

# Computing Methods for Experimental Physics and Data Analysis

Data Analysis in Medical Physics

Lecture 1: Intro to Medical Image Processing and Analysis

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INFN - Pisa

# Course objectives

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- to provide students with an overview of the computational and mathematical methods used in **medical image processing**
- to complement the programming skills you acquired in the first two modules of this course with the knowledge of the **Matlab** programming language

## In demos and hands-on lectures:

- practical examples of several methods currently used to enhance and extract relevant information from medical images
- a variety of data sets (2D, 3D, 4D), acquired with the main medical imaging modalities (e.g. RX, CT, MRI) will be made available

# Course Calendar

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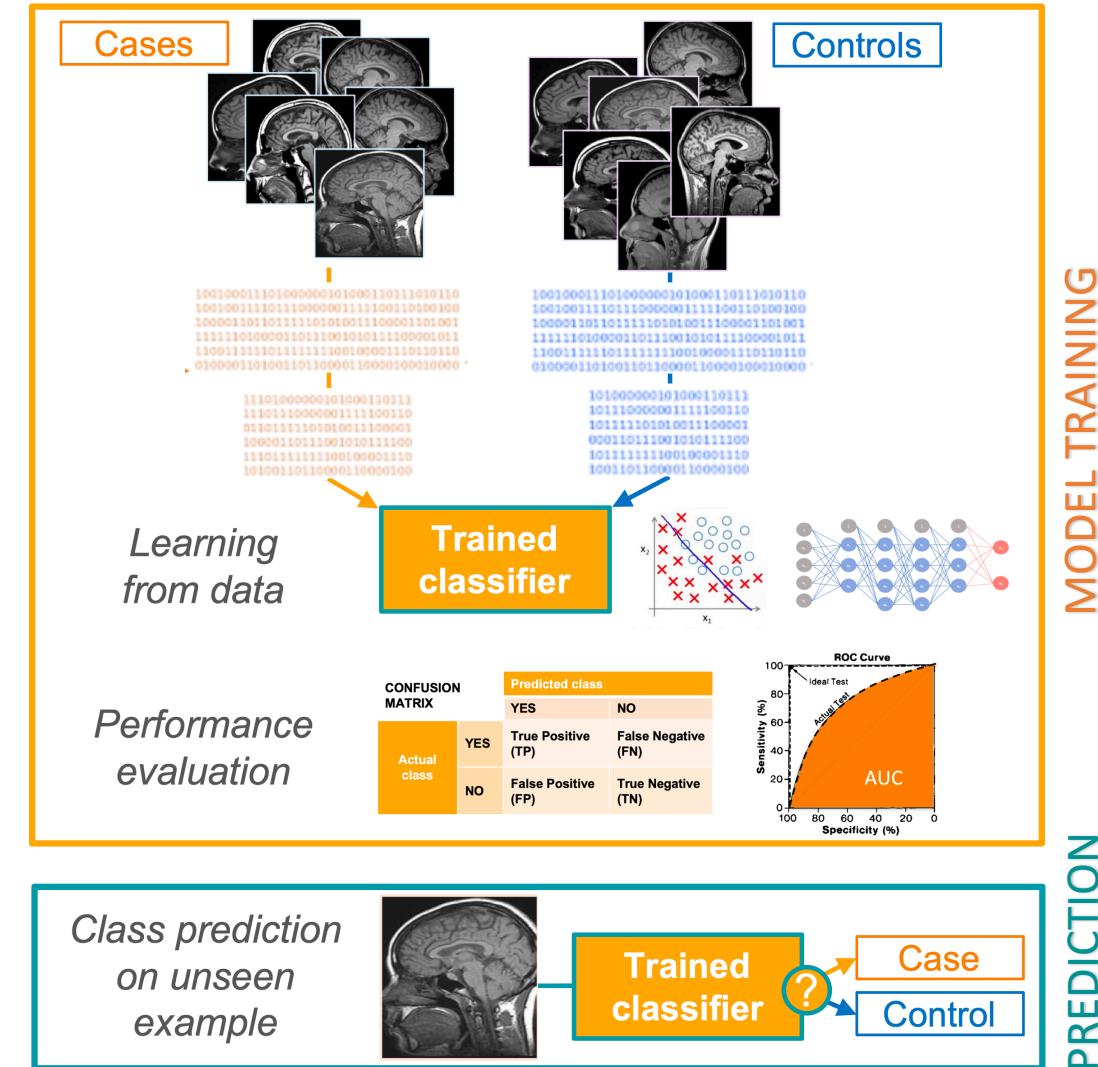
- **1. Thursday, November 19 9->11**
- **2. Monday, November 23 9->11**
- **3. Monday, November 23 16->18**
- **4. Thursday, November 26 9->11**
- **5. Monday, November 30 9->11**
- **6. Monday, November 30 16->18.00 – MedPhys + HEP**
- **7. Thursday, December 3 9->11**
- **8. Monday, December 7 9->11**
- **9. Monday, December 7 16->18**
- **10. Thursday, December 10 9->11**
- **11. Monday, December 14 9->11**
- **12. Monday, December 14 16->18**
- **13. Monday, December 17 9->11**

Bring your laptop!

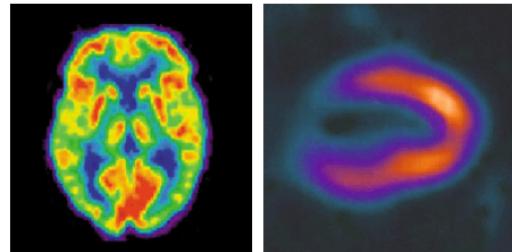
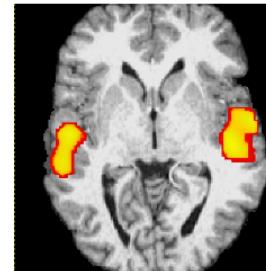
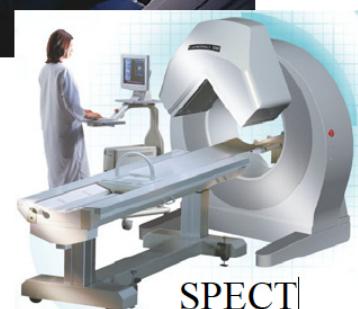
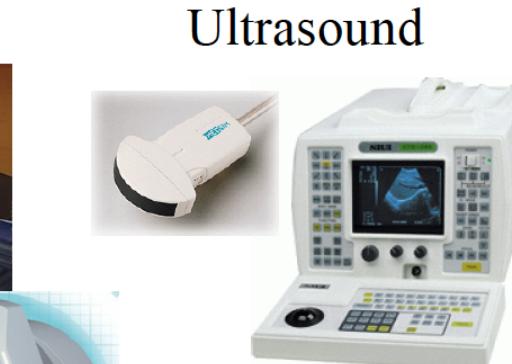
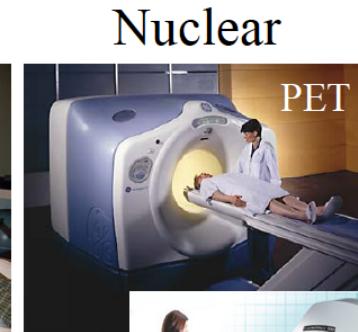
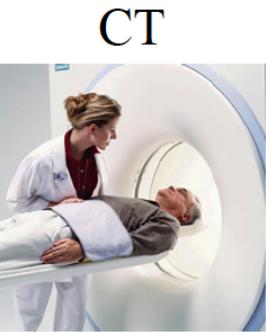
# Course overview

Medical data processing, feature extraction, feature/image classification:

- Handling standard-format medical data (DICOM), data anonymization, visualization
- Deriving features from images, image segmentation
- Data quality control pipelines, outlier removal, dimensionality reduction
- Data analysis and classification
- Performance evaluation: figures of merit, cross-validation, permutation test
- Machine-learning and deep-learning tools for segmentation and classification
- Data augmentation, transfer learning, retrieving localization information



# Medical Imaging: focus on diagnostic modalities



X-ray

magnetic spin

metabolic tracer X-ray emission

sound waves

*Medical images are more than pictures!!!*



<http://www.humanconnectomeproject.org/>

# Medical vs. non-medical images

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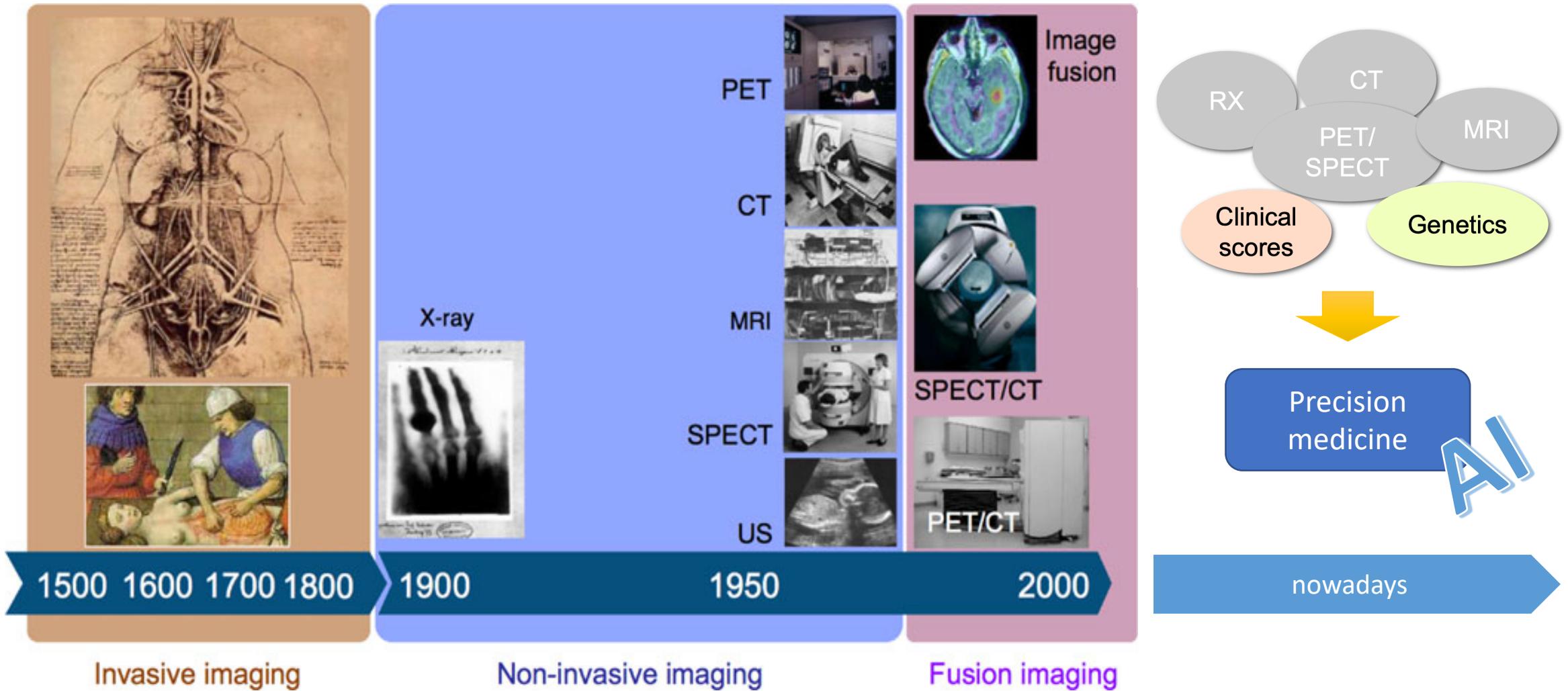
- Medical images reflect various physical properties of the body
  - Images are formed by interaction of radiation/ultrasounds with tissues/organs
- Medical images are not limited to 2D, as natural pictures generally are
- Medical images may allow to study:
  - the structure of tissues/organs
  - the function tissues/organs

# The main diagnostic imaging modalities

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TECHNIQUE	MEASURED PHYSICAL QUANTITY	INDIRECT CLINICAL INFORMATION
X-Ray (Radiography, Computed tomography)	Intensity of transmitted X rays	Mapping tissue radiodensity. Mapping body cavity (e.g. vessels) through contrast agents
Nuclear Medicine (SPECT, PET)	Intensity of emitted $\gamma$ rays	Distribution functional radiotracers
Ultrasound (US, Doppler)	Intensity / frequency of reflected ultrasound	Mapping tissue with different acoustic properties. Mapping blood velocity
Nuclear Magnetic Resonance (MRI, fMRI, ...)	Intensity of stimulated emission of radiofrequency waves	Mapping tissues with different relaxation properties, different water content and different water aggregation status
...	...	...

# Medical Imaging: historical perspective



# The role of computers in Medical Imaging

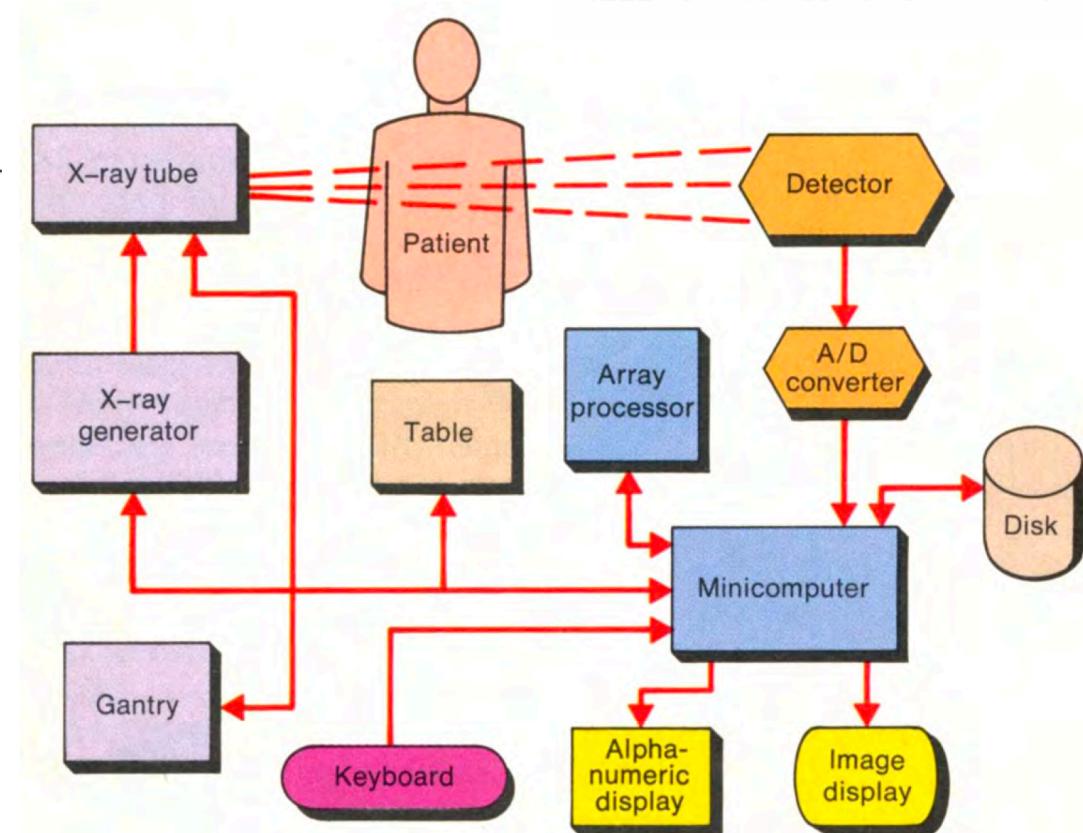
## ***The computer revolution in medical imaging systems***

Louis J. Heitlinger

**Computers have become an essential part of medical imaging, where their many uses include control of hardware and image enhancement and generation**

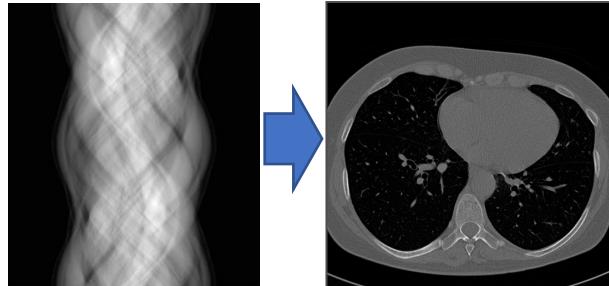
In a computerized tomography system, the X-ray data are acquired by a detector and converted to digital data. These data are then processed by the array processor and the minicomputer to generate the image, which is then displayed on a CRT.

IEEE POTENTIALS • DECEMBER 1984

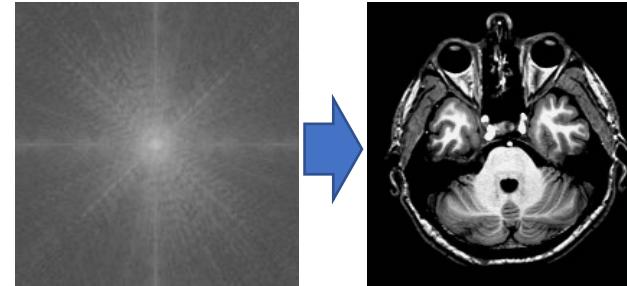


# The role of computers in Medical Imaging

- To reconstruct acquired signals to form diagnostic images

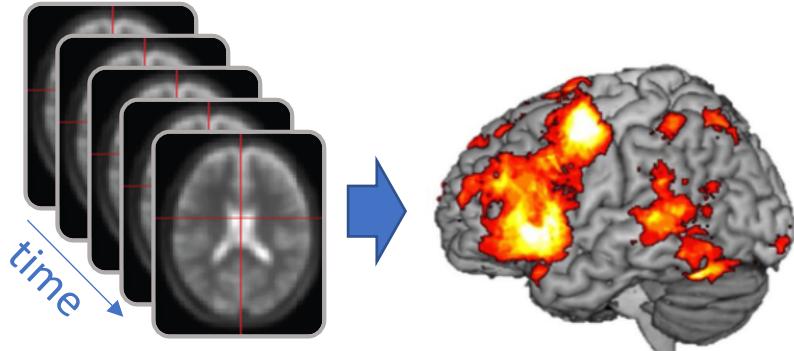


*CT image reconstruction from sinogram*

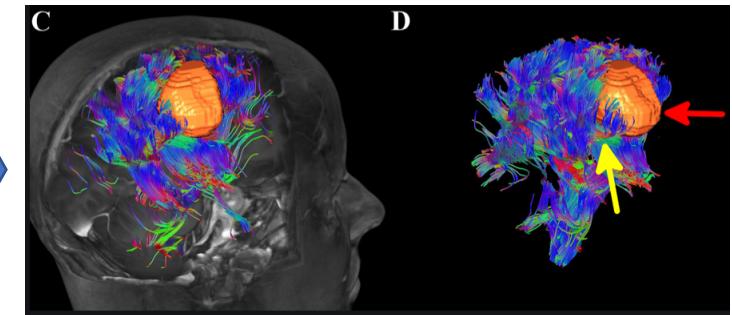


*MRI image reconstruction from k-space data*

- To allow the visualization of the acquired information



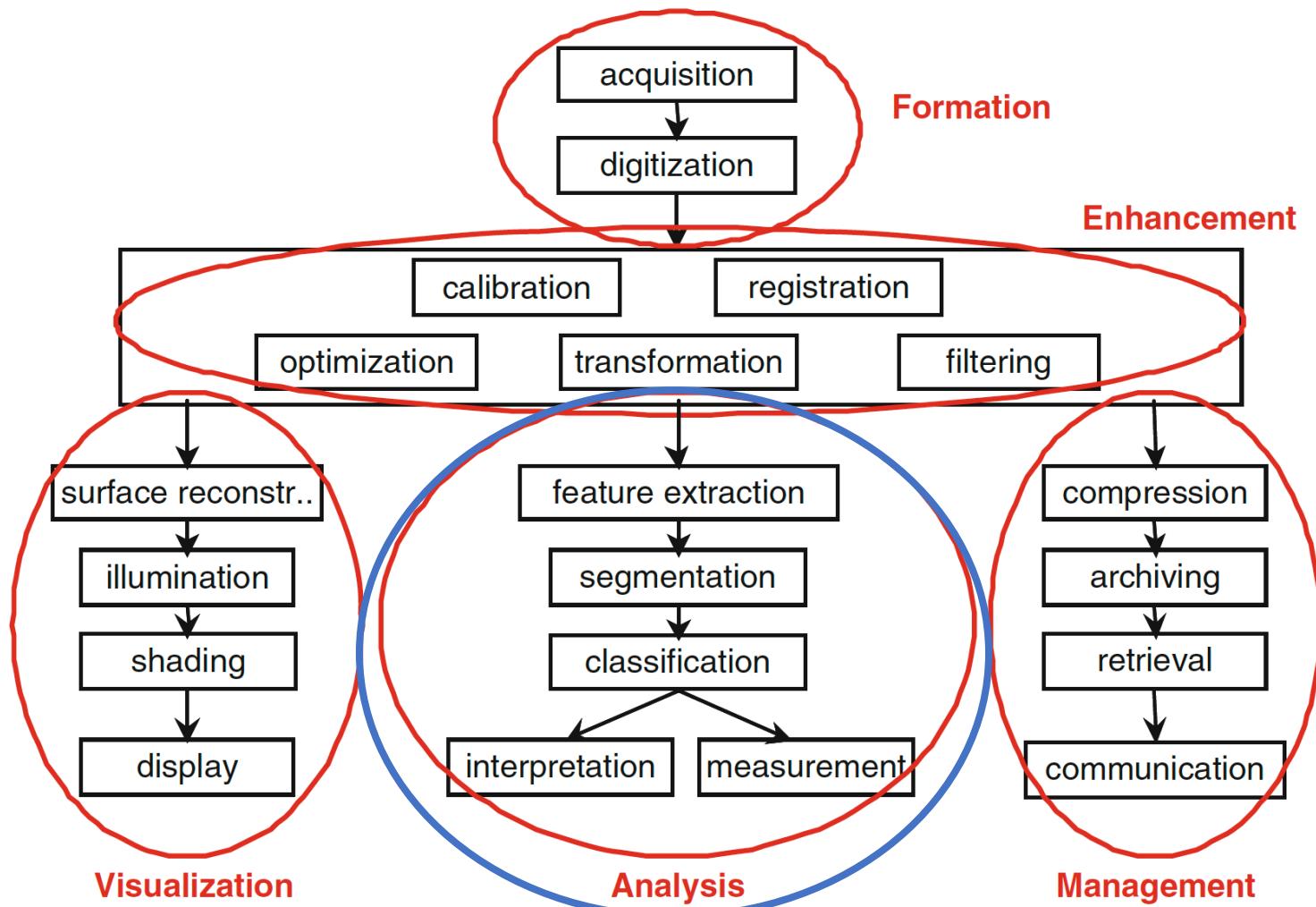
*Activated brain areas while executing a task*



*Axon fiber tracts from diffusion-weighted MRI data*

# The role of computers in Medical Imaging

- Algorithms for information processing enter at many levels in the image formation pipeline
- We will focus in this course on the high-level **Analysis** procedures, and we will not cover the countless other possibilities



# Biomedical image processing and analysis

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The main objectives are:

- To detect abnormalities in diagnostic images (lesions, etc.)
- To follow up pathological conditions (e.g. measuring the grow rate of lesions)
- To assess treatment efficacy



*The aim is to assist clinicians in their tasks,  
not to replace them:*

→ Computer Aided Detection/ Diagnosis (CAD) systems

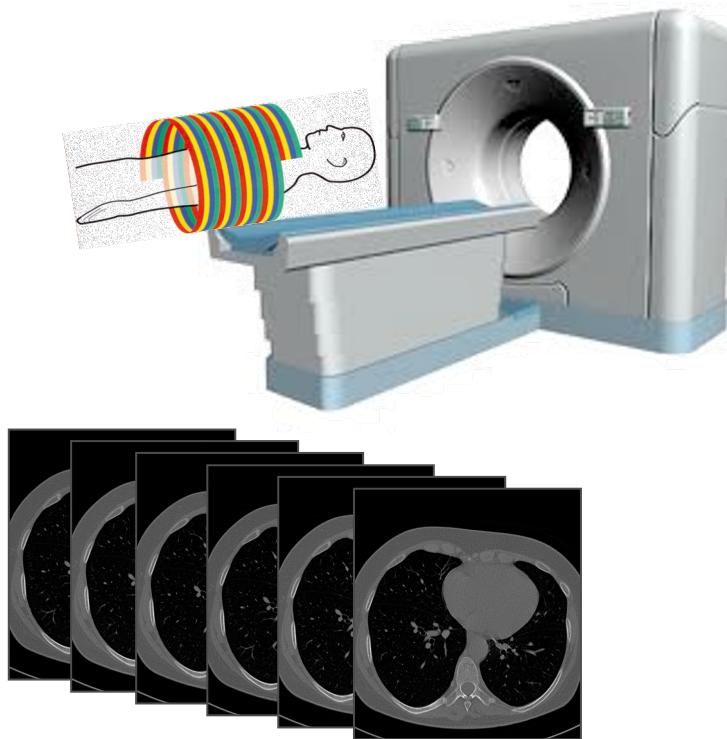
Historical development of CAD systems:

- 1) '90 - Old-fashion systems (Rule-based decision systems)
- 2) 2000-today - Hand-crafted feature and Machine Learning classification
- 3) since 2016 - Deep Learning-based data/image classification

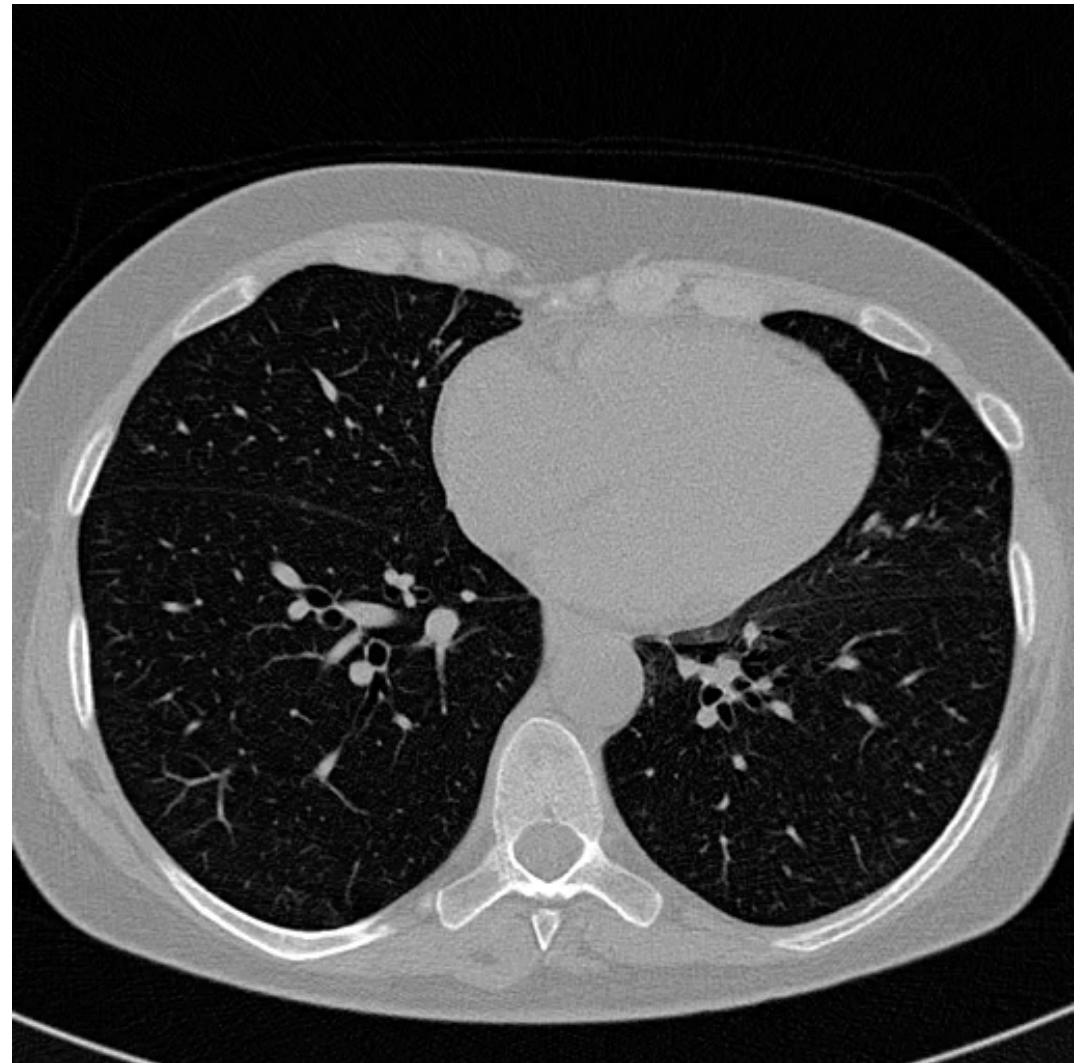
# Data dimensionality: 3D images

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- Computed Tomography

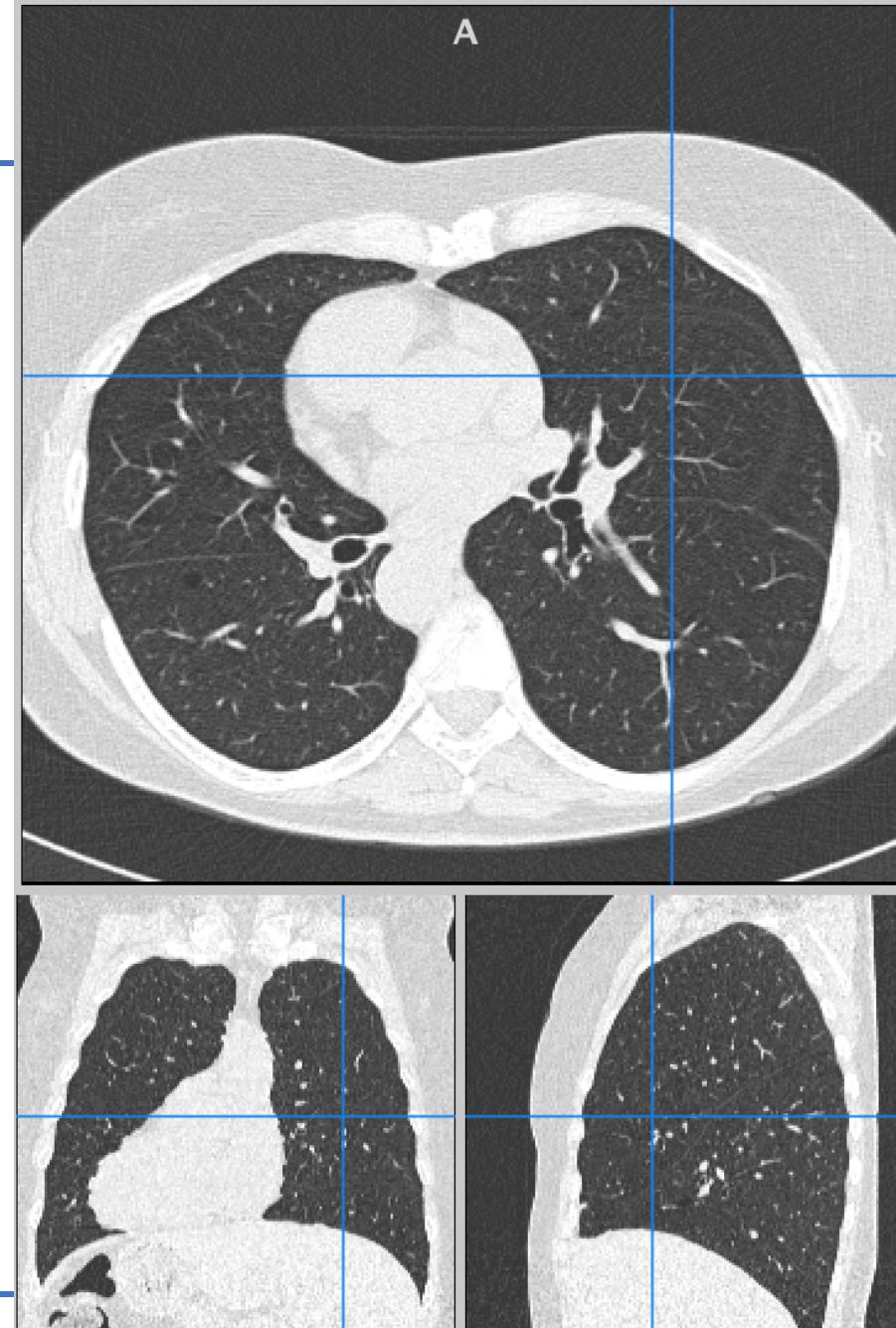
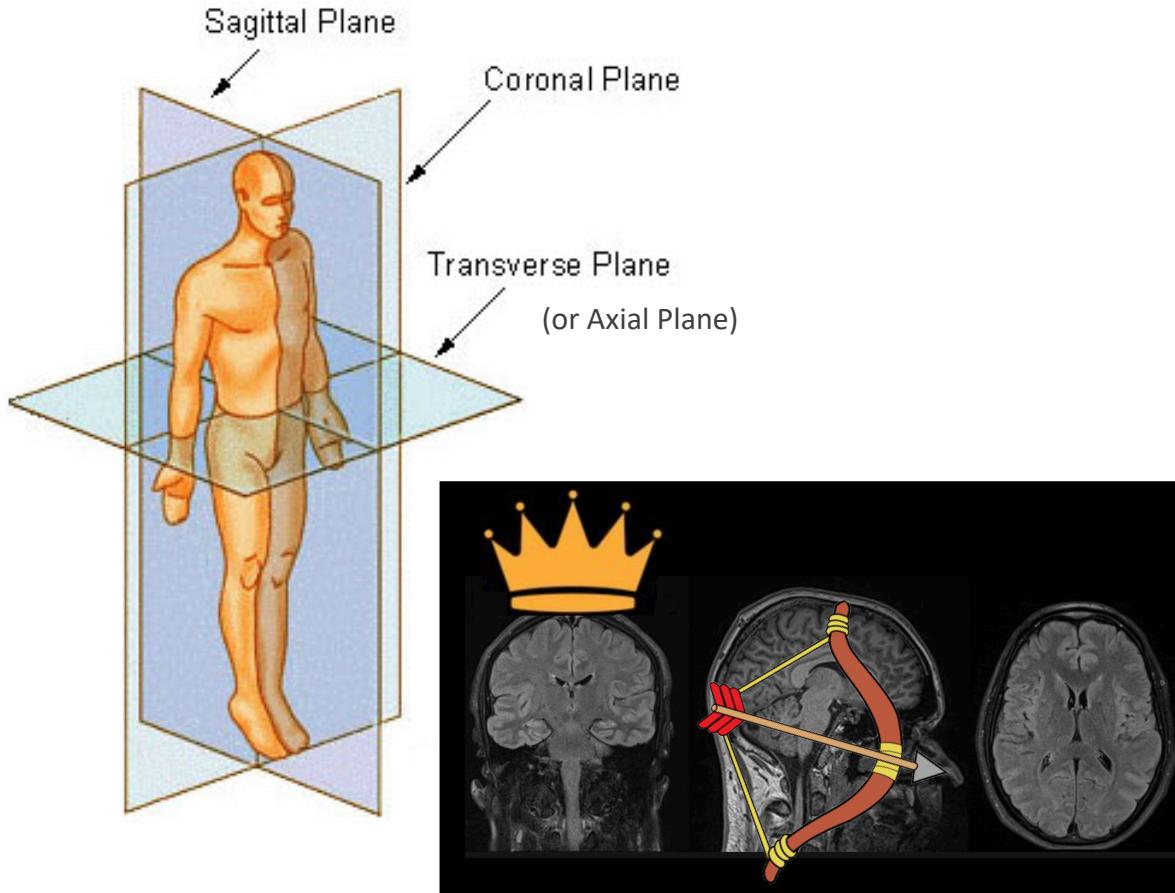


A thin-slice lung CT exam generates  
~300 slices per subject

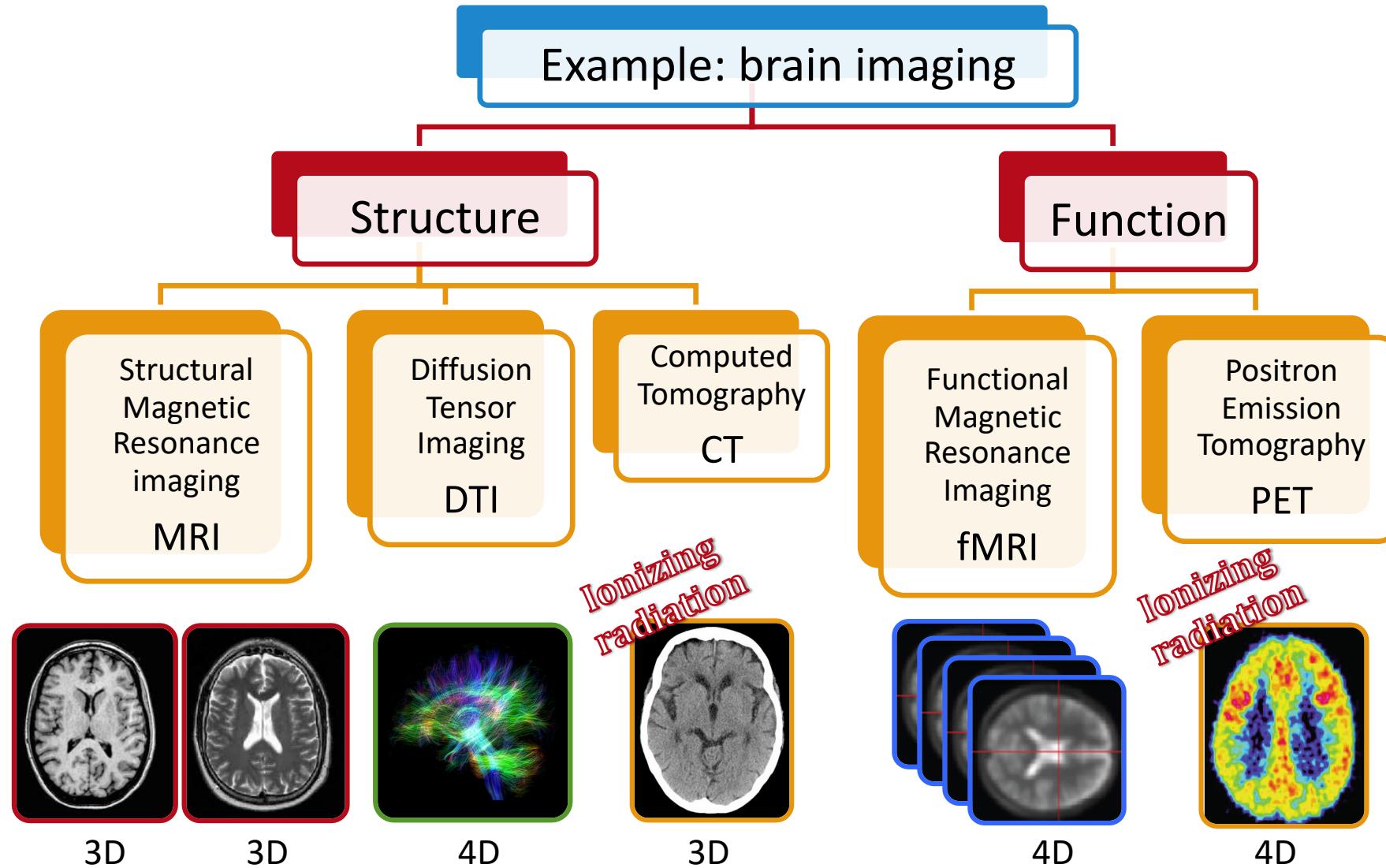


# Volume display planes

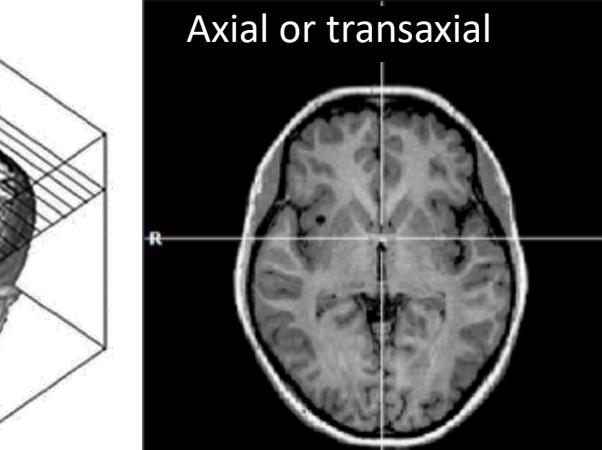
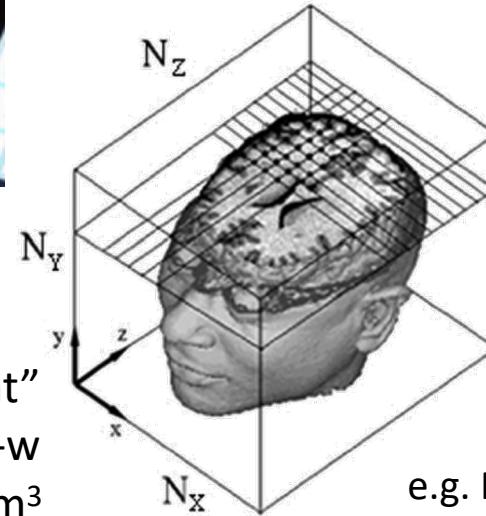
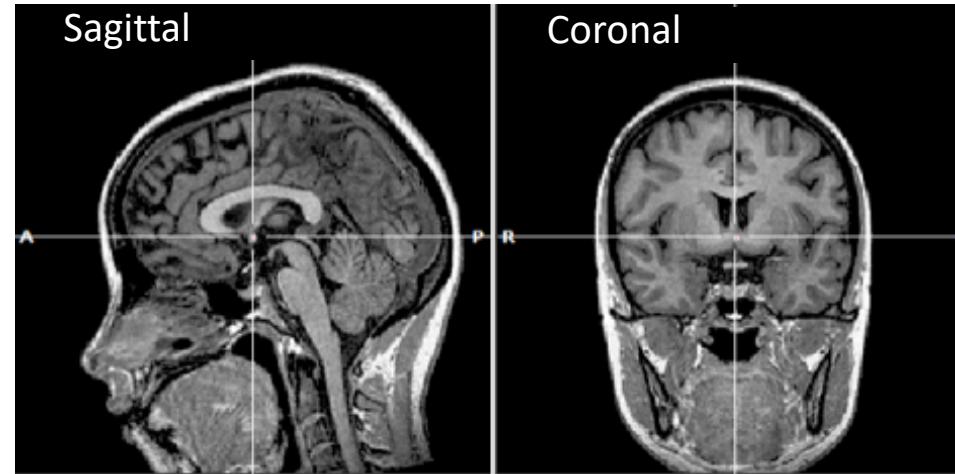
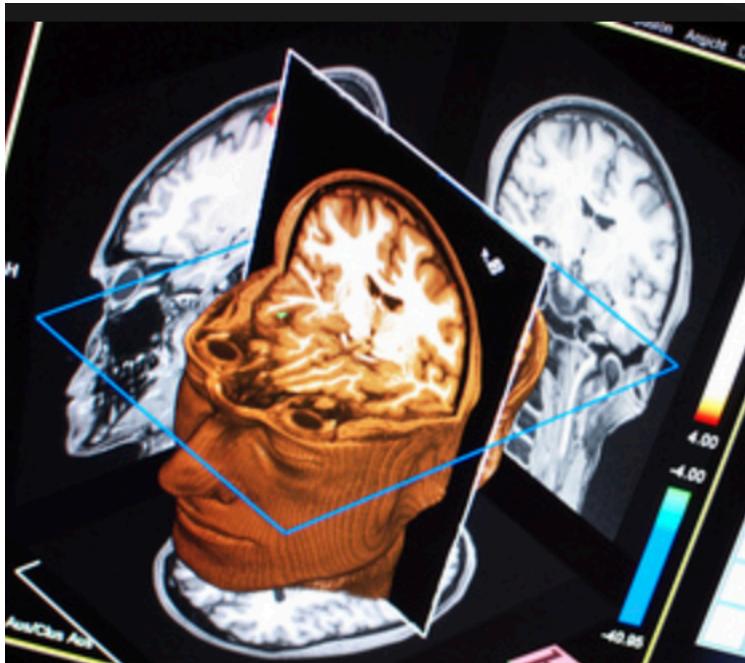
- Medical images are usually displayed by anatomical planes



# Data dimensionality: 3D and more...



# Structural MRI T<sub>1</sub>-weighted images



3D array of data:

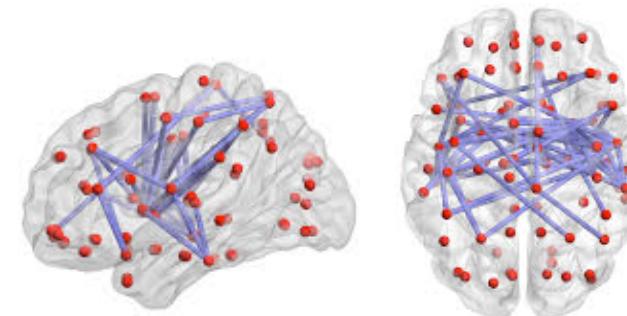
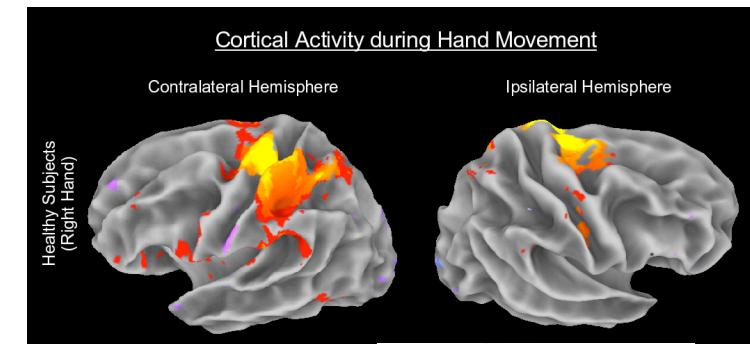
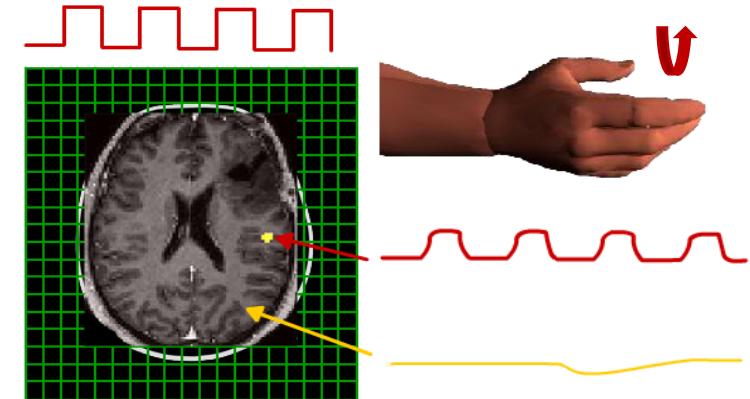


- The voxel is the “volume element”
- Typical “High resolution” MRI T<sub>1</sub>-w images:  $v_x \times v_y \times v_z \approx 1 \times 1 \times 1 \text{ mm}^3$

e.g.  $N_x \times N_y \times N_z = 256 \times 256 \times 150$   
16 bit  
~20 MB

# Functional MRI (fMRI)

- BOLD Response: blood-oxygen-level-dependent (BOLD) contrast
- Typically, 1 volume per second ( $3 \times 3 \times 3 \text{ mm}^3$ ) is acquired for 4-5 min
- Stimuli (visual, auditory, tactile, ...) are administered to the subject during the scan
- Analysis of data time series to look for up-and-down signals that match the stimulus time series

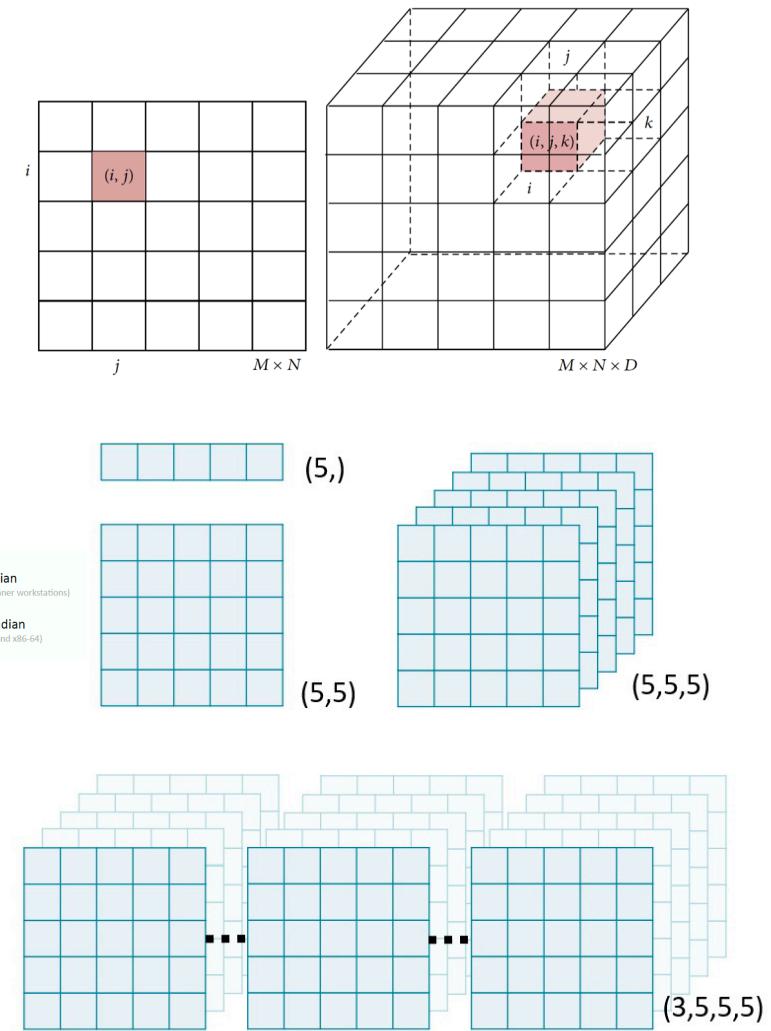
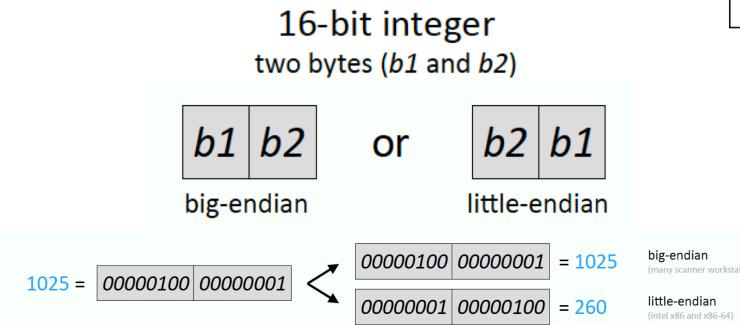


## Functional connectivity

- Resting state rs-fMRI: study of temporal correlations between spatially remote neurophysiological events

# Data dimensionality: 2D/3D/4D/..nD images

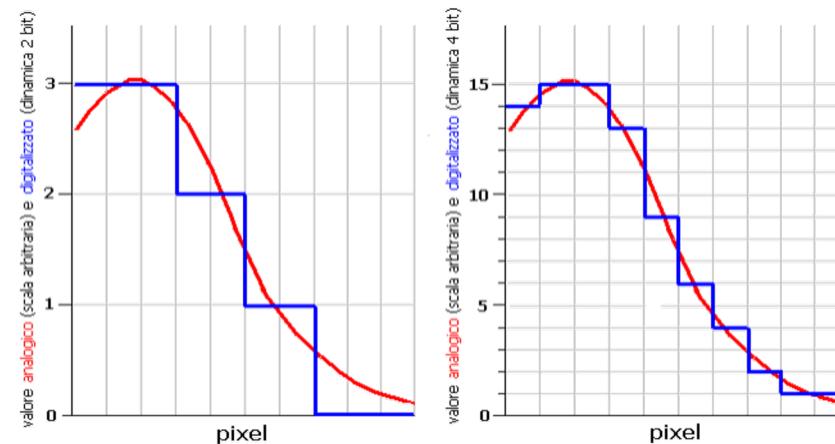
- Image element:
    - pixel: picture element
    - voxel: volume element
  - Images are n-D matrices of numbers (integer, real, complex)
  - Typical data types:
    - signed/unsigned byte (8-bit)
    - signed/unsigned short (16-bit)
    - signed/unsigned int (32-bit)
    - float (32-bit) and double (64-bit)
  - Metadata is required for correct interpretation, e.g.:
    - pixel/voxel dimension (e.g. voxel size (mm):  $v_x, v_y, v_z$ )
    - machine settings
    - parameters used to produce the image
    - patient's positioning
    - patient's information (age, sex,...)
    - Intensity scaling information (only integers are often stored)



# Errors introduced by the analog to digital conversion

- Most medical imaging devices need to convert analog signals into digital ones
- The bit dynamic range should be appropriate to catch the signal variability with minimum errors

2-bit vs. 4-bit dynamics



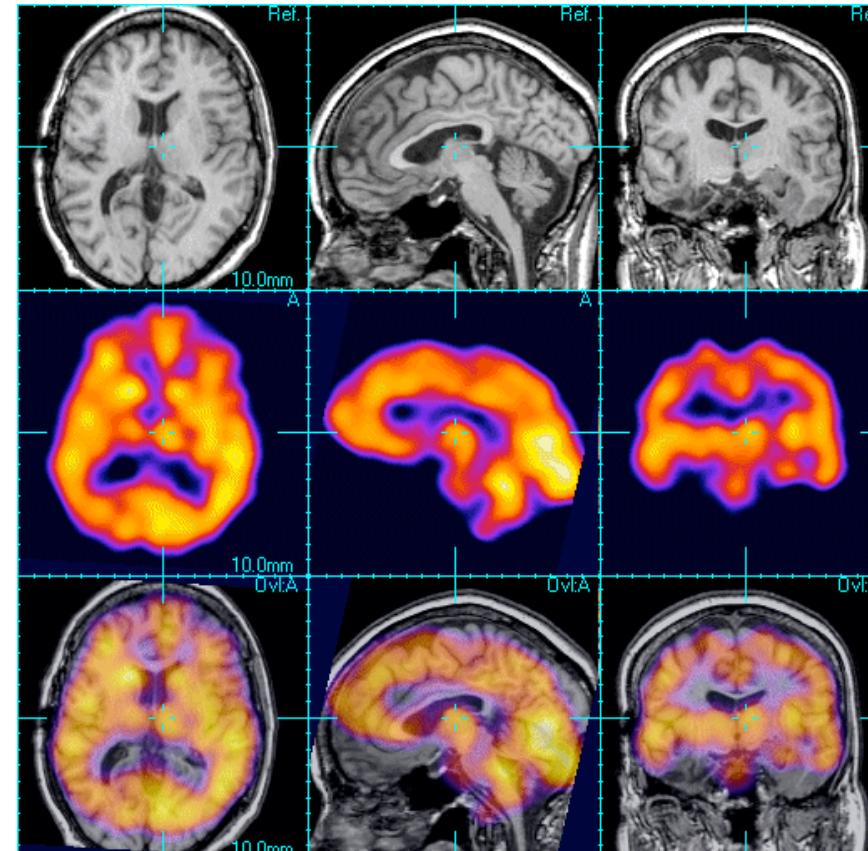
N. bit	Conversion dynamics	Decimal (binary) range of values	Maximum conversion error
2	$2^2 = 4$	from 0 (00) to 3 (11)	$1/4 * 1/2 = 12.5\%$
4	$2^4 = 16$	from 0 (0000) to 15 (1111)	$1/16 * 1/2 = 3.1\%$
8	$2^8 = 256$	from 0 (00000000) to 255 (11111111)	$1/256 * 1/2 = 0.2\%$
12	$2^{12} = 4096$	from 0 (000000000000) to 4095 (111111111111)	$1/4096 * 1/2 = 0.012\%$
16	$2^{16} = 65536$	from 0 (0000000000000000) to 65535 (1111111111111111)	$1/65536 * 1/2 = 7 \cdot 10^{-4} \%$

# Image representation with gray and color scales

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- Usually, anatomical images are represented in gray scales, functional images and parametric maps in color scales

- Structural image

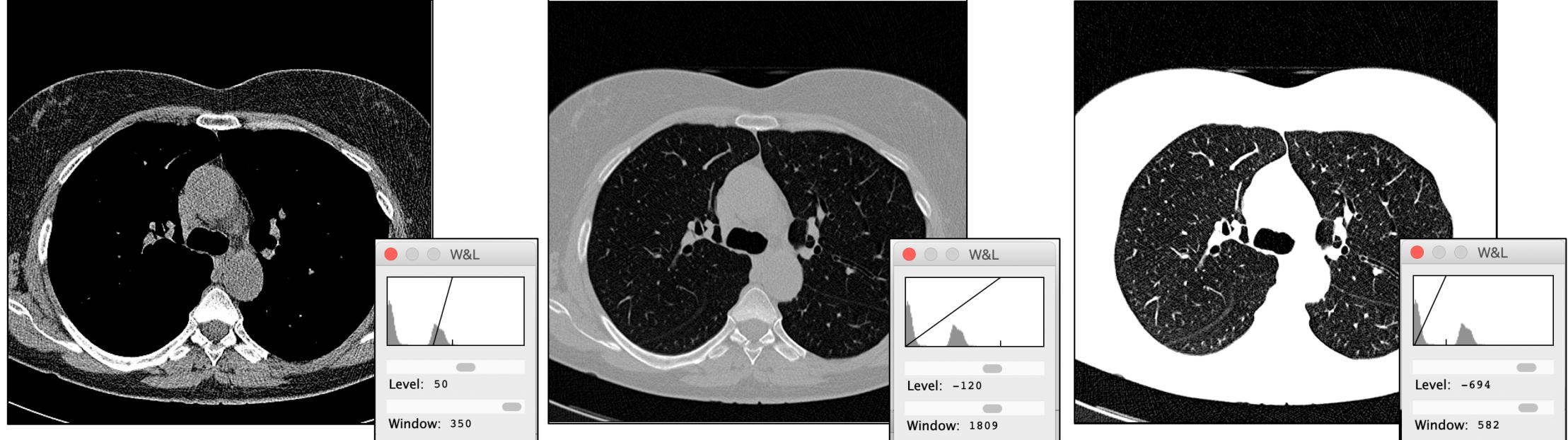


- Functional image

- Image overlay

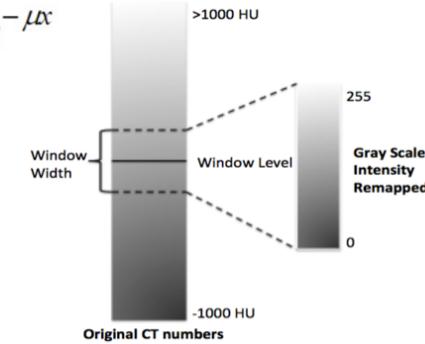
# Image windowing

- In Computed Tomography (CT) voxel intensities are expressed in Hounsfield Units (HU)
- The HU scale is a linear transformation of the original linear attenuation coefficient measurement into one in which the radiodensity of distilled water at standard pressure and temperature is defined as zero HU
- HU values of human tissues range in [-1000, 1000]
- Windowing:
  - To visualize CT images a window level (WL) and a window width (WW) should be chosen
  - A linear transform is applied that maps the lower bound to zero and the upper bound to the maximum gray level (i.e., 255 for 8 bit images) for visualization



$$HU = 1000 \times \frac{\mu - \mu_{\text{water}}}{\mu_{\text{water}} - \mu_{\text{air}}}$$

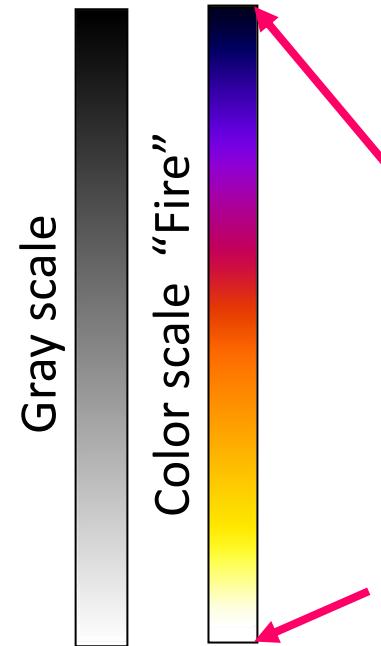
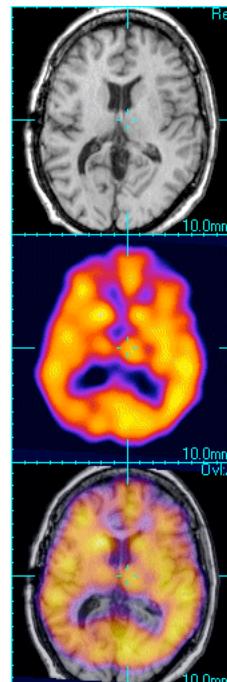
$$I = I_0 e^{-\mu x}$$



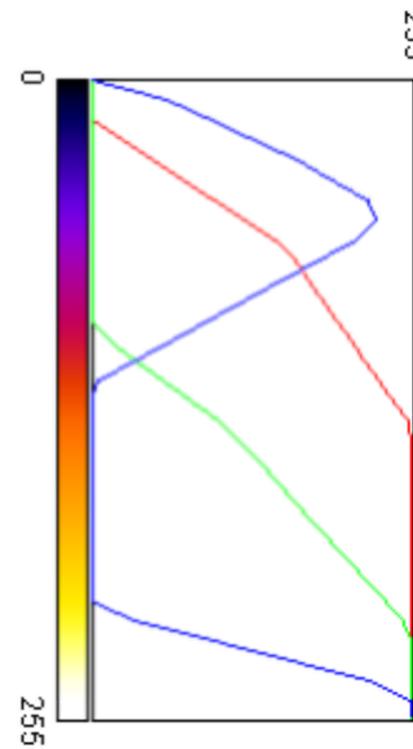
Hounsfield Units for human body	
Bone	1000
Liver	40 to 60
White Matter	46
Grey Matter	43
Blood	40
Muscle	10 to 40
Kidney	30
Cerebrospinal Fluid	15
Water	0
Fat	-50 to -100
Air	-1000

# Look-up tables (LUTs)

- Look-up tables (LUTs) are used for pseudo coloring.
  - Computer graphic boards typically limit the number of gray scales to 256 (8 bit), but offer  $256^3$  ( $\sim 16 \cdot 10^6$ ) colors.
  - Pseudo coloring allows presentation of data without reducing the information as it would result from windowing



LUT RGB			
value	red	green	blue
0	0	0	0
1	0	0	7
2	0	0	15
...			
128	240	79	0
...			
255	255	255	255



# Medical image file formats

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- The file format describes how the image data are organized in the file and how pixel data should be interpreted by a software for correct reading and visualization
- Numerical values of the pixels depend on image modality, acquisition protocol, reconstruction and applied post processing algorithms:
  - A medical image which is separated from the context information is meaningless
- Medical image file formats belong to two categories:
  - Those intended to standardize images generated by different diagnostic modalities, e.g. the DICOM standard
  - Those aiming to facilitate the post-processing analysis (e.g. NIfTI for neuroimaging)

# DICOM file format

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- In response to the increased use of digital images in radiology the American College of Radiology (ACR) and the National Electrical Manufacturers Association (NEMA) formed a joint committee in **1983** to create a **standard format for storing and transmitting medical images**.
- The committee published the original ACR-NEMA standard in **1985**.
- This standard has subsequently been revised and in **1993** it was renamed **DICOM**.

- **Digital Imaging and COmmunication in Medicine (DICOM)**:
  - It is both a file format and communication protocol
  - It is the standard format used in digital imaging medical devices
  - The header and the image are contained in the same file
  - The header contains many information on the imaging device, the acquisition parameters, the patient and the physician
- DICOM supports most imaging modalities

# DICOM file format

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- DICOM is most commonly used for storing and transmitting medical images enabling the integration of medical imaging devices such as scanners, servers, workstations, printers, network hardware, and picture archiving and communication systems (PACS) from multiple manufacturers.
- Pixel data cannot be separated from the description of the image formation procedure:
  - Images should be self-descriptive
- The DICOM format (“anachronistically”) stores volume as a sequence of 2D slices.
  - (A 3D DICOM format exists, but it is not widespread).
  - It is not that bad to have images stored slice by slice, as some acquisition parameters may change slice-wise during the acquisition.
- The DICOM format only stores integer numbers as pixel values, thus a *slope* and an *intercept* to linearly transform data in the allowed range are specified

# DICOM metadata

A **DICOM data element, or attribute**, is composed of the following most important parts:

- a **tag** that identifies the attribute, usually in the format (XXXX,XXXX) with hexadecimal numbers, and may be divided further into DICOM Group Number and DICOM Element Number;
- a **DICOM Value Representation (VR)** that describes the data type and format of the attribute value (e.g. PN: Person Name; DT: Date Time; SS: Signed Short).

Check <https://www.dicomlibrary.com/dicom/dicom-tags/>

- most fields are optional
- there are also vendors' private keys
- some examples:
  - Session Name and Study Number
    - (0008, 0090) ID Referring Physician
    - (0010,0010) Patient's Name
  - Image "Shape"
    - (0028, 0010) IMG Rows
    - (0028, 0011) IMG Columns
    - (0028, 0030) IMG Pixel Spacing
    - (0018, 0050) ACQ Slice Thickness
  - How and where the image data is stored
    - (0028, 0100) IMG Bits Allocated
    - (0028, 0101) IMG Bits Stored
    - (0028, 0102) IMG High Bit

Info for SEO	
0010,0040	Patient's Sex:
0010,1010	Patient's Age: 028Y
0010,1030	Patient's Weight: 60
0010,21B0	Additional Patient History:
0018,0020	Scanning Sequence: GR
0018,0021	Sequence Variant: SS\SK
0018,0022	Scan Options: FAST_GEMS\EDR_GEMS\ACC_GEMS
0018,0023	MR Acquisition Type: 3D
0018,0025	Angio Flag: N
0018,0050	Slice Thickness: 1
0018,0080	Repetition Time: 6.12
0018,0081	Echo Time: 2.256
0018,0082	Inversion Time: 600
0018,0083	Number of Averages: 1
0018,0084	Imaging Frequency: 298.137269
0018,0085	Imaged Nucleus: 1H
0018,0086	Echo Numbers(s): 1
0018,0087	Magnetic Field Strength: 7
0018,0088	Spacing Between Slices: 1
0018,0091	Echo Train Length: 1
0018,0093	Percent Sampling: 100
0018,0094	Percent Phase Field of View: 100
0018,0095	Pixel Bandwidth: 488.281
0018,1000	Device Serial Number: 0000000A5278404
0018,1020	Software Versions(s): 22\LX\MR Software release:7T21.0_V01
0018,1030	Protocol Name: Human V0.5b (Ric)
0018,1088	Heart Rate: 0
0018,1090	Cardiac Number of Images: 0
0018,1094	Trigger Window: 0
0018,1100	Reconstruction Diameter: 256
0018,1250	Receiving Coil: RM:Nova32Ch Head
0018,1310	Acquisition Matrix: 0 256 256 0
0018,1312	Phase Encoding Direction: ROW
0018,1314	Flip Angle: 8
0018,1315	Variable Flip Angle Flag: N

## The Neuroimaging Informatics Technology Initiative (NIIfTI) file format

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Another file format commonly used to store brain imaging data obtained using Magnetic Resonance Imaging methods is the Neuroimaging Informatics Technology Initiative (**NIIfTI**)

- The NIIfTI stored volumes can be:
  - in dual file format: file.hdr, file.img
  - in a single file: file.nii
- NIIfTI is the default file format of most software packages for neuroimaging post-processing:
  - FSL, SPM, itk-SNAP, 3D Slicer, ITK & VTK, nipy, etc.
- The NIIfTI format allows a double way to store the orientation of the image volume in the space:
  1. rotation + translation to be used to map voxel coordinates to the scanner reference frame
  2. 12-parameter or more general transformation adopted to realign the volume to a standard template coordinate system.

# Images viewers

- MANGO: <http://ric.uthscsa.edu/mango/index.html>

## Mango for the Desktop



Mango supports Mac, Windows, and Linux operating systems.

[Download for Mac \(v4.1\)](#)

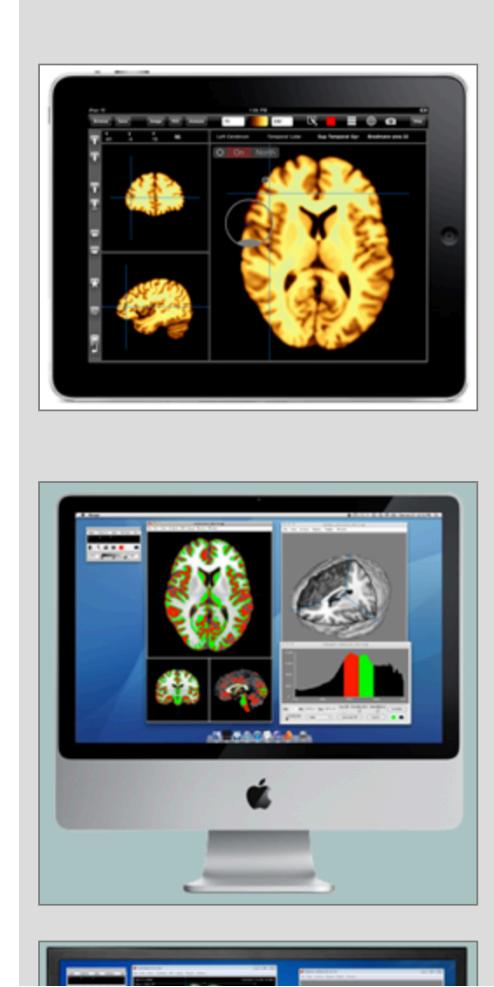
[Download for Windows \(v4.1\)](#)

[Download for Linux \(v4.1\)](#)

[Download Previous Versions](#)

## License

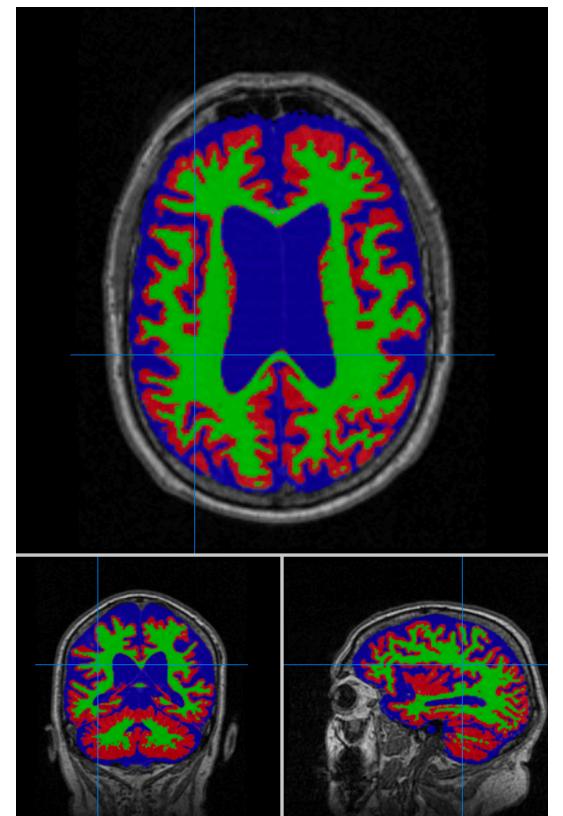
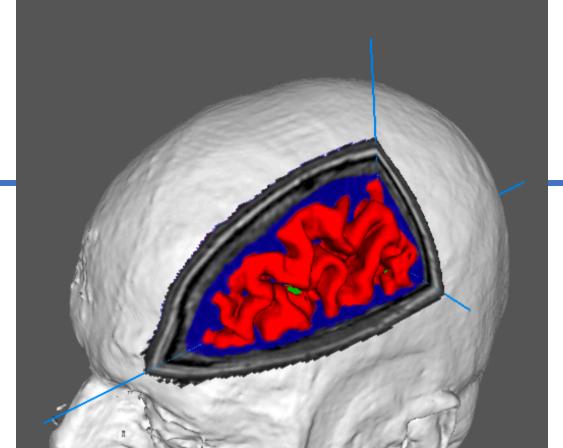
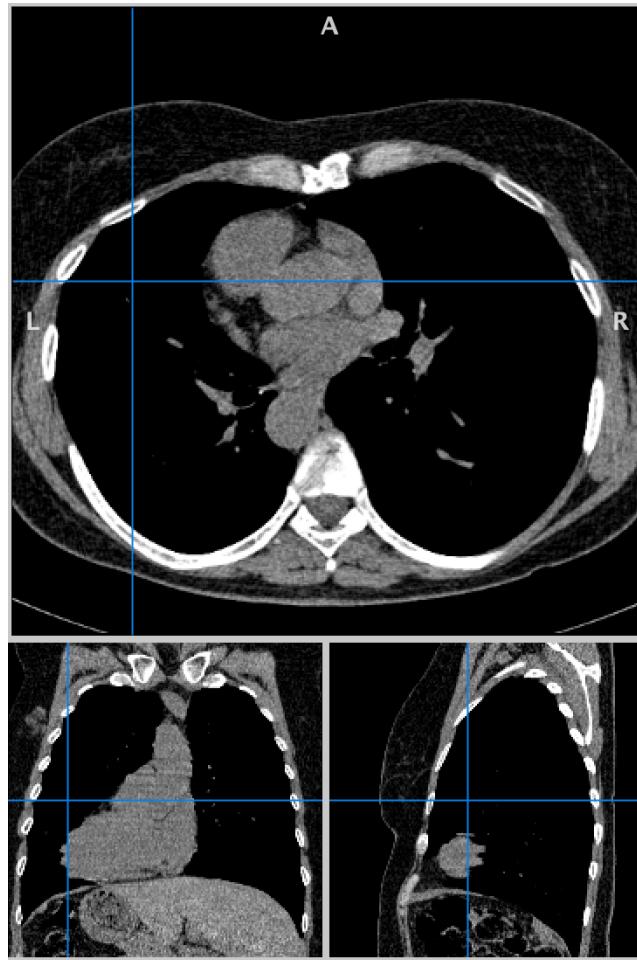
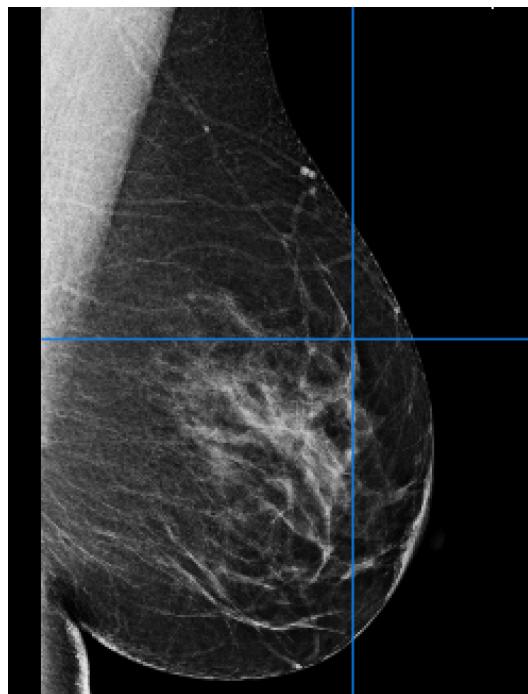
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Mango

# Let's read real images with Mango ...

- Explore 2D, 3D and 4D data sample
- Make image overlays



# Other medical image viewers

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- ImageJ
  - <https://imagej.nih.gov/ij/>
- OsiriX (only for Mac, iPhone, iPad)
  - <https://www.osirix-viewer.com>
- 3DSlicer
  - <https://www.slicer.org>
- itk-SNAP
  - <http://www.itksnap.org/pmwiki/pmwiki.php>
- ...

# The pydicom library

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- Pydicom is a pure Python package for working with DICOM files such as medical images, reports, and radiotherapy objects.
- Pydicom makes it easy to read these complex files into natural pythonic structures for easy manipulation.
- Requirements:
  - numpy library is recommended, (it is only required if manipulating pixel data)
  - matplotlib is necessary to visualize data

See demo code (Jupyter notebooks):

- Lecture1\_demo1\_read\_DICOM\_file.ipynb (read and visualize dicom files, e.g. a 2D slice)
- Lecture1\_demo2\_read\_DICOM\_dir.ipynb (read and visualize dicom dirs, e.g. a 3D volume)

# The General Data Protection Regulation (GDPR)

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- The EU General Data Protection Regulation (GDPR) was approved by the EU Parliament on 14 April 2016
- It is designed to:
  - Harmonize data privacy laws across Europe
  - Protect and empower all EU citizens data privacy
  - Reshape the way organizations across the region approach data privacy
- GDPR and Data Science
  - De-identifying medical imaging is a fundamental prerequisite for data storing, processing and sharing within research projects in order to be compliant with GDPR:
    - Anonymization: using the Hash function (non-reversible)
    - Pseudo-anonymization: data is tokenized, a separate lookup file (with the original entry and the token) is generated and stored in a restricted database.

See demo code (Jupyter notebooks):

- Lecture1\_demo3\_anonymize.ipynb (how to anonymize dicom files)

# References and sources

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- Books
  - The Essential Physics of Medical Imaging, Jerrold T. Bushberg
  - Digital Image Processing for Medical Applications, Geoff Dougherty
  - Handbook of Medical Image Processing and Analysis, Isaac N. Bankman
  - Image Processing and Acquisition using Python, Ravishankar Chityala & Sridevi Pudipeddi
- Sources
  - <https://www.dicomstandard.org>
  - <https://eugdpr.org>
  - <https://pydicom.github.io/pydicom/>

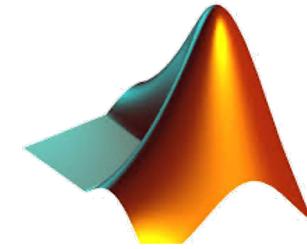
You will find the repository of course materials on [https://github.com/retico/cmepda\\_medphys](https://github.com/retico/cmepda_medphys) and the image data samples to use on <https://pandora.infn.it/public/cmepda> and on [https://drive.google.com/open?id=1YqK7ZkM-P2lrqfD7Pj-SCmjz-GWd\\_1-Y](https://drive.google.com/open?id=1YqK7ZkM-P2lrqfD7Pj-SCmjz-GWd_1-Y)

# To do ....

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University of Pisa has a Campus license for MATLAB

- Download and install MATLAB on your laptop  
<https://start.unipi.it/personale-t-a/strumenti-di-lavoro/strumenti-informatici/software-e-servizi-cloud/software-matlab/>



Current stable release  
is R2020b

Download and install a Dicom viewer, e.g. Mango:

- <http://ric.uthscsa.edu/mango/index.html>

# System requirements and useful MATLAB toolboxes

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- System requirements for installation:
  - Administrative rights
  - **Processors – Minimum:** any Intel or AMD x86-64 processor
  - **RAM – Minimum:** 3.3 GB. **Recommended:** 8 GB
  - **DISK – Minimum:** 3.3 GB of HDD space for **MATLAB** only, 5-8 GB for a typical installation. **Recommended:** An SSD is recommended
- During the installation you have to specify the products to be installed, i.e. the **MATLAB toolboxes**.
- You may add some toolboxes you like to the suggested ones (e.g. the Image Processing, Curve Fitting, Wavelet and Deep Learning toolboxes we will use in the exercises)
- You can add more toolboxes whenever you need from the Add-Ons drop down menu from the MATLAB desktop HOME tab.