

Data Science and AI for Medicine Training School

TRAINING: Application of Deep Learning in Medical Imaging (Radiology)

SPEAKERS: Leo Misera, Asier Rabasco

GEFÖRDERT VOM



Bundesministerium
für Forschung, Technologie
und Raumfahrt



Diese Maßnahme wird gefördert durch die Bundesregierung
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der Grundlage des von den Abgeordneten des Sächsischen
Landtags beschlossenen Haushaltes.



Come2Data
Kompetenzzentrum für
interdisziplinäre Datenwissenschaften

Data Science and AI for Medicine Training School
Training: Application of Deep Learning in AI
(Radiology)

Slide 1

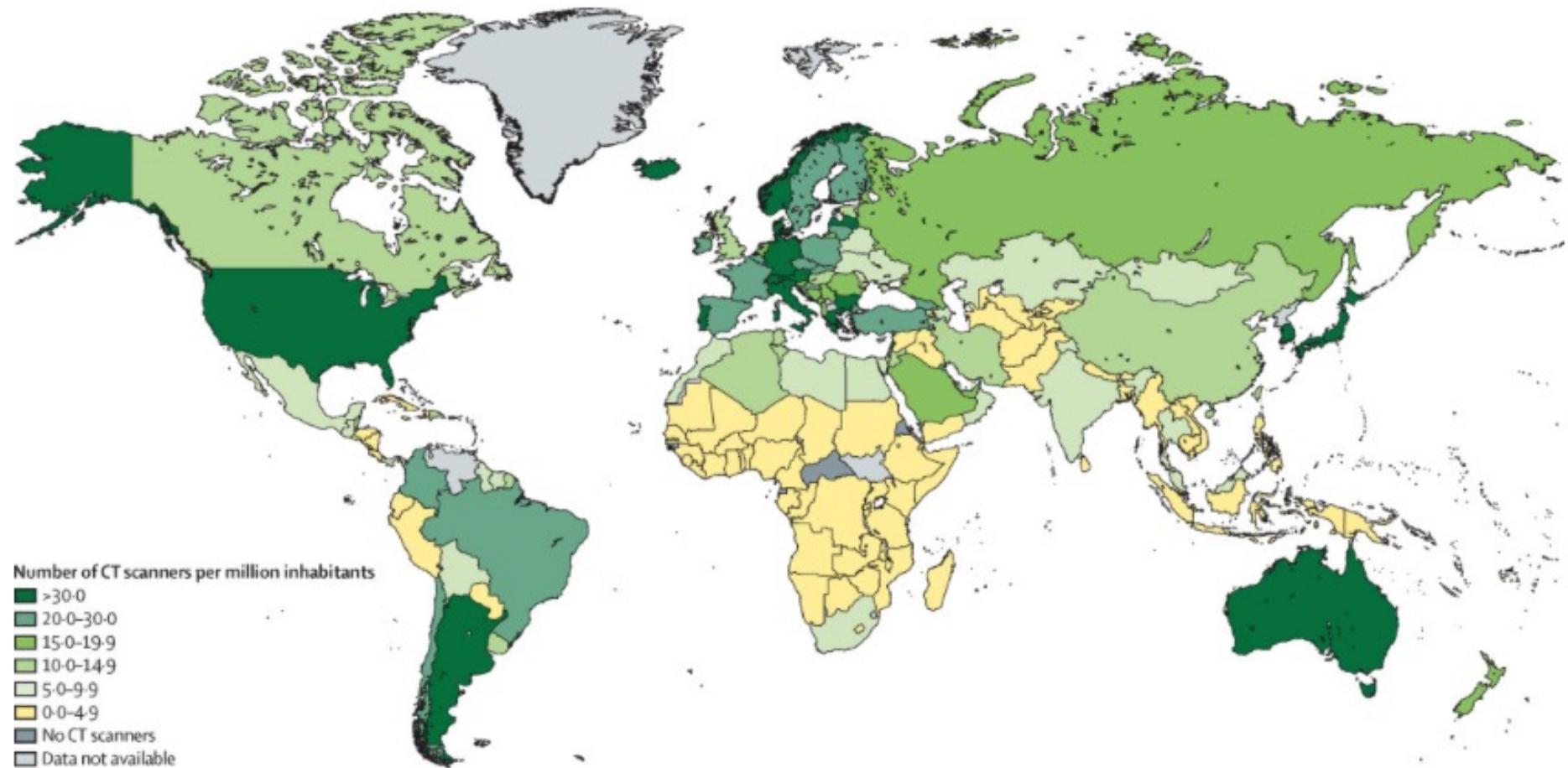
ScaDS.AI
DRESDEN LEIPZIG

Deep Learning is everywhere...



And so is
Radiology!

Radiology in Medicine: how every is everywhere?





**Radiology in
Medicine**

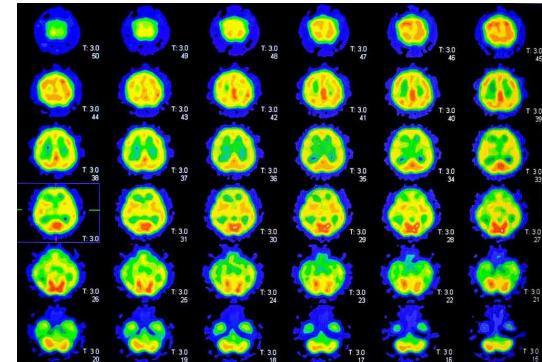
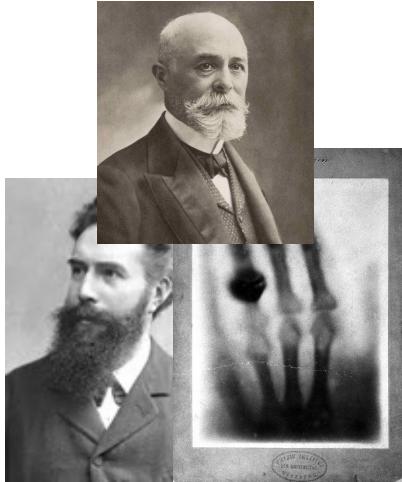
**Imaging
modalities**

**Applications
of Radiology
in AI**

**Hands-On
Session**

Radiology in Medicine: history and definition

Radiology: leverage of human body imaging for treatment guidance and diagnosis



1895-1896
Discovery of X-rays
and
radioactivity¹

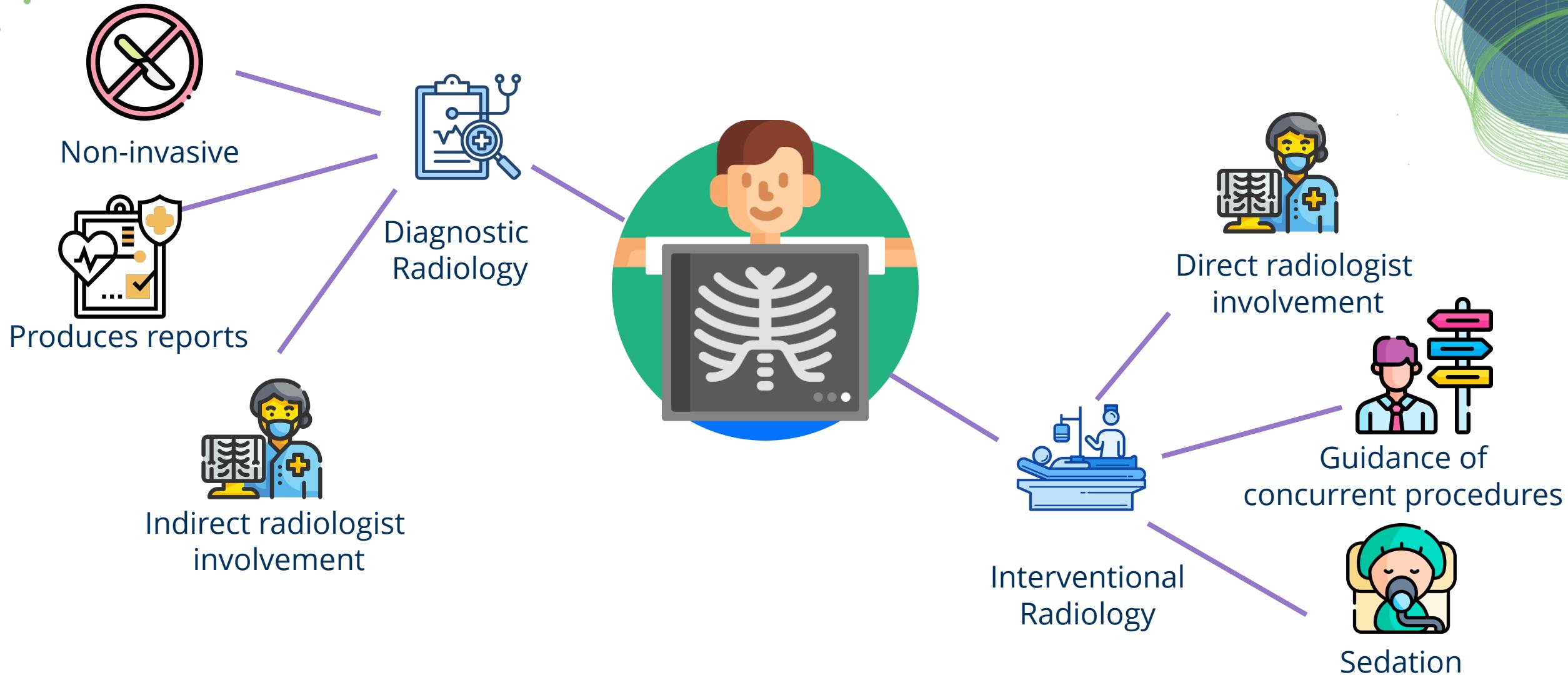
1914-1918
Film used for
radiology
Usage in WWI

1946-1958
Discovery of NMR
Usage of ultrasound in
gynecology

1970s-1980s
First CT and MRI
images

1990s+
Refinement of
radiological
technologies²

Radiology in Medicine: subtypes





**Radiology in
Medicine**

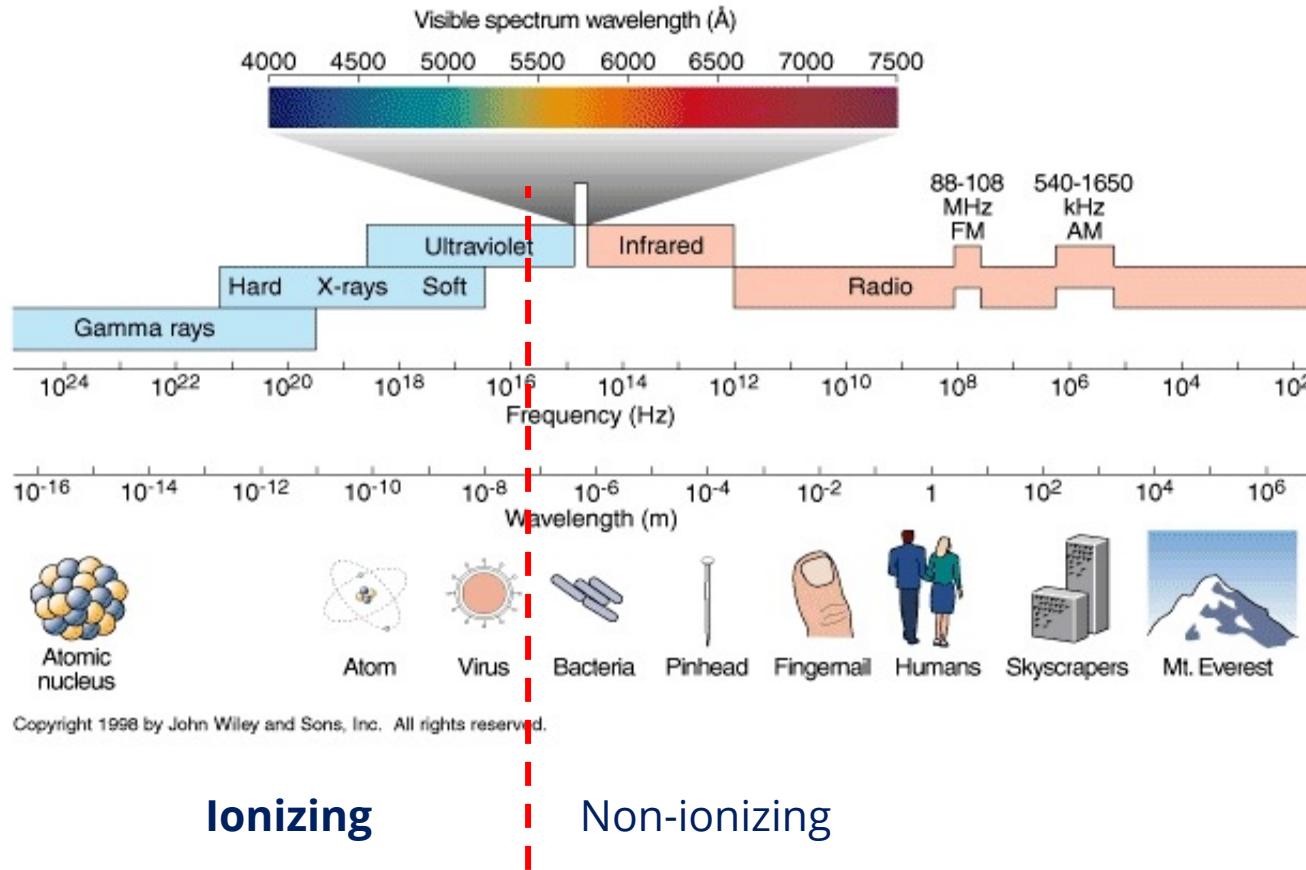
**Imaging
modalities**

**Applications
of Radiology
in AI**

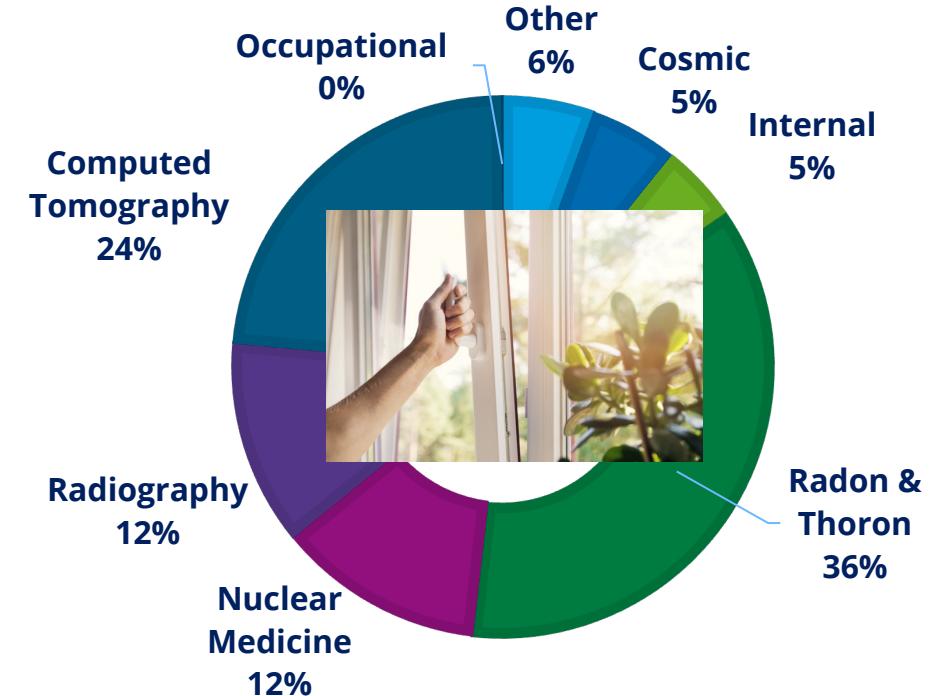
**Hands-On
Session**

Imaging modalities: brief summary of radiation

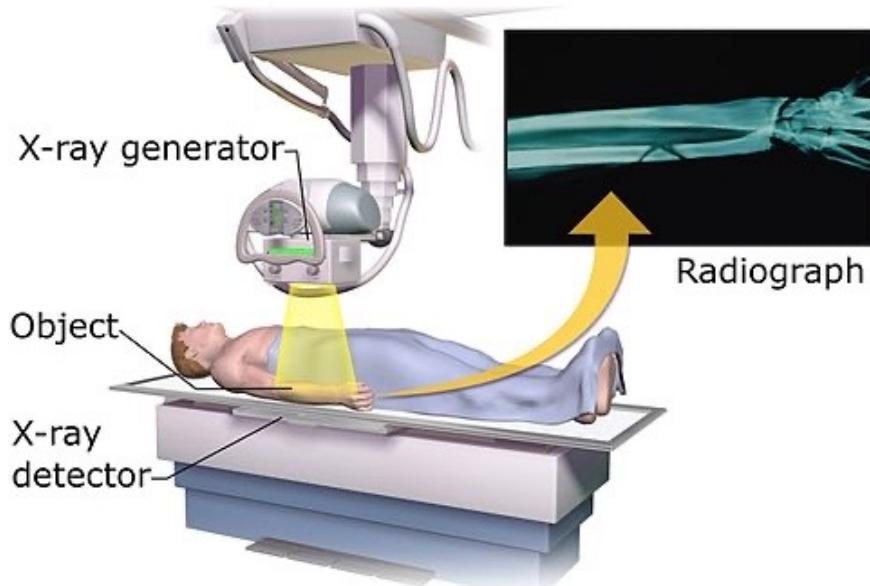
Radiation: emission/transmission of energy through space via waves or particles



AVERAGE DOSE OF IONIZING RADIATION PER YEAR³



Imaging modalities: projectional radiography



X-rays are ionising sources of radiation.
Small doses used to produce 2D images of body structures



Contrast can be limited due to overlapping of structures in one single view

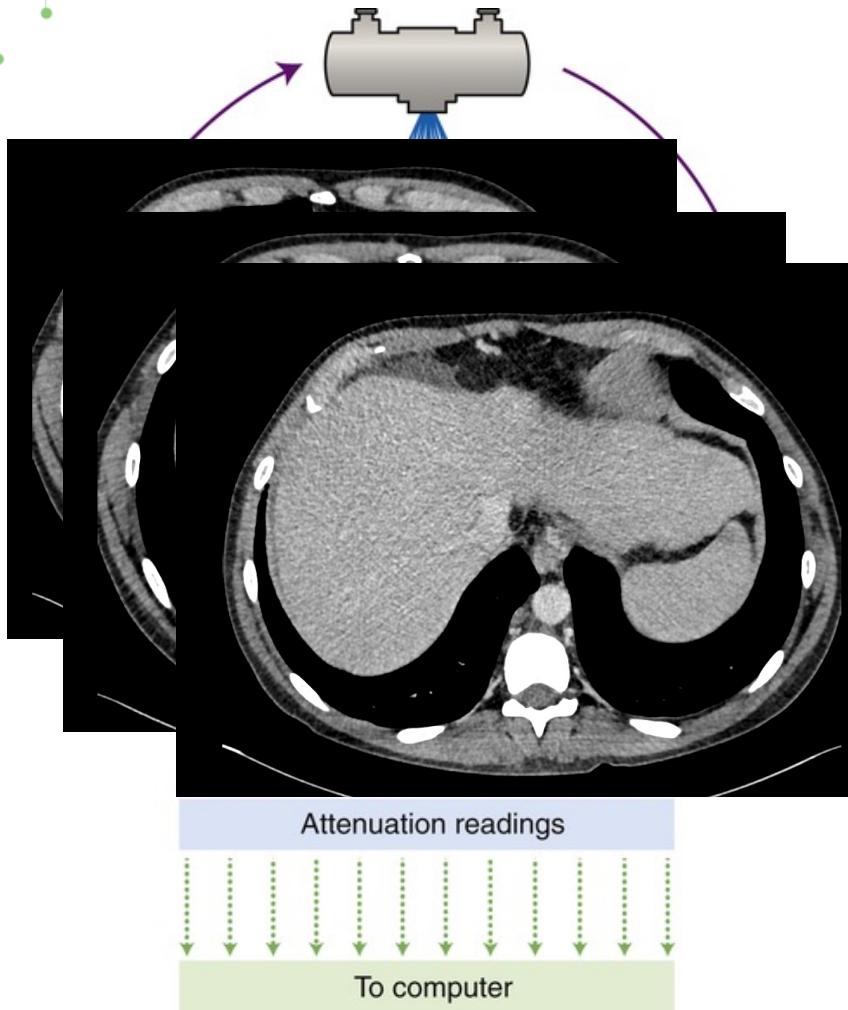


Radiography is used to diagnose broken bones, foreign objects in soft tissue or screen for infections



Oldest and most used form of medical imaging

Imaging modalities: computerised tomography



More radiation than X-rays, but allows for 3D scanning of the body/area of interest



Contrast is used intravenously to highlight different parts of the anatomy in real time

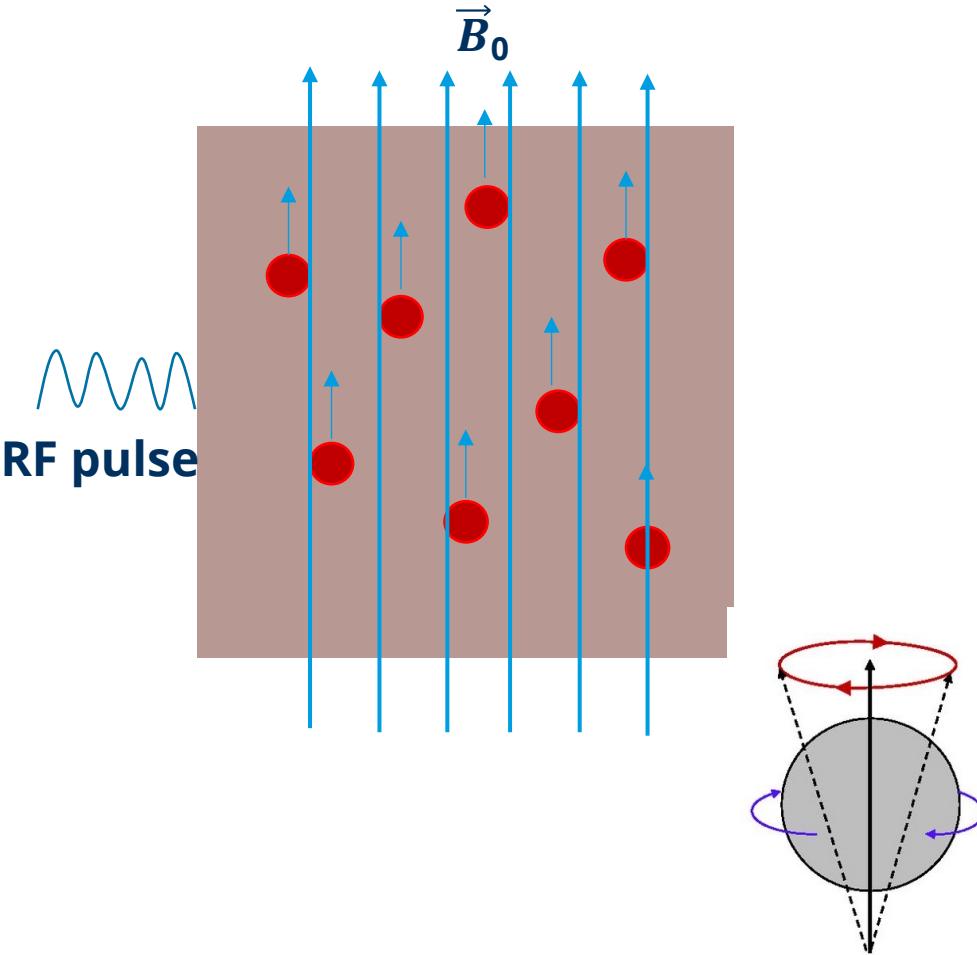


Allows for locating lesions within the body, assess sizes and make first impressions on a diagnosis



CT units have physical meaning, related to the attenuation of water.

Imaging modalities: magnetic resonance imaging



MRI does not produce ionizing radiation, it is based on nuclear magnetic resonance from hydrogen nuclei (protons)



MRIs are better at contrast resolution than CTs but lower at spatial resolution

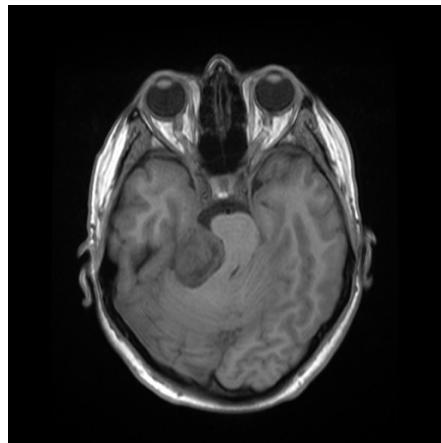


MRI is very very diverse. Sequences use different resonance aspects, highlighting different phenomena

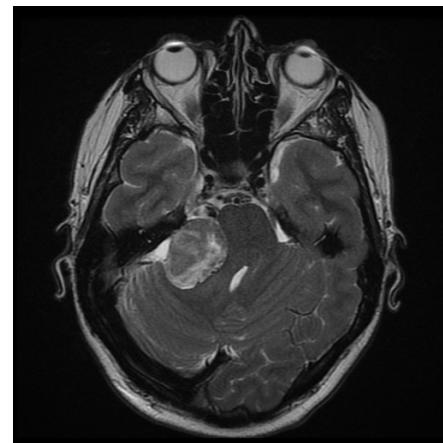


MRI units are dimensionless and can vary from person to person for the same exam.

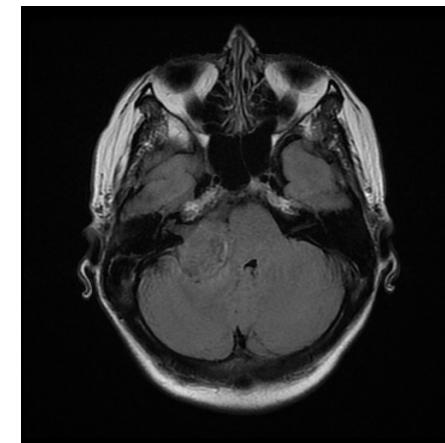
Imaging modalities: magnetic resonance imaging



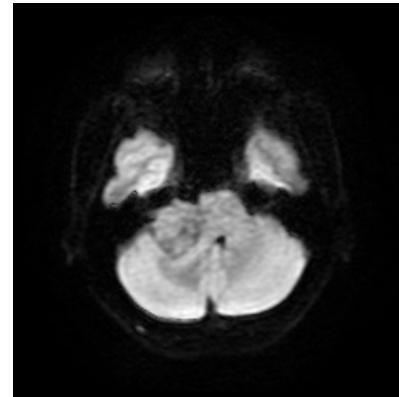
T1-weighted



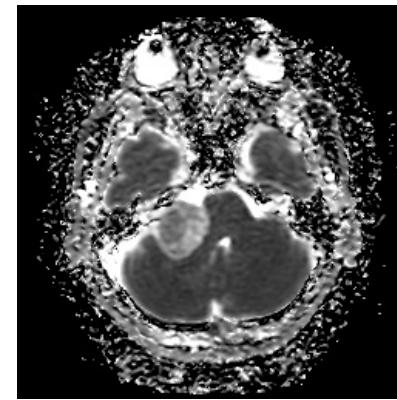
T2-weighted



FLAIR



Diffusion-weighted



Apparent diffusion
weighted



**Radiology in
Medicine**

**Imaging
modalities**

**Applications
of Radiology
in AI**

**Hands-On
Session**



Applications of Radiology in AI



“Radiologists will be obsolete in 5 years”

**George Hinton, 2016, Godfather of AI and Nobel prize in Physics.
Not a radiologist**

Applications of Radiology in AI: the need for AI

Prevalence of burnout amongst German radiologists: A call to action

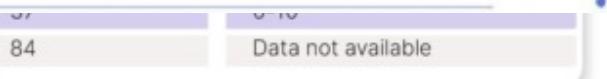
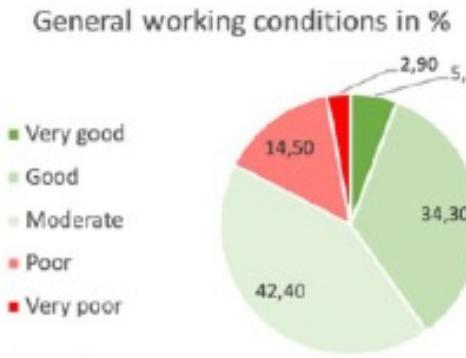
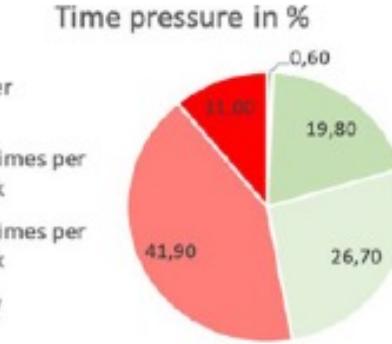
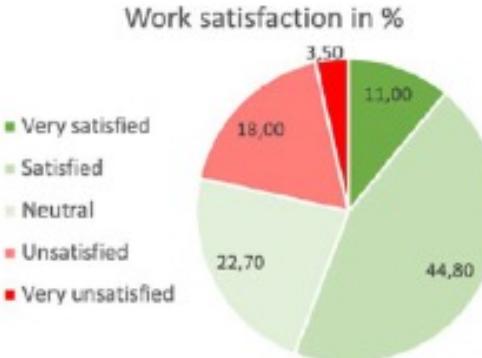
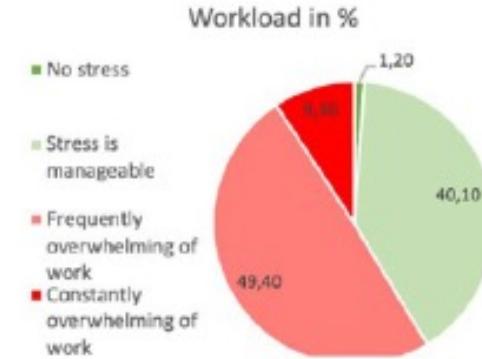
ESR European Society of Radiology

Burnout is prevalent among medical professionals, including radiologists

A burnout-rate of 76.7% was observed in 172 participants

Distributed burnout focused questionnaires to members of the German Society of Radiology and Interventional Radiology

Further observations included:



Map of radiologists from: <https://harrison.ai/news/global-radiology-6-key-challenges-how-ai-can-help/>

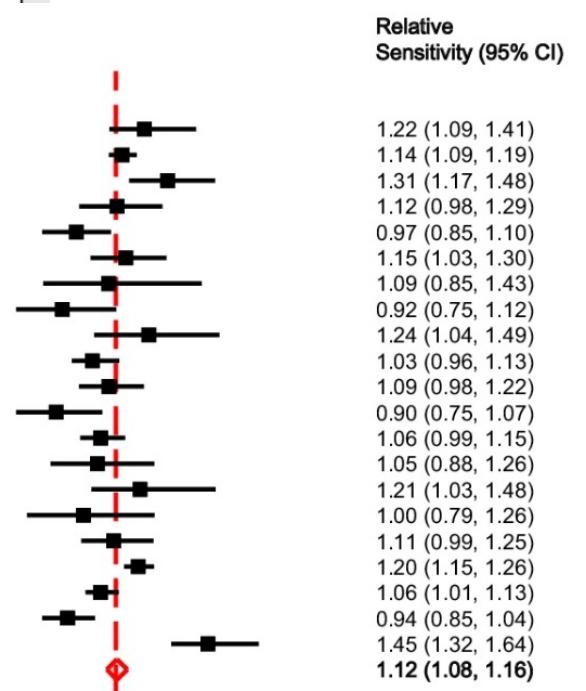
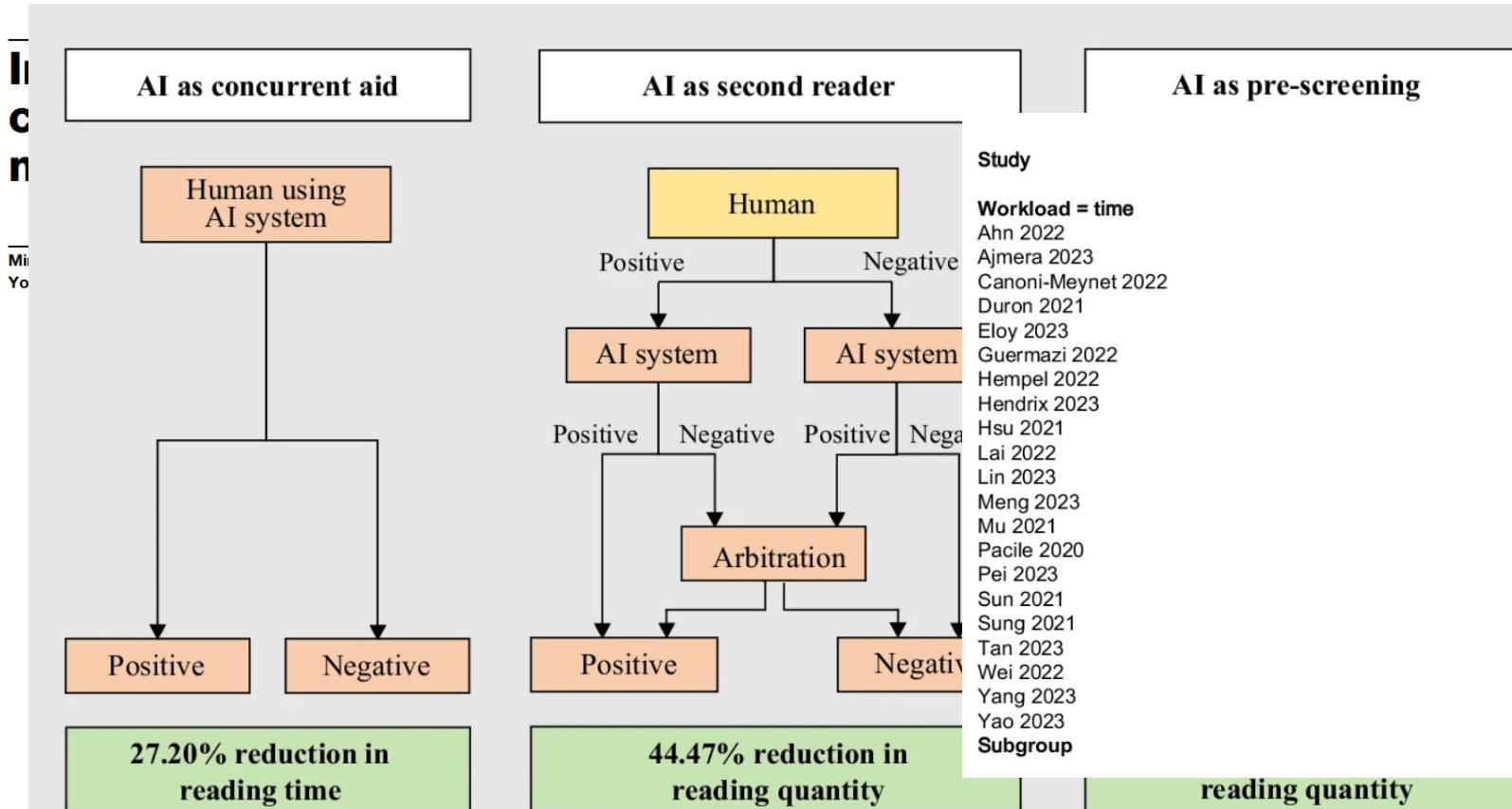
Burnout abstract:
Bastian, M. et al. European Radiology (2025).

Applications of Radiology in AI: synergistic fields

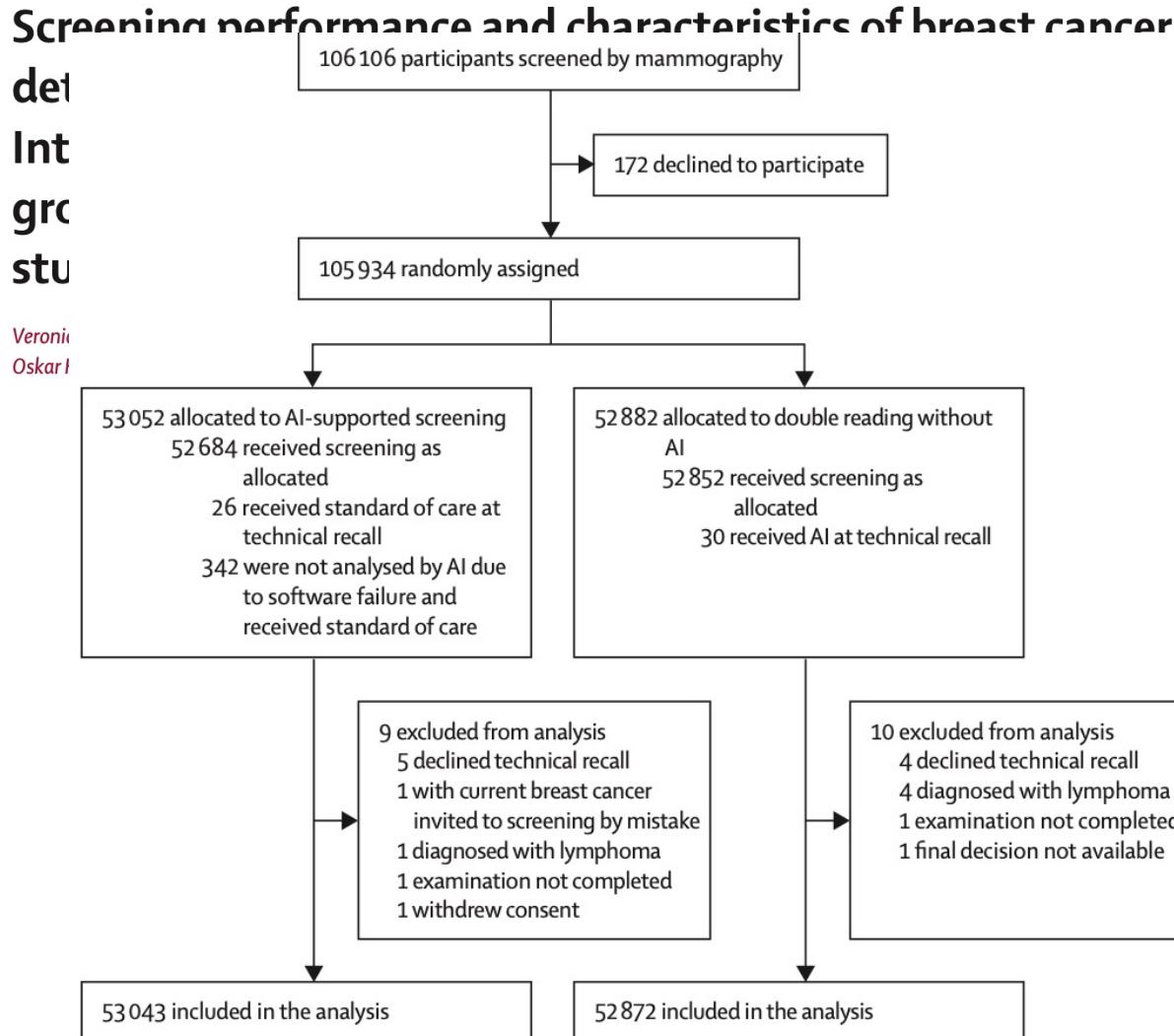
npj | digital medicine

Published in partnership with Seoul National University Bundang Hospital

Article

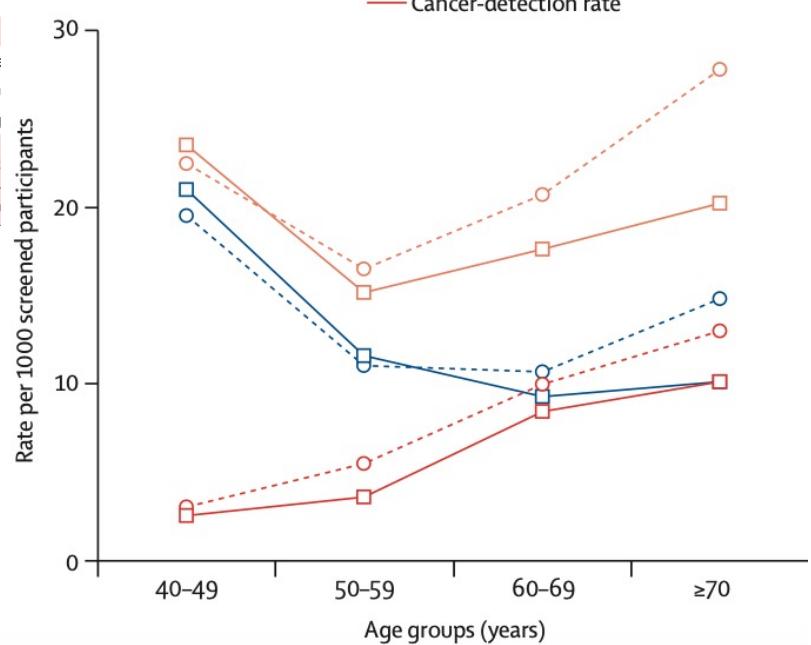


Applications of Radiology in AI: the MASAI trial

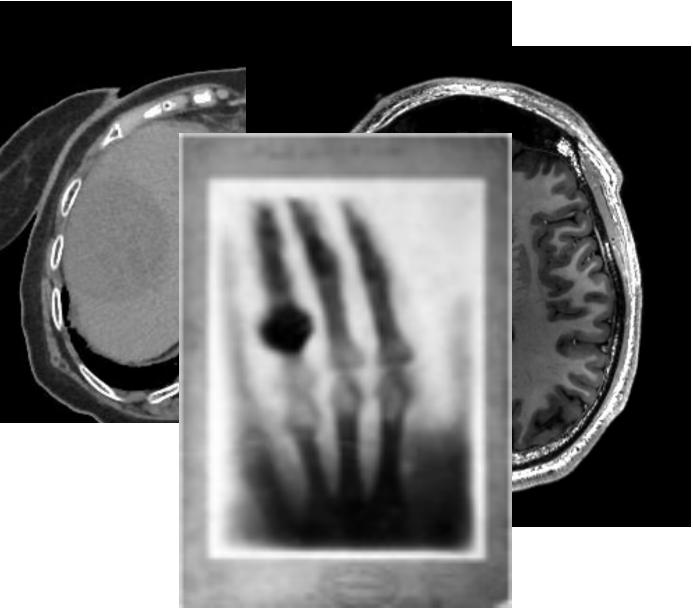


	Intervention group (n=53 043)	Control group (n=52 872)	Proportion ratio	p value
Early screening performance				
Number of recalls	1110	1027
Recall rate	2.1% (2.0-2.2)	1.9% (1.8-2.1)	1.08% (0.99-1.17)	0.084
Number of detected cancers	338	262
Cancer-detection rate, per 1000	6.4 (5.7-7.1)	5.0 (4.4-5.6)	1.29 (1.09-1.51)	0.0021
Number of false positives	772	765
False positive rate				
Positive predictive v				
Workload				
Number of screen re				
Number of consens				
Consensus meeting				
Data are n or point est				

Table 2: Early screen

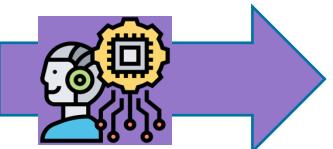
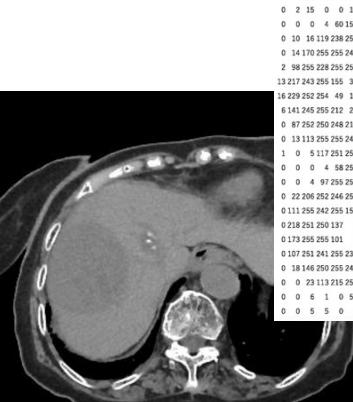


Applications of Radiology in AI: from data to insights



0	2	15	0	11	10	0	0	0	9	9	0	0
0	0	0	4	60	151	236	255	277	95	61	32	0
0	10	16	138	235	255	244	245	250	249	255	222	103
0	14	170	255	255	244	254	255	253	245	255	249	174
0	28	95	228	225	251	254	211	141	116	231	251	235
13	217	243	255	151	136	255	255	255	252	2	10	13
16	229	252	254	49	12	0	0	7	7	0	70	237
6	141	245	255	212	25	11	9	3	0	115	236	243
0	87	250	250	248	215	60	0	1	252	255	255	146
0	13	111	255	255	245	255	182	181	248	252	242	208
1	0	1	57	215	255	241	255	247	255	241	16	17
0	0	4	58	251	255	246	254	253	252	11	0	1
0	0	4	97	255	255	255	245	255	244	255	182	10
0	22	206	252	246	251	241	100	24	113	255	245	194
0	0	111	255	245	225	153	24	0	6	35	255	230
0	18	218	251	250	137	7	11	0	0	2	255	250
0	173	255	256	101	9	20	0	3	13	182	251	245
0	101	251	241	255	230	98	55	19	110	247	248	235
0	18	146	250	255	247	255	255	255	249	250	255	129
0	0	23	113	215	255	250	248	255	248	248	113	14
0	0	6	1	92	153	233	255	252	147	37	0	1
0	5	0	0	0	0	0	0	0	14	1	0	6

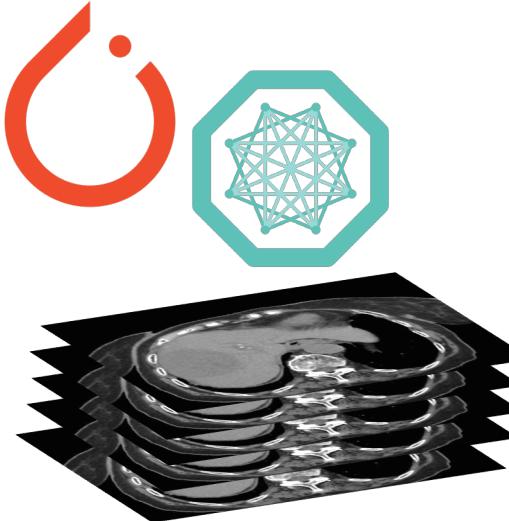
0	2	15	0	0	11	10	0	0	0	9	9	0	0
0	0	0	4	60	157	235	255	255	177	95	61	32	0
0	10	16	119	238	255	244	245	245	250	250	255	222	103
0	14	170	255	255	244	254	255	253	245	255	249	253	124
0	13	217	243	255	155	133	255	252	0	10	12	235	255
16	229	252	254	49	12	0	0	7	7	0	70	237	252
6	141	245	255	212	25	11	9	3	0	115	236	243	255
0	87	252	250	248	215	0	1	12	252	255	248	144	0
0	13	113	255	255	245	255	182	181	248	252	242	208	36
1	0	5	117	251	254	241	255	247	255	241	162	17	0
0	0	0	4	58	251	255	254	256	253	255	255	220	11
0	0	4	97	255	255	255	254	252	255	254	245	182	10
0	22	206	252	254	251	241	100	24	113	255	245	255	194
0	111	255	242	255	158	24	0	6	39	255	230	230	56
0	218	251	250	137	77	11	0	0	2	62	255	250	120
0	173	255	255	101	9	20	0	13	3	13	251	251	61
0	107	251	241	255	230	98	55	19	118	217	248	253	52
0	18	146	250	255	247	255	255	255	255	249	255	255	129
0	0	23	113	215	255	250	248	255	255	248	248	118	14
0	0	6	1	0	52	153	233	255	252	147	37	0	0
0	0	5	5	0	0	0	0	14	1	0	6	6	0



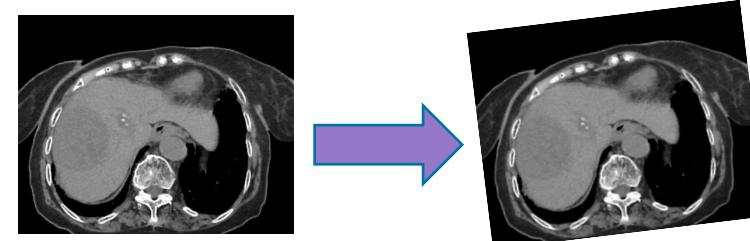
OpenCV RADIOMICS



Applications of Radiology in AI: how is data used?



**Radiology dataset
(2D, 3D...)**



Data Loading

- Package to load images
- Functions to feed data to the model
- Augmentations on images

Model training

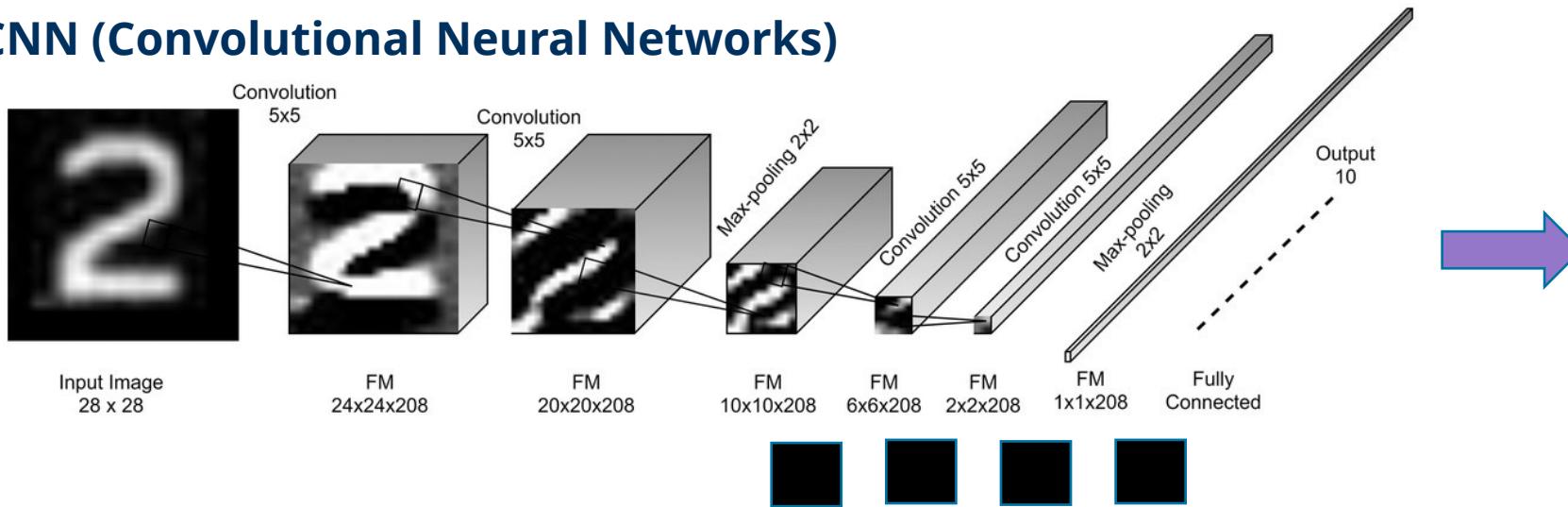
- Select parameters for training
- Choose architecture
- Make loops for training and tracking

Test & evaluate

- Use different data to deploy model
- Test model through different metrics
- Explainability

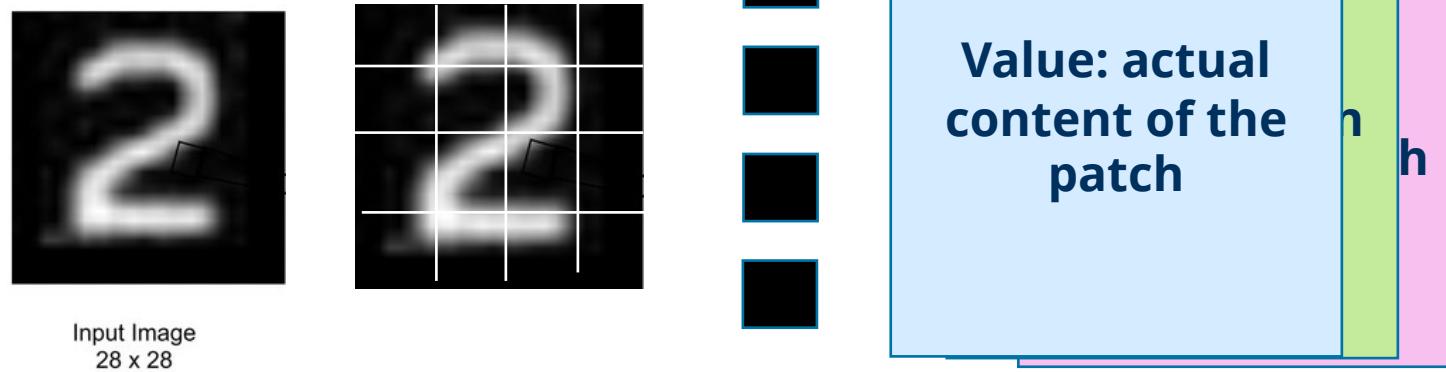
Applications of Radiology in AI: architectures

CNN (Convolutional Neural Networks)



Features based on relationships of **neighbouring** pixels through convolution

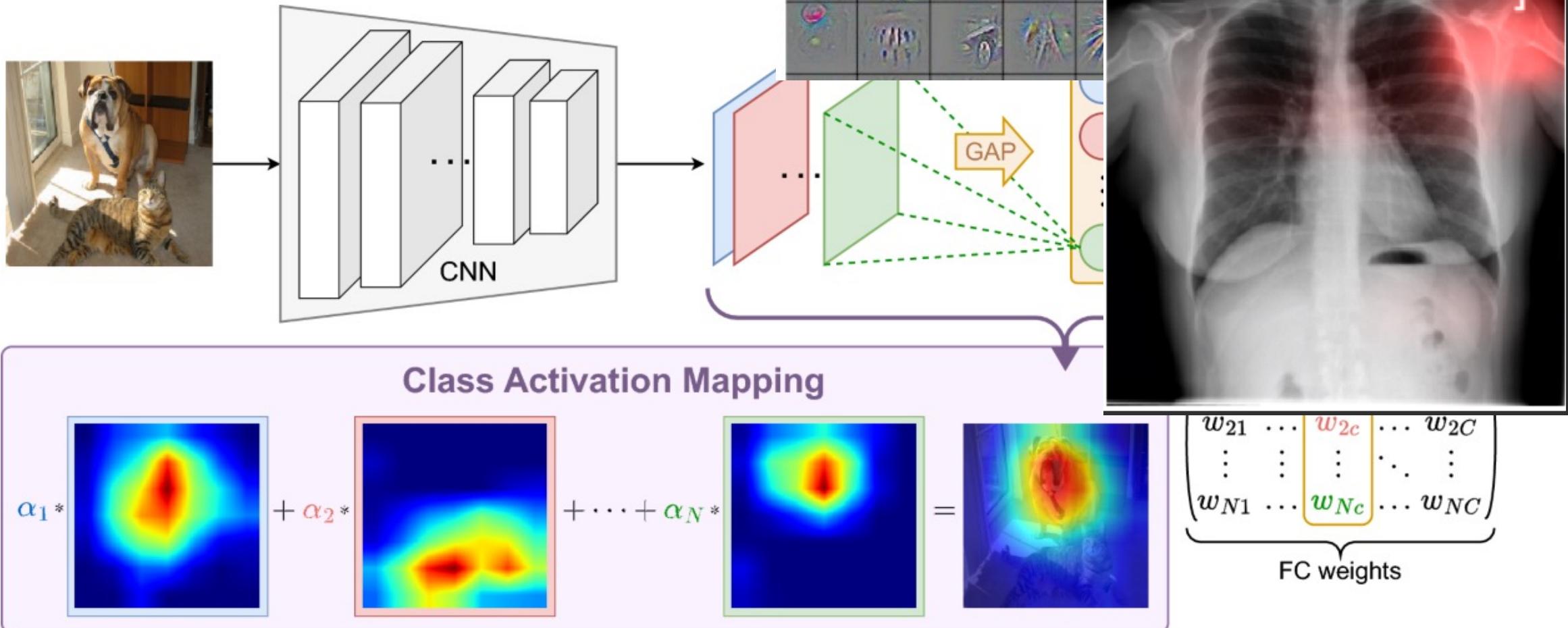
(Vision) Transformers



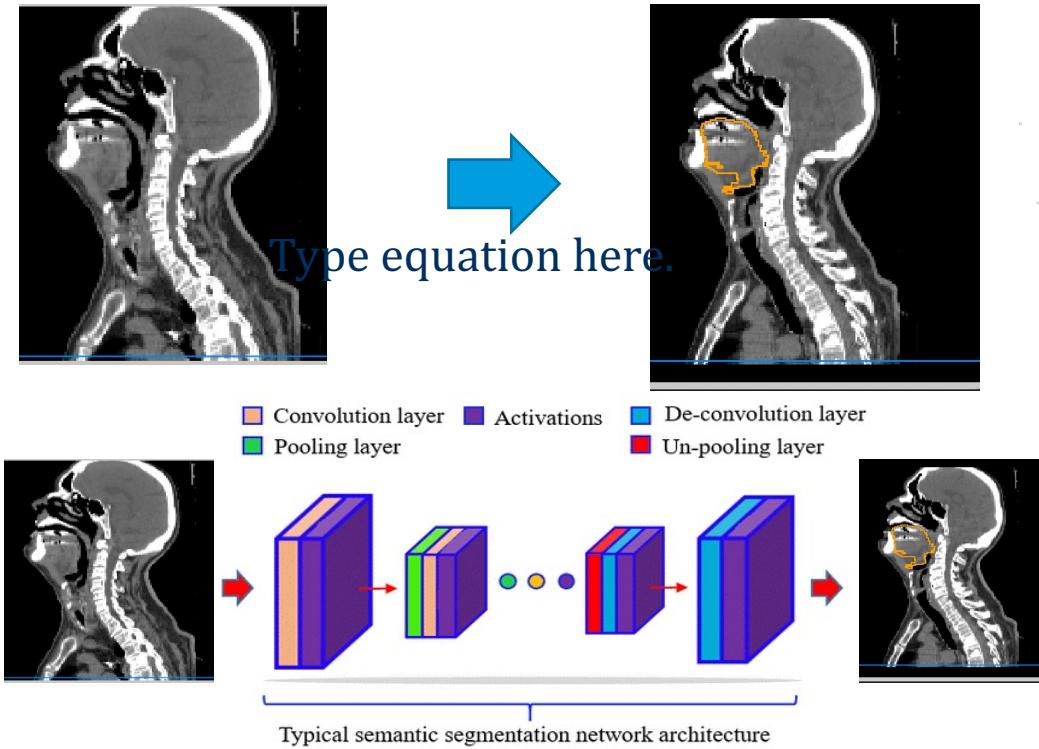
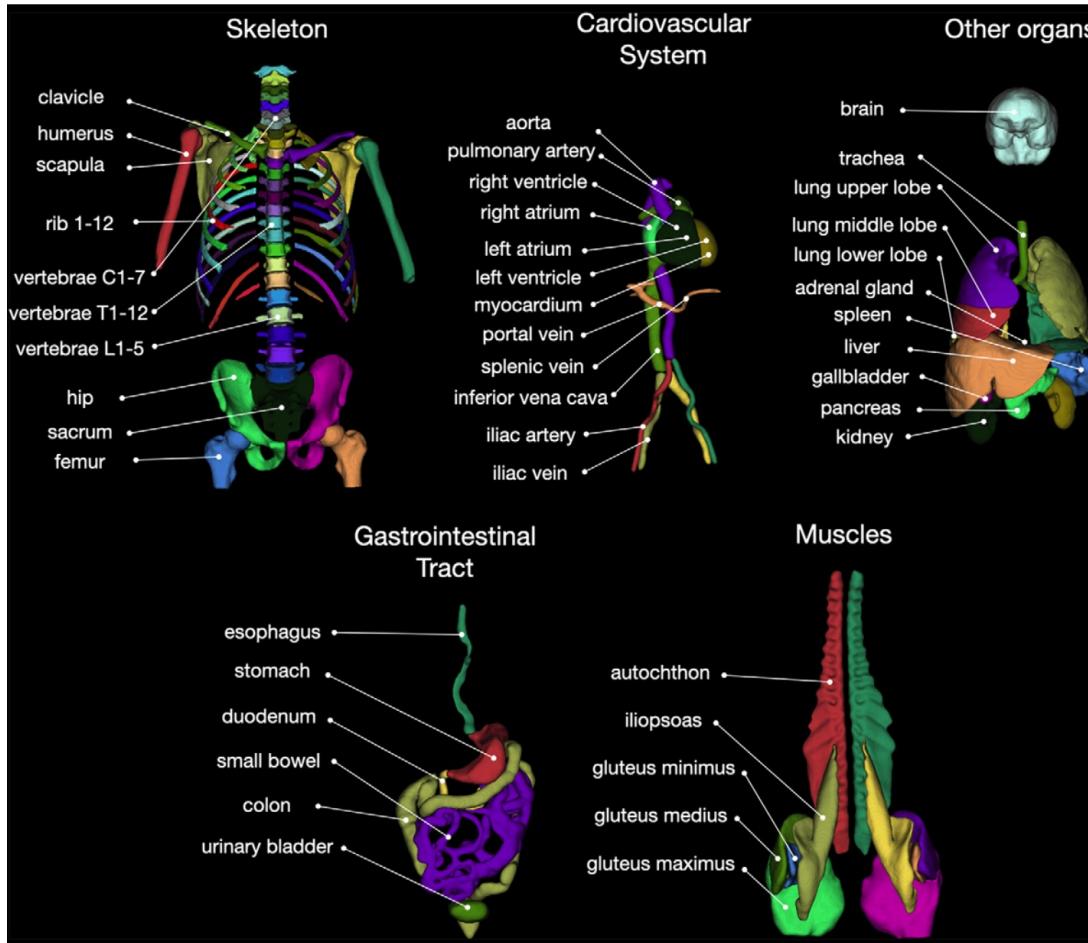
Features based on learned relationships between tokens, **very weak inductive bias**

Applications of Radiology in AI: explainability

CAM and Grad-CAM



Applications of Radiology in AI: segmentation



Segmentation Loss (Dice)

$$L_{dice} = 1 - \frac{2 \sum_{n=1}^N t_n y_n}{\sum_{n=1}^N (t_n + y_n)}$$

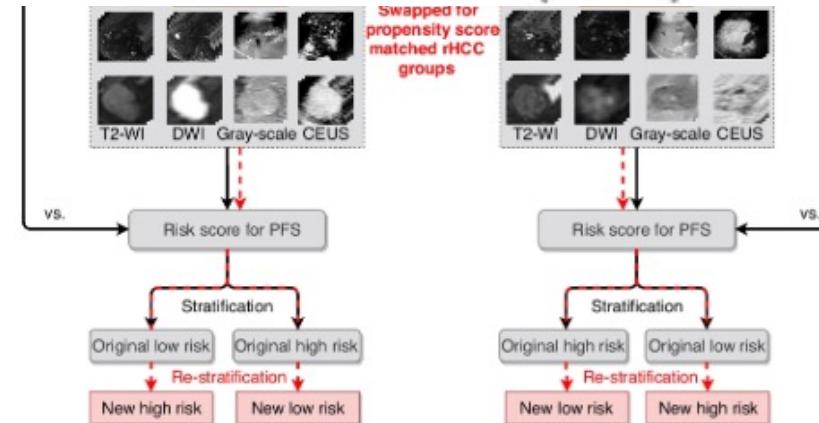
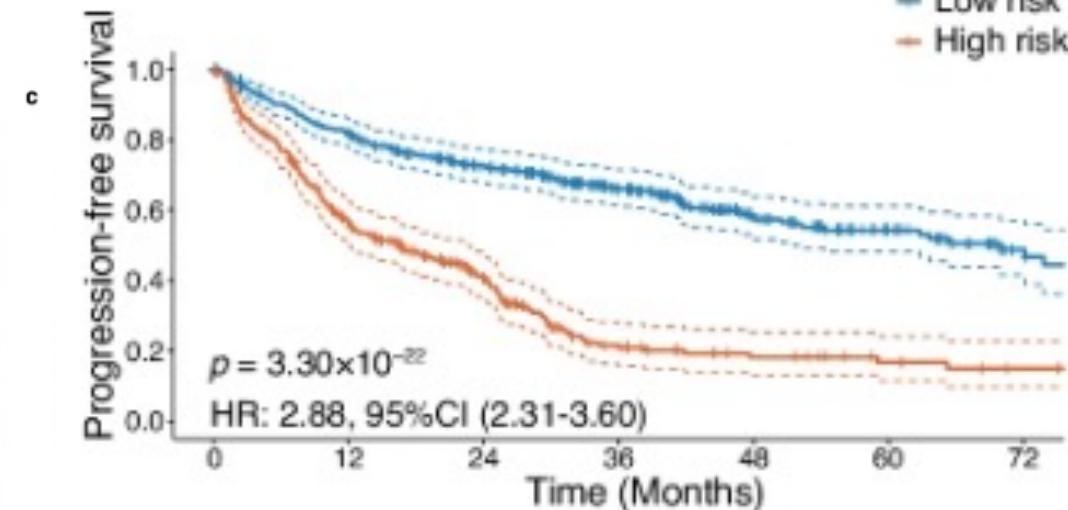
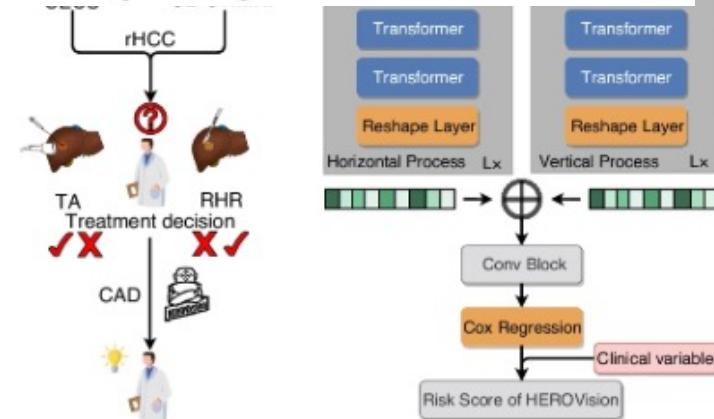
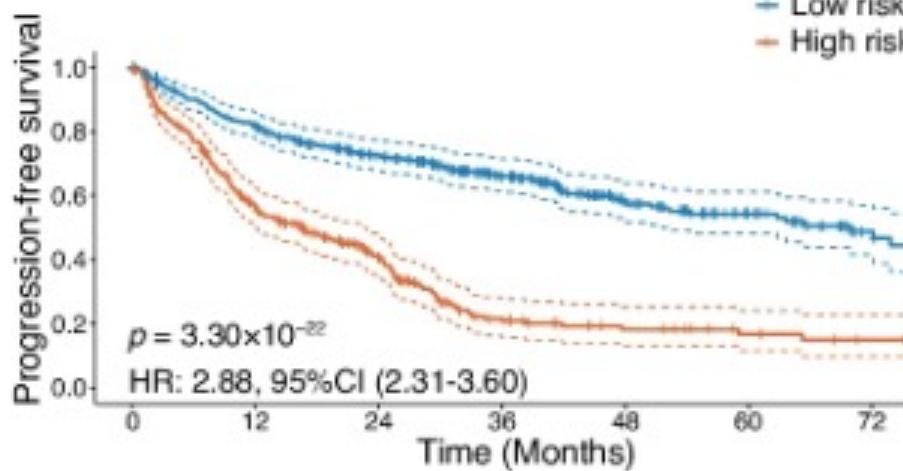
Segmentation tool and image of body:
Wasserthal., J. et al. *Radiology AI* (2023). 5(5)
H&N images: Welch, M et al. (2023).
Computed Tomography Images from Large Head and Neck Cohort (RADCURE) (Version 4) The Cancer Imaging Archive. https://doi.org/10.7937/J47W-NM11

Slide 23

Applications of Radiology in AI: prognosis

Article | [Open access](#) | Published: 01 May 2025

Vision transformer-based model can optimize curative-intent treatment for patients with recurrent



Conclusions

- Radiology is a long-lived field with an early adoption of electronic technology
- Radiological images are varied and diverse, making Radiology a complex field.
- The complexity, number and link of radiological images with diseases makes them a rich source f informative data for modeling
- Radiology can benefit from AI, reducing workloads and enhancing radiologist readings
- Ai in Radiology is an ongoing field of research. There is evolution in architectures, training styles and applications

Now on to the hands-on session!

