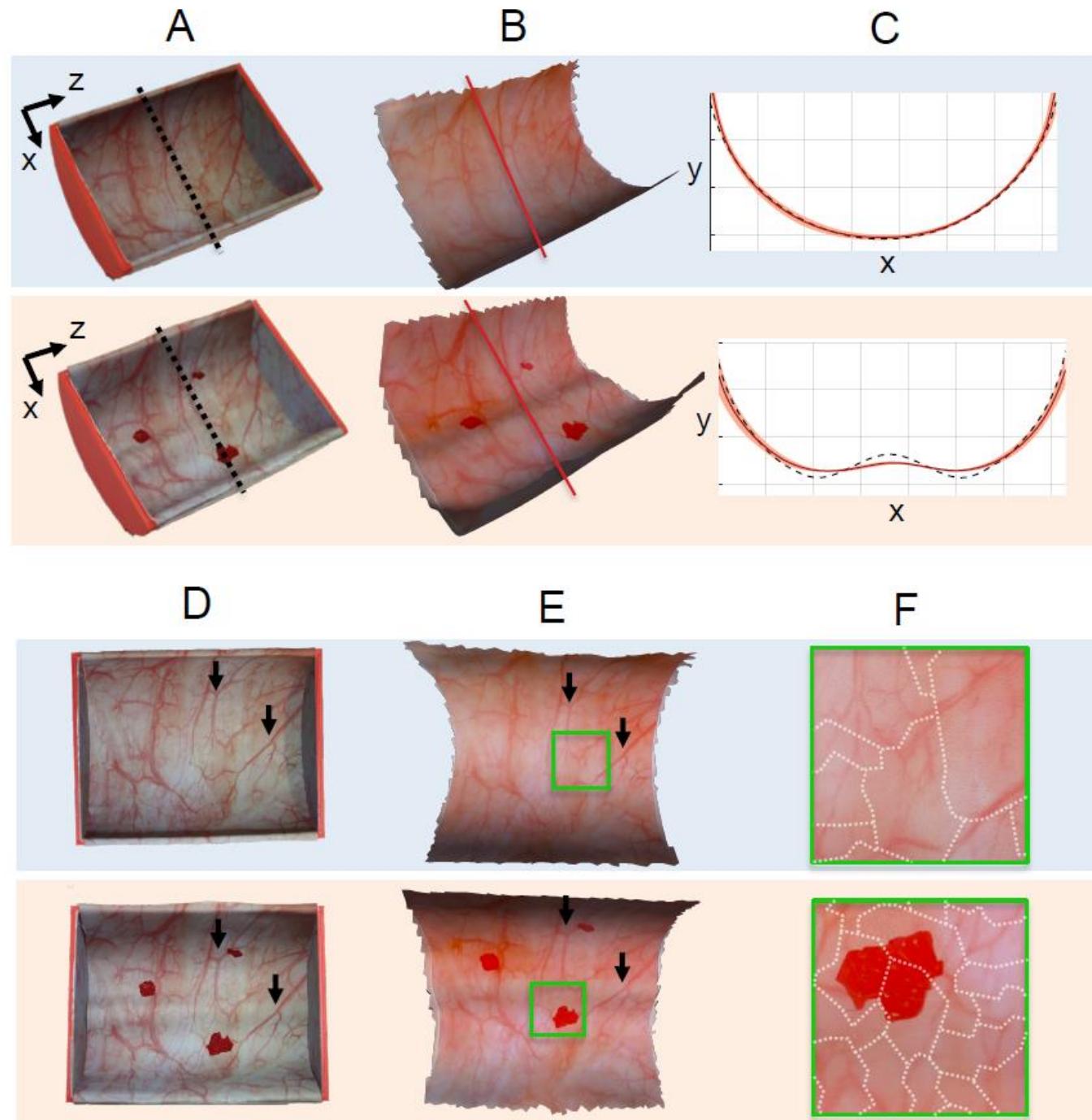
- SfM: Structure from motion
- TEX: texture reconstruction



Reconstruction of tissue-mimicking phantom datasets

- Top row (blue background): original semi-cylinder phantom; bottom row (orange background): modified semi-cylinder phantom.

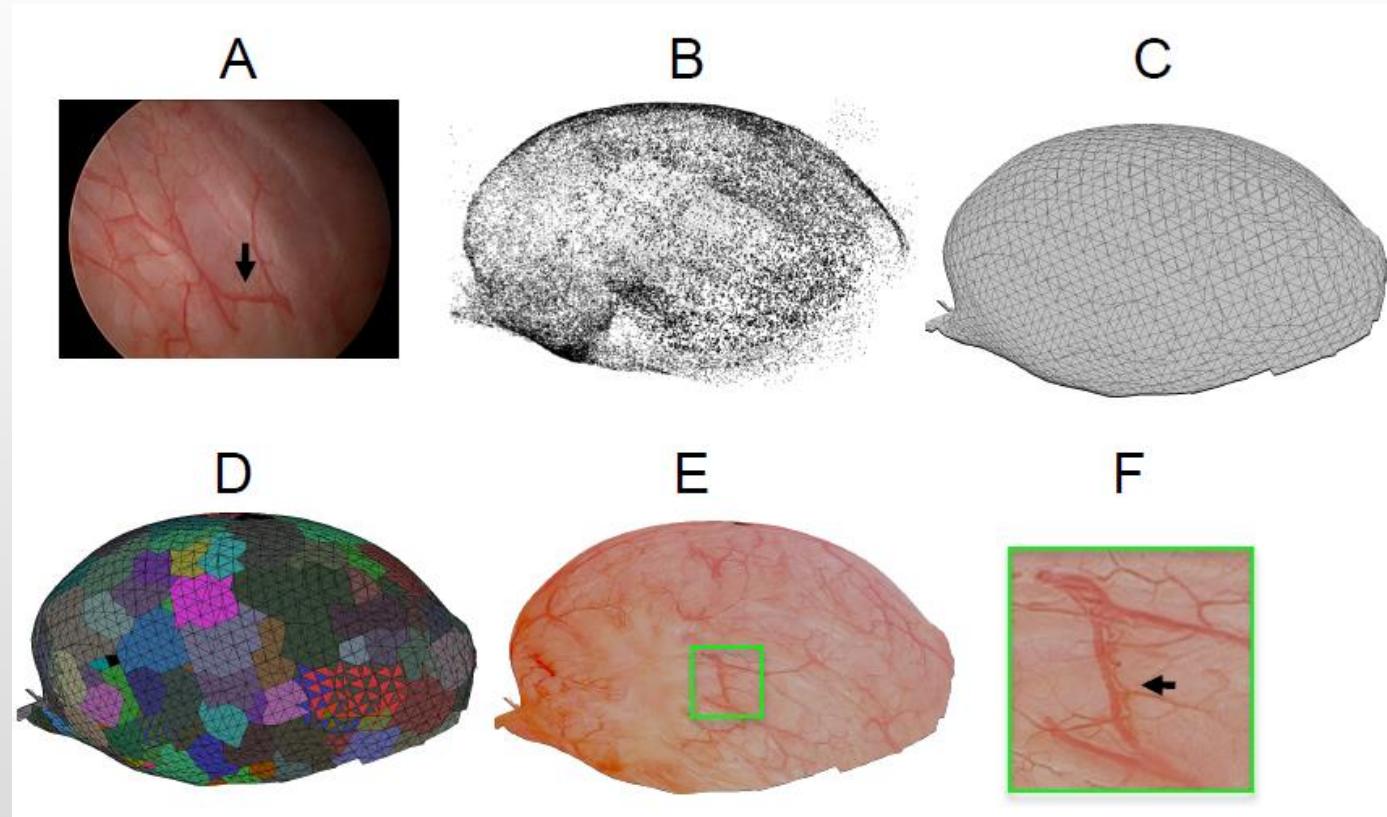
- Standard digital camera images of the phantoms highlighting their shaped are compared with the reconstructed mesh.
- Cross-sections of the expected mesh (dotted black line) and average reconstructed cross section (red) are compared.
- Standard digital camera images of the phantoms highlighting their surface appearance
- the reconstructed textured phantoms viewed from approximately the same camera angles. Black arrows are added to highlight similar features between the original and reconstructed images. Green boxes indicate regions of texture shown in greater detail in F
- emphasize the seamlessness between regions composed of different images. The dotted white lines in (F) indicate boundaries between mesh faces that are composed of different original images.



Reconstruction of clinical datasets

- A. a representative, original WLC image,
- B. point cloud from the structure-from-motion step before outlier removal,
- C. mesh from the mesh-generation step,
- D. labeled texture (faces with the same color are labeled with the same input image)
- E. textured mesh from texture-generation steps. The green box shows a similar region between subfigures indicating clear continuity of vessels despite the use of multiple input images to construct this region. The green box is approximately the size of a single WLC image.

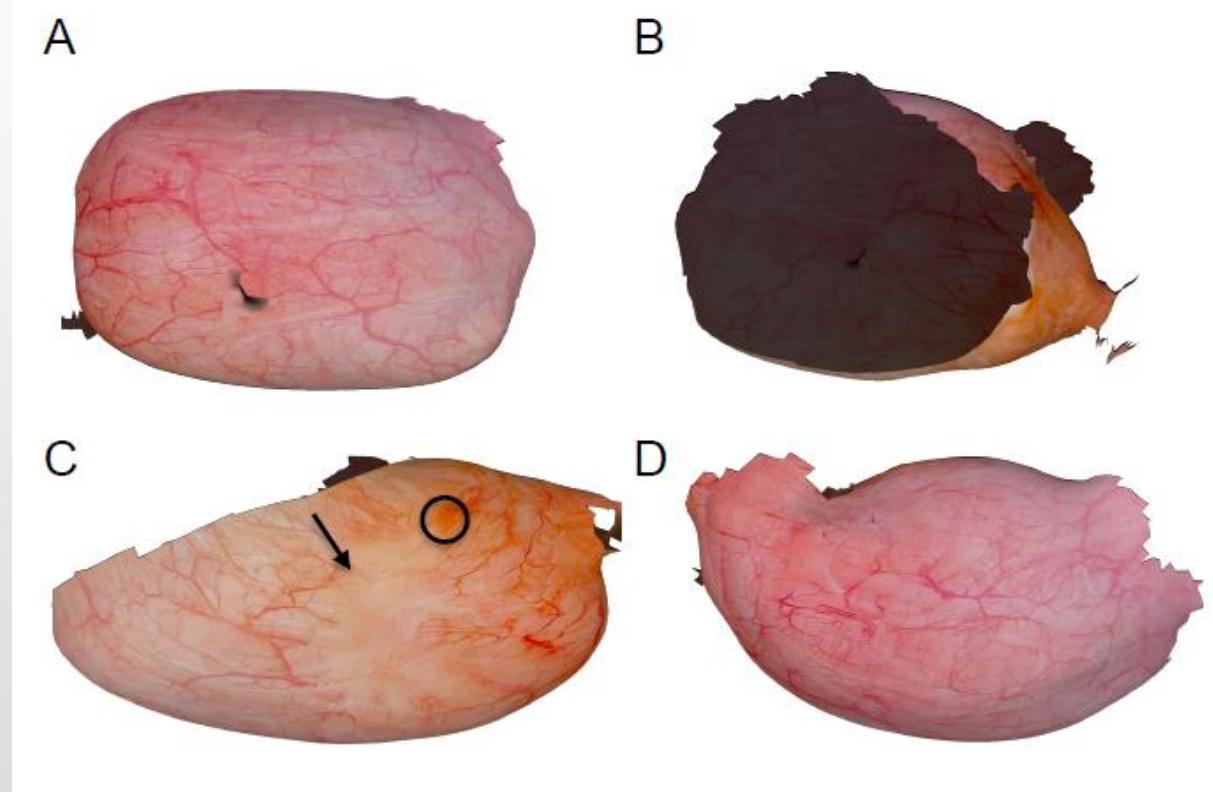
Black arrows in (A) and (F) indicate similar regions of the bladder.



Reconstruction of clinical datasets

- A. anterior,
- B. posterior,
- C. left lateral,
- D. right lateral walls

Black circle and arrow in (C) show regions of a papillary tumor and scarring, respectively. Regions that appear dark represent the interior of the bladder.



Conclusion

- The proposed algorithm can serve as the foundation for surgical planning, quality assessment of the procedure, optical annotation, and integration with other optical technologies.
- A longitudinal record of the bladder appearance can enable new quantitative studies of the time-varying appearance in the bladder wall (for example, to predict the location of early tumors or to stratify patient outcomes).
- The reconstructions presented in this work are based on rigid cystoscopies, but the proposed method is extendable to flexible cystoscopes.
- Importantly, the shape-agnostic nature of the algorithm may make it extendable to reconstructions of other organs using their respective endoscopy derivatives.



Thank You!

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Questions?

