

Florence Workshop: "Artificial Intelligence For Physical Science" (AI4PHYS)

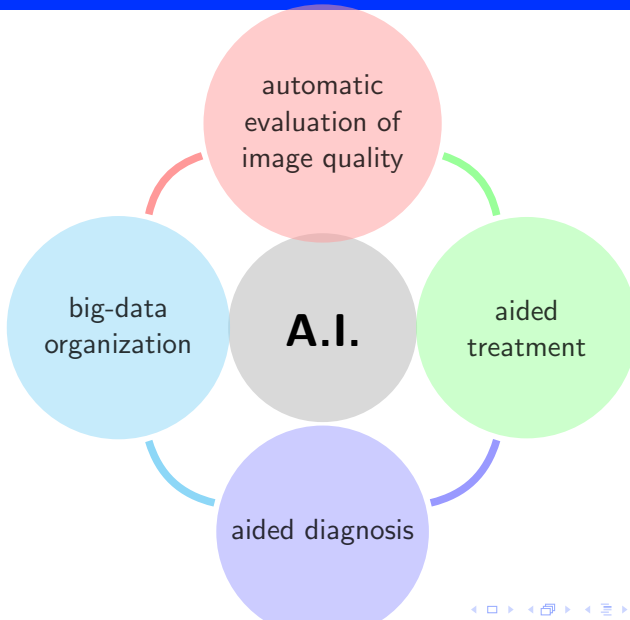
Physics for Medicine



29/01/2025

Sandra Doria

Artificial Intelligence in medicine



Medical imaging modalities

- Computer Vision algorithms have important applications in medical imaging

Nuclear Medicine



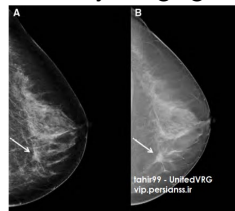
Ultrasound imaging



Magnetic resonance



X-ray imaging



Development of (X-ray) medical imaging

German physics professor Conrad Roentgen



Hand of Anna Bertha Ludwig (1895)

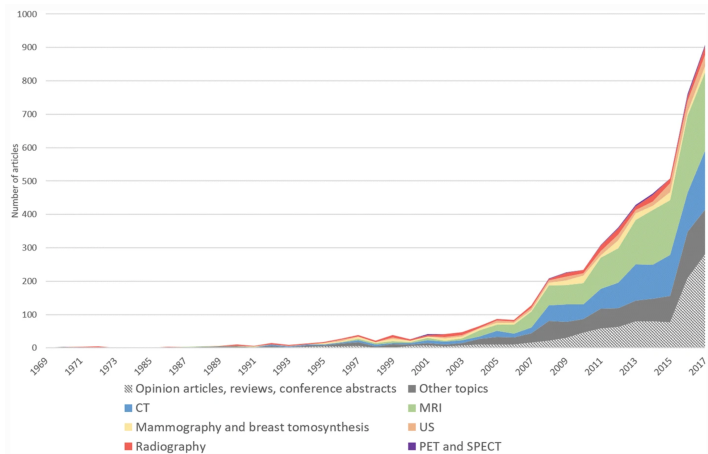


Today



- Understanding of human body
- Computer technology/networking (images sharing among sites)
- Massive amounts of information about patients
- Technology advancements, equipment and digital images

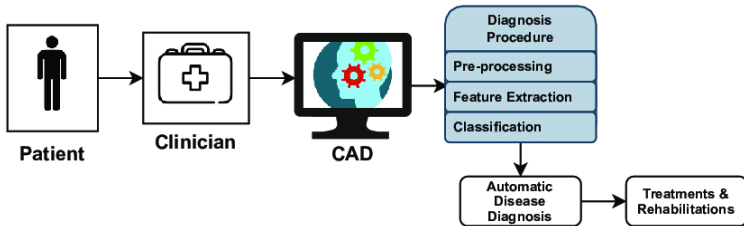
AI in medical imaging



Pesapane et al. *European Radiology Experimental* (2018) 2:35
<https://doi.org/10.1186/s41747-018-0061-6>

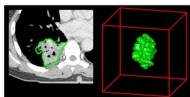
CNN in medical imaging

- **Computed Aided Diagnosis (CAD):** uses a computer program to detect features likely to be of clinical significance in images.

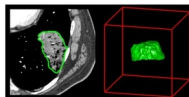


Radiomic analysis

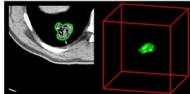
- **Radiomics** uses descriptors aiming to find associations between qualitative and quantitative informations extracted from clinical images and clinical data



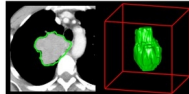
Standard deviation: 208.9
Shape sphericity: 0.46
Texture RLGL nonuniformity: 3,622.3
LoG entropy: 3.89



Standard deviation: 116.0
Shape sphericity: 0.62
Texture RLGL nonuniformity: 5,276.5
LoG entropy: 3.44



Standard deviation: 133.2
Shape sphericity: 0.669
Texture RLGL nonuniformity: 265.1
LoG entropy: 3.93



Standard deviation: 80.0
Shape sphericity: 0.65
Texture RLGL nonuniformity: 9,339.6
LoG entropy: 3.04

- Radiomic features exhibit different levels of complexity and express properties of: lesion shape, voxel intensity histogram, intensity spatial arrangement (texture)

Radiomic examples: pneumonia detection

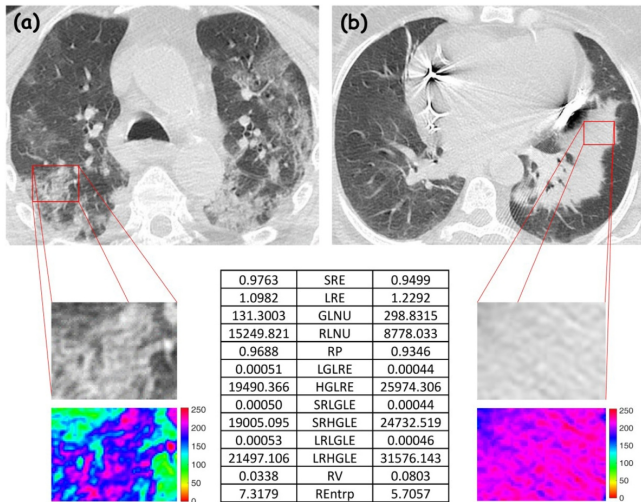
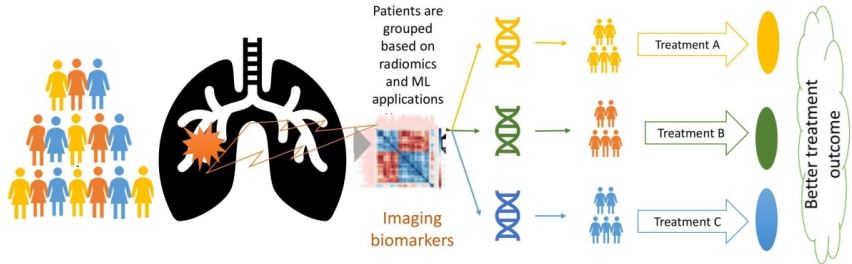


Fig. 7. HRCT images of patient with COVID-19 (a), non-COVID-19 pneumonia and their corresponding GLRLM features.

Why radiomics?

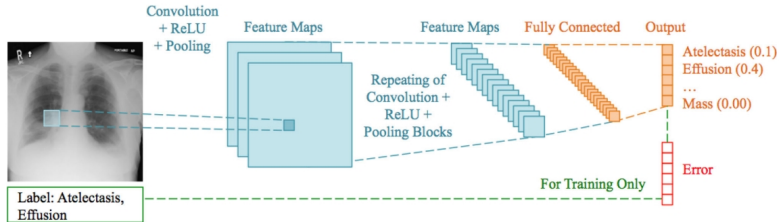
- The goal is to find biomarkers for specific treatment / pathology (e.g. features indicators of tumor aggressiveness)
- Big limitation: standardization of radiomic features across hospitals, protocols, acquisition settings, image processing... Still not overcome!



- Radiomics was used as a pre-processing step for ML algorithms → Current ML algorithms are powerful enough to learn directly from images

CNN in medical imaging

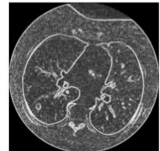
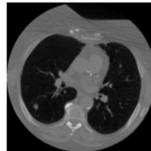
- CNNs are the preferred neural networks in image processing because of their ability to maintain spatial information (segmentation, localization etc.)



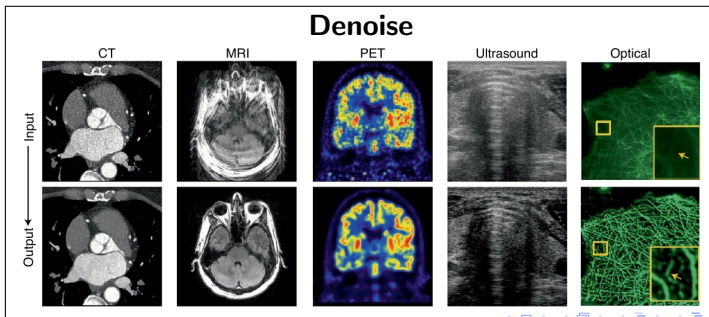
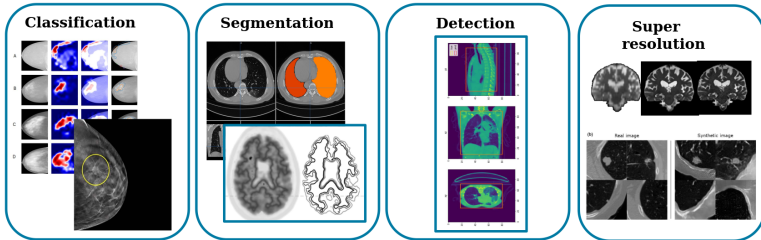
- **Feature extractions:** identifying textures (edges, pixel intensity spatial distributions, spatial periodicity)

↓
Example: detecting edges with the **Sobel filter**

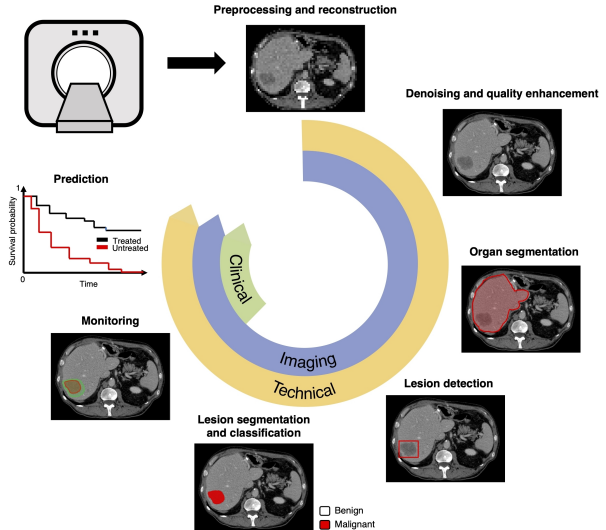
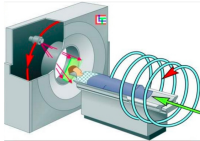
-1	0	-1
-2	0	2
-1	0	1



AI in medical imaging: applications

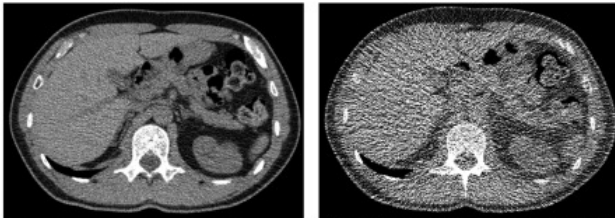


CNN in Computed Tomography



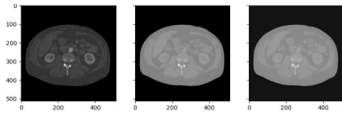
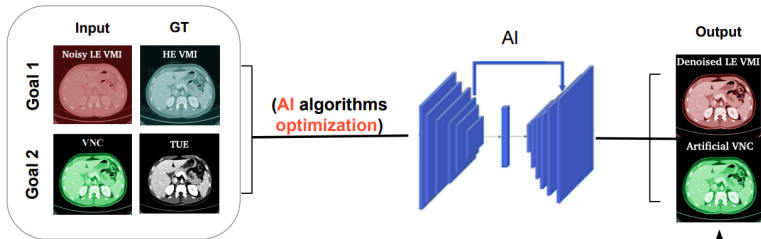
Application: Image Quality enhancement in CT

- Quality enhancement is particularly important in Computed Tomography (and in general x-ray applications) because there is a compromise between radiation dose and noise.



- Increase interest in algorithms for image quality evaluation and enhancement

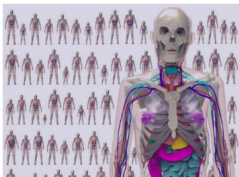
Research line: Dual Energy Computed Tomography



Azienda
Ospedaliero
Universitaria
Careggi

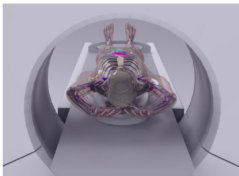


Last frontier: Virtual Patients



TR&D 1 - Virtual Patients

We are developing a framework from which researchers can generate vast populations of realistic, customizable virtual subjects for 3D and 4D research, significantly advancing human modeling to enable virtual imaging trials.

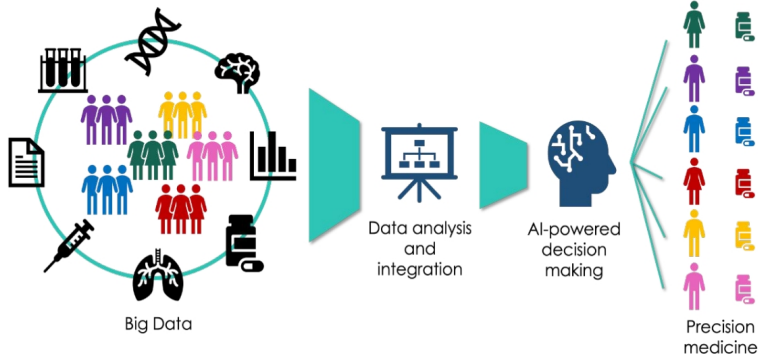


TR&D 2 - Virtual Scanners

We are developing a platform for rapid, accurate, and realistic CT simulations capable of generating 3D and 4D CT images and radiation dose estimates for highly detailed patient-specific anatomies with physiological and perfusion dynamics.

- Virtual datasets represent a powerful method to generate synthetic realistic data to optimize deep learning algorithms for medical imaging

Last frontier: Multimodal analysis



- Multimodal analysis are the future challenge of DL in medicine

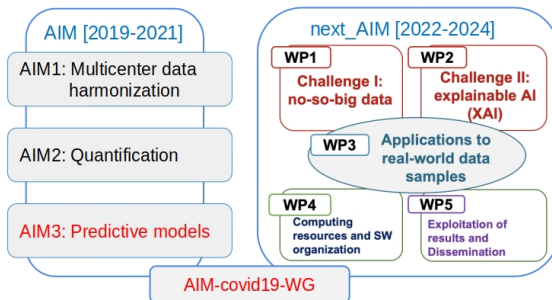
Previous projects @ INFN

Artificial Intelligence to become the next revolution in **medical diagnostics** and **therapy**.

- New image processing and data analysis strategies, including radiomics approaches, need to be developed and extensively validated.

INFN-FLORENCE GROUP

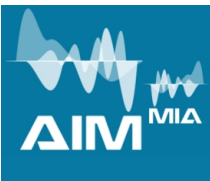
C. Talamonti
L. Ubaldi
S. Pallotta
L. Marrazzo
D. Greto
M. Loi
(S. Piffer
M. Berrettini)



Resp. Naz.: A. Retico

14 Research Units:
Bari (S. Tangaro)
Bologna (D. Remondini)
Cagliari (P. Oliva)
Catania (M. Marrale)
Ferrara (G. Paternò)
Firenze (C. Talamonti)
Genova (A. Chincarini)
Lab. Naz. Sud (G. Russo)
Lecce (G. De Nunzio)
Milano (C. Lenardi)
Napoli (G. Mettivier)
Pavia (A. Lascialfari)
Padova (A. Zucchetta)
Pisa (M.E. Fantacci)

Researchers from INFN divisions and University Departments collaborate closely with Clinicians and Medical Physicists of many Italian hospitals and IRCCS, and with international consortia sharing data



Artificial Intelligence in Medicine: focus on Multi-Input Analysis (AIM_MIA)



2025-2017

General goal: *to take a step forward in the development and validation of AI-based tools for medical data analysis*

Resp. Naz.: A. Retico

12 Research Units:

Bari (S. Tangaro)
Bologna (D. Remondini)
Cagliari (P. Oliva)
Catania (M. Marrale)
Ferrara (G. Domenico)
Firenze (C. Talamonti)
Genova (A. Chincarini)
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(G. Russo)
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Objectives

1. Mining multi-modal information
2. Handling incomplete/missing/limited datasets
3. Development of a dedicated data and computing platform

