

Machine Learning in Clinical Reality

Georg Langs

Medical University of Vienna
Department for Biomedical Imaging and Image-guided Therapy
Computational Imaging Research Lab
CAIM - Comprehensive Center for AI in Medicine

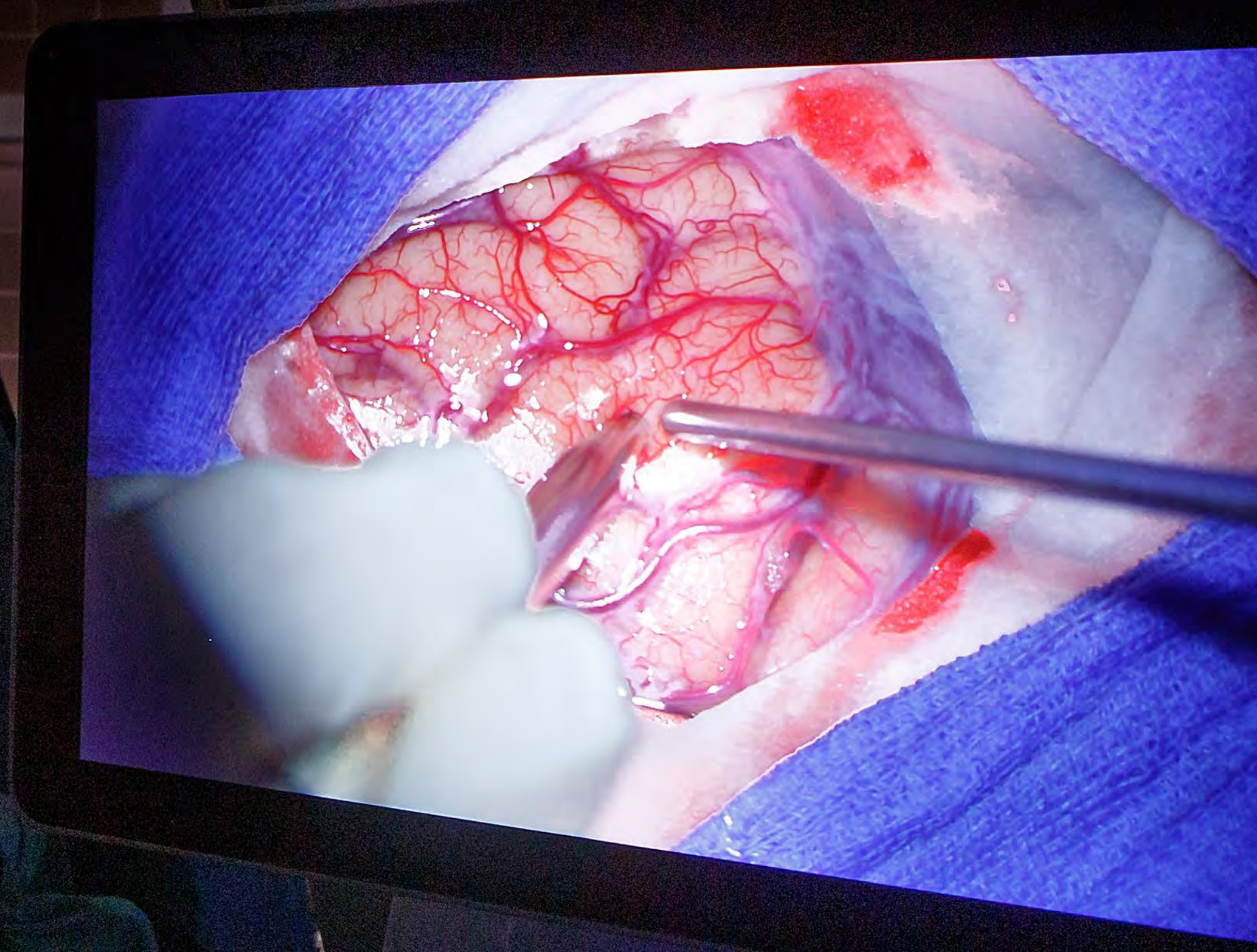
Disclosure

Co-founder, shareholder: contextflow GmbH

Research grants: Siemens Healthineers, Novartis, Boehringer Ingelheim, NVIDIA

Honoraria: Roche

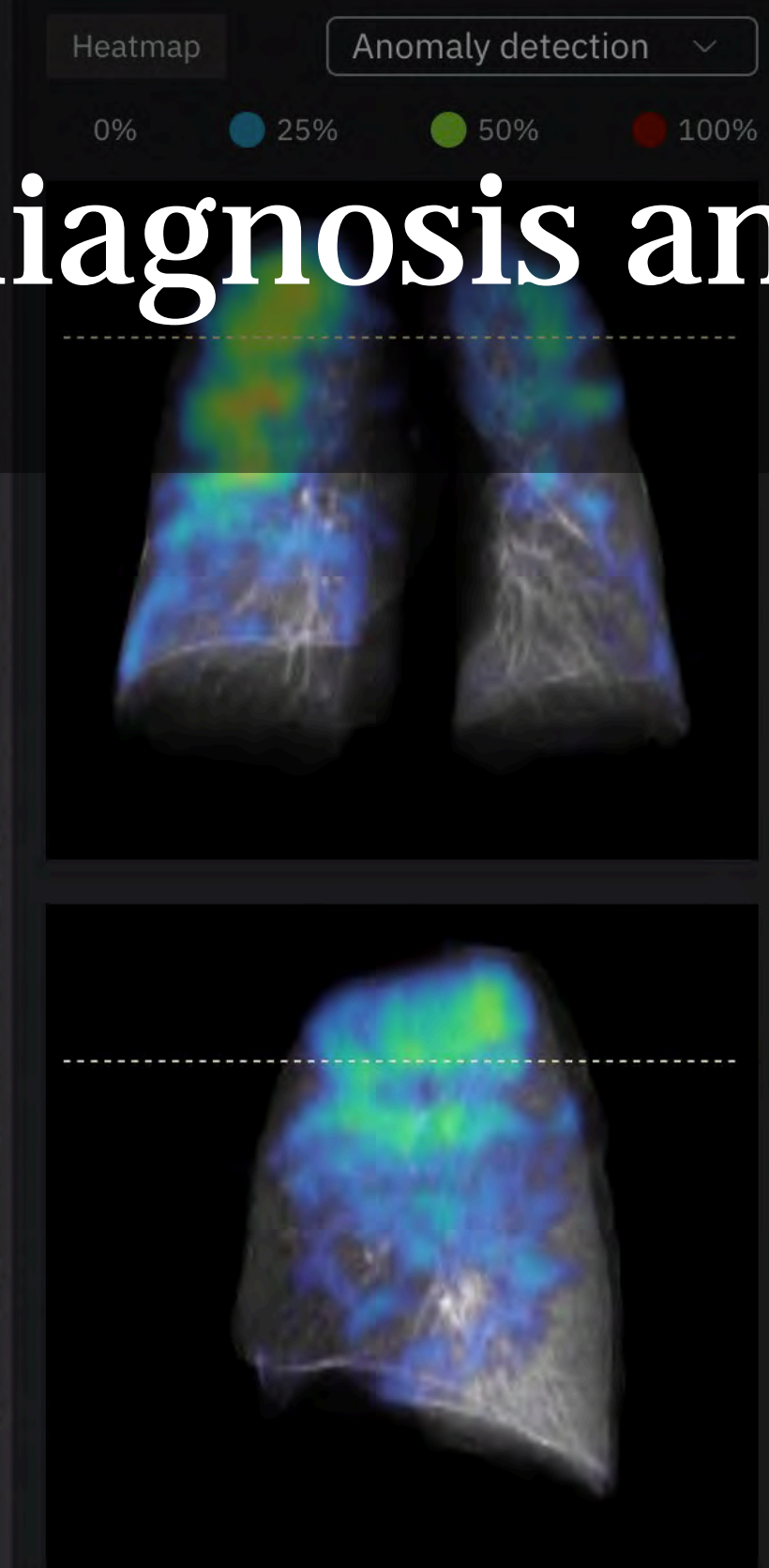
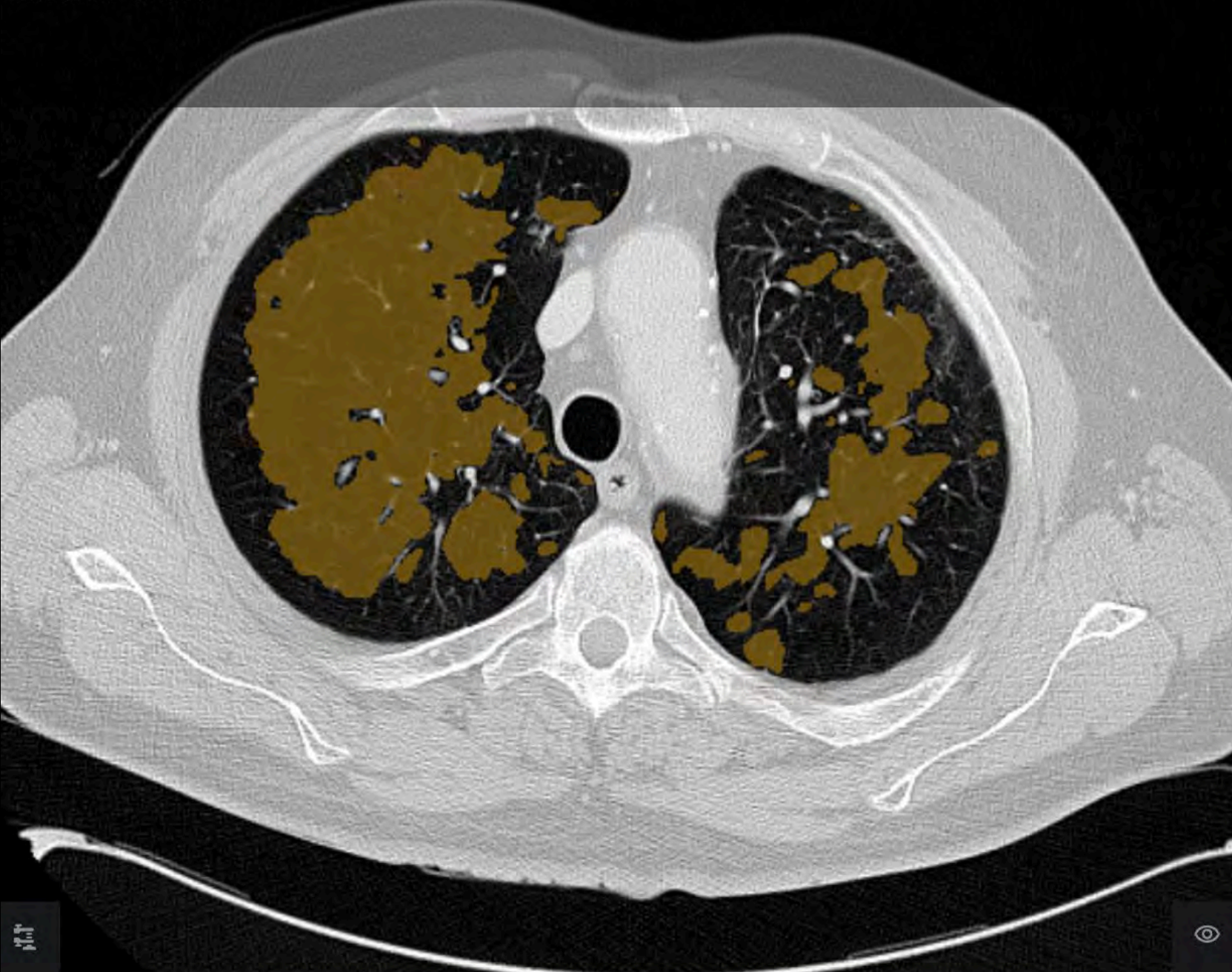
Rapid diagnosis during neuro surgery



Georg Widhalm et al. Neuro Surgery MedUni Wien

Computer aided diagnosis and quantification in radiology

Series description:
 Name: Patient-4593
 Age: 0 years
 Sex: M
 Slice: 37 of 156



Lung Tissue Analysis Results

Consolidation Effusion Emphysema

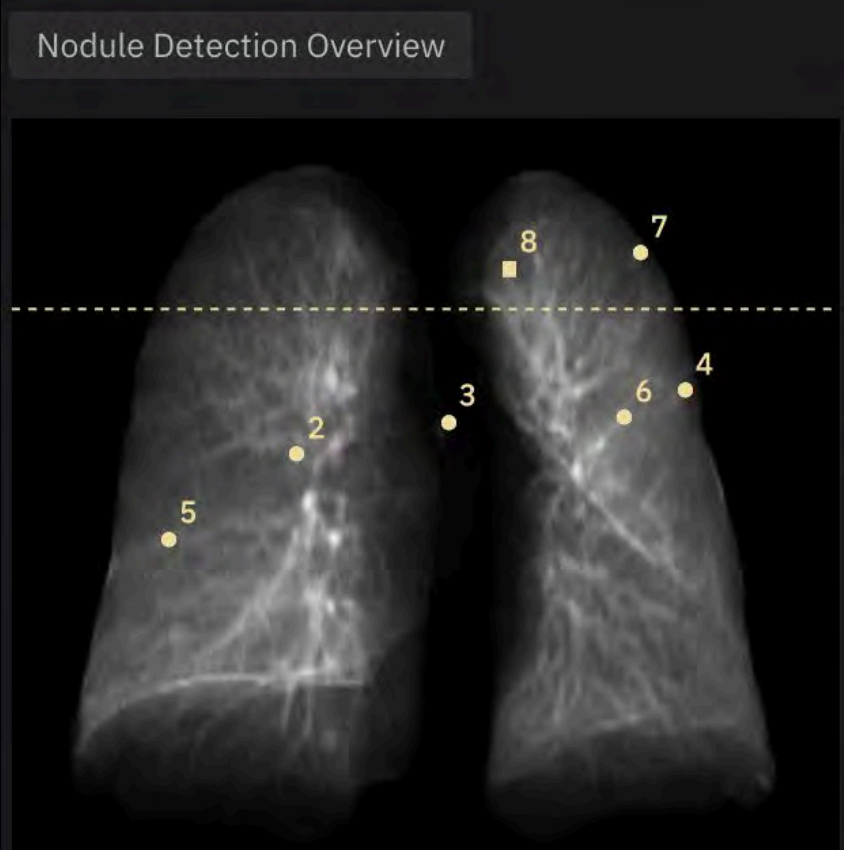
Ground-glass o... Honeycombing Pneumothorax

Reticular pattern Other

21% Lung anomalies Unremarkable 79%

*% coverage values denote the percentage of total lung volume affected

Category	Coverage	Volume
Total Potential Lung Volume	100%	6.4L
Lung mask		
Pleural Cavity		
Effusion	0%	0.0L
Pneumothorax	0%	0.0L
Lung Parenchyma	100%	6.4L
Consolidation	0%	0.0L
Emphysema	16%	1.0L
Ground-glass opacity	2%	0.2L
Honeycombing	< 1%	0.0L
Reticular pattern	< 1%	0.1L
Other	2%	0.1L
Unremarkable	79%	5.0L



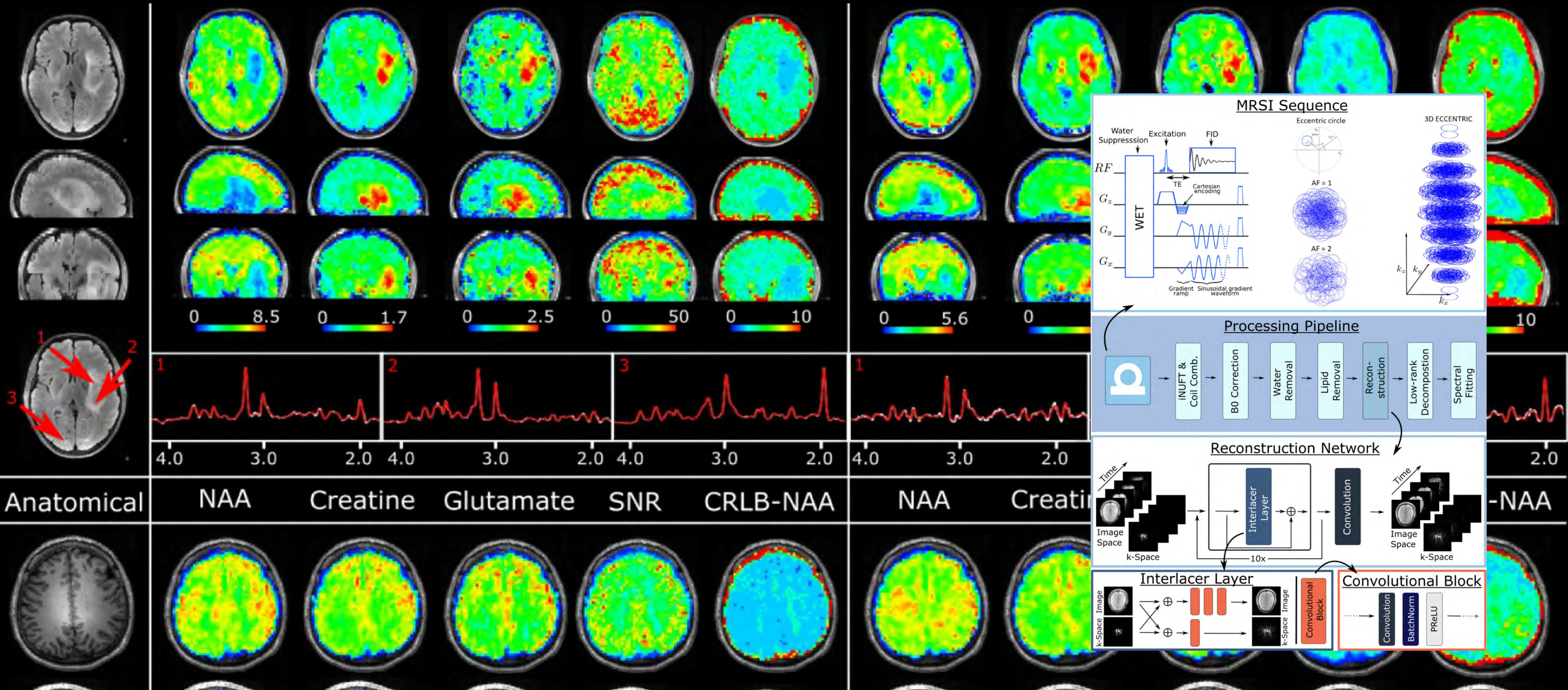
Cumulative volume	1581mm ³
Largest nodule	432mm ³
Detected nodules	7
● Solid (nodules)	6
■ Part-solid (nodules)	1
▲ Non-solid (nodules)	0

1-5/7 Guideline Lung-RADS Sort by Ø average

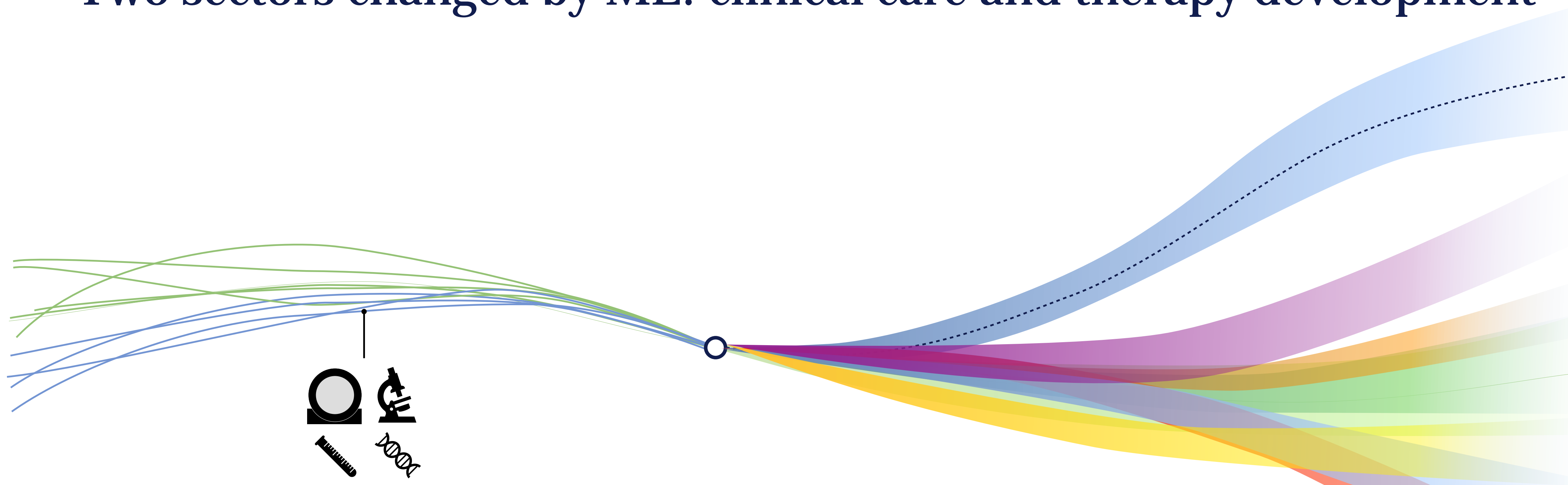
Nodule ID	Volume	Ø A	Ø L	Ø S
1	432mm ³	9mm	10mm	8mm
2	401mm ³	8mm	9mm	7mm
3	34mm ³	7mm	7mm	6mm
4	34mm ³	6mm	9mm	4mm
5	34mm ³	6mm	7mm	5mm

Röhrich et al. EURA, Dep. Radiology MedUni Wien

Reconstructing MR spectroscopy imaging with DL



Two sectors changed by ML: clinical care and therapy development



A. ML to treat individual patients

- Predict disease course
- Predict response to treatment
- Optimal patient management
- Early intervention and prevention



B. ML to develop novel therapies

- Novel fine-grained early digital endpoints
- Patient stratification
- Biology, hypotheses, and therapeutic targets
- Identify phenotypes in real world populations



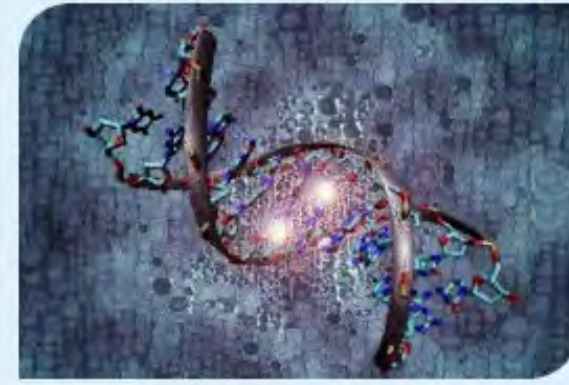


**COMPREHENSIVE CENTER
FOR ARTIFICIAL INTELLIGENCE IN MEDICINE
MEDICAL UNIVERSITY OF VIENNA**



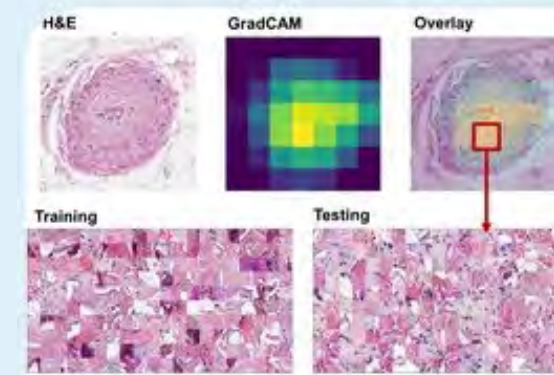
**Dynamics of
neural systems**

Adam Gosztolai



**Machine learning
in biomedicine**

Christoph Bock



**Dermatologic
imaging and
informatics**

Philipp Tschandl
Harald Kittler



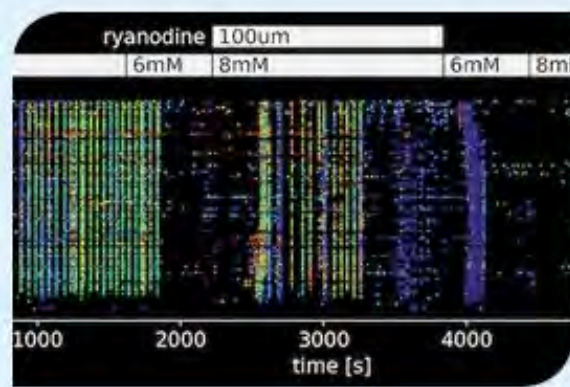
**Speech and
hearing science**

Philipp Aichinger



**Clinical pathways
and health
services**

Dominik Roth



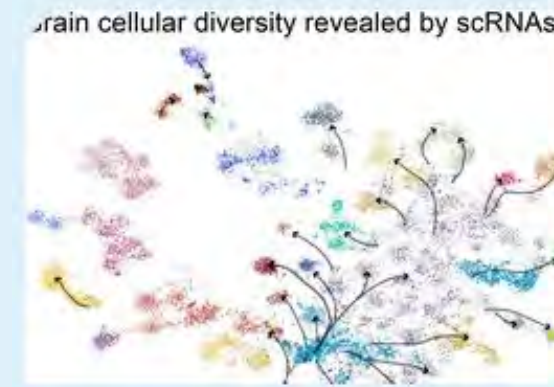
**Cell and tissue
networks**

Marjan Slak Rupnik



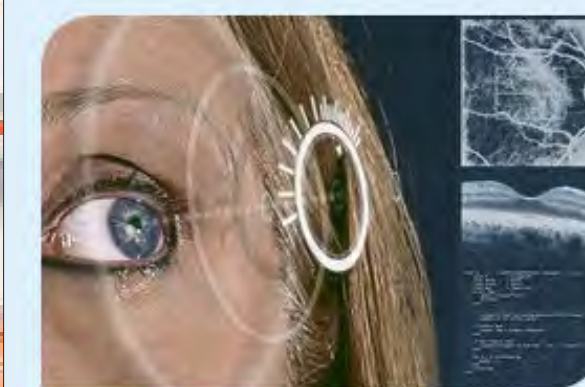
**Applied quantum
computing**

Laszlo Papp



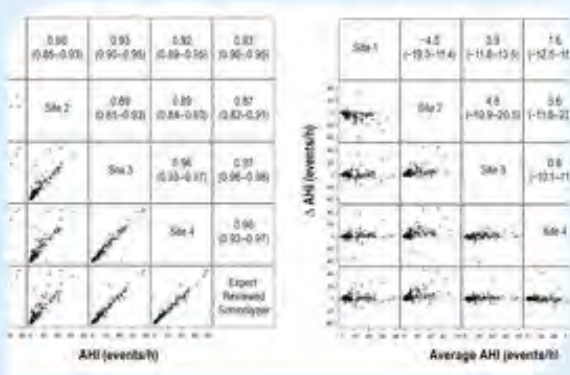
**Brain cellular
heterogeneity**

Roman Romanov



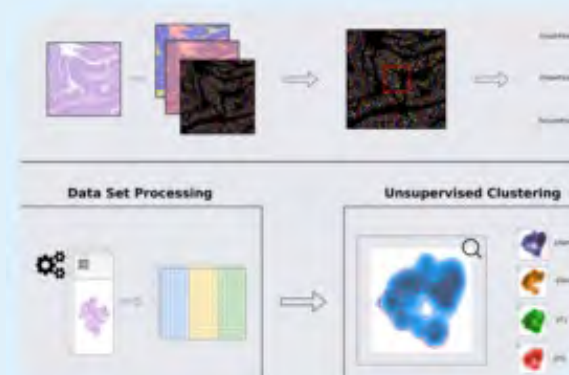
**Ophthalmic image
analysis**

Hrvoje Bogunovic



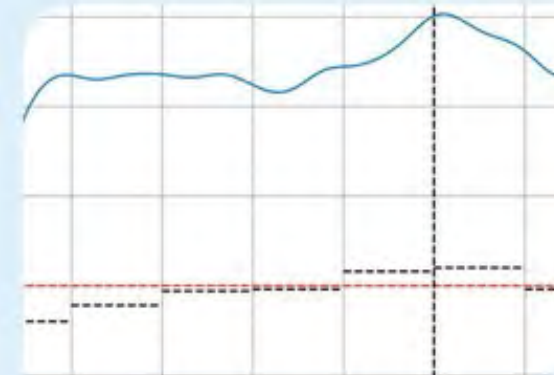
**ML for prediction
and signal
processing**

Georg Dorffner



**Computational
pathology**

Christopher Kaltenecker



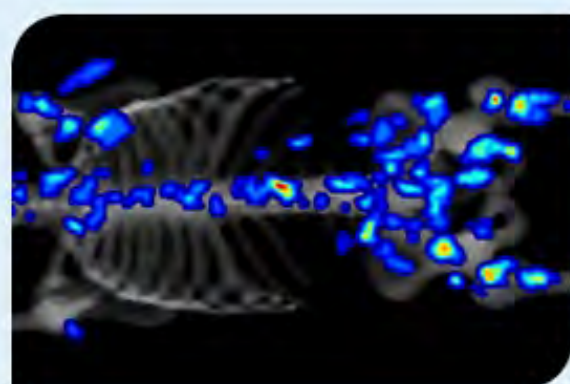
**Intensive care
and anesthesia**

Oliver Kimberger



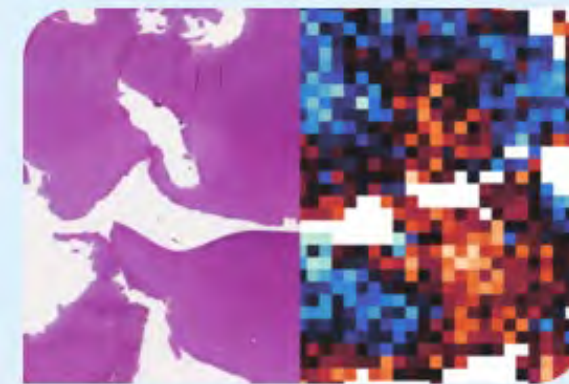
**Trustworthy
artificial
intelligence**

Matthias Samwald



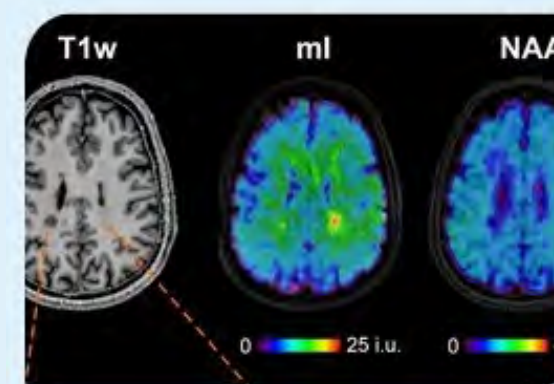
**Computational
imaging research**

Georg Langs



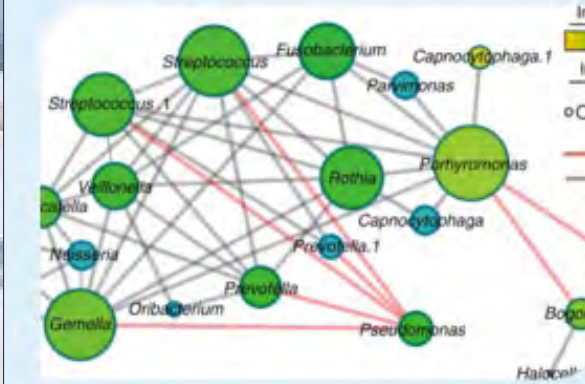
**Computational
neuropathology**

Adelheid Wöhrer



**Metabolic MR
imaging**

Wolfgang Bogner



**Multi-omics
microbial systems**

Stephanie Widder

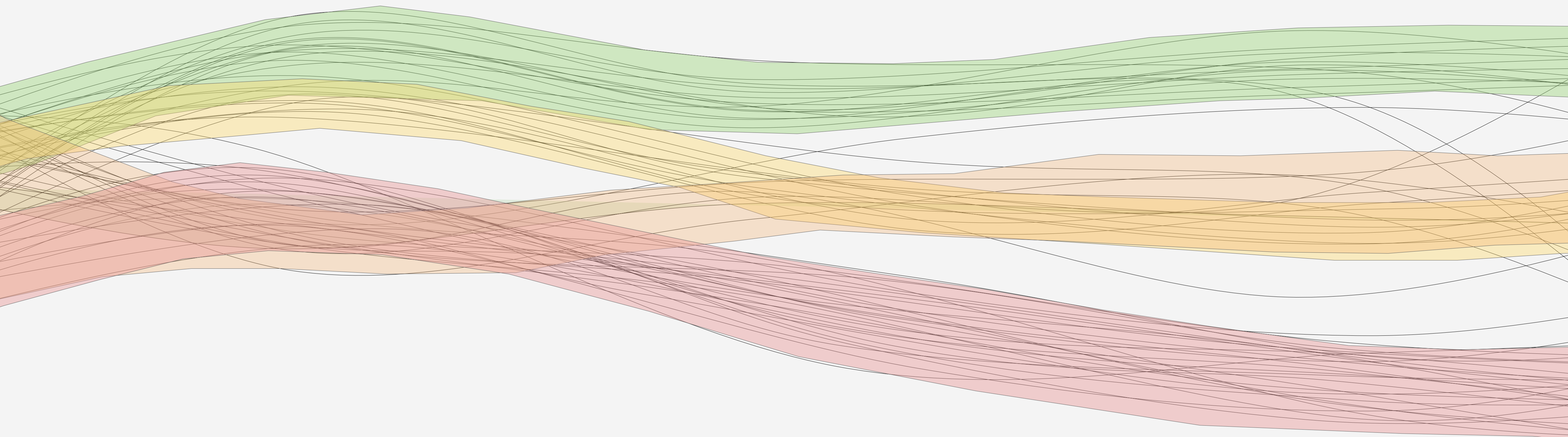




- **Integration:** clinical real world use & biomedical research & machine learning development
- **Recruiting:** attractive center with 25 labs, 150 researchers across all disciplines
- ***Dicke Bretter* are still open:** robustness, explainability, learn from few patients, biological mechanisms

Agenda

- **Expanding role of AI** in real world data use
- **Adoption in the clinic** and value propositions of AI
 - Efficiency
 - Quality of diagnosis
 - Downstream impact on patient pathways
- **Agentic AI in healthcare** - robustness and deeper integration
- **Implementing AI** in the live clinical environment



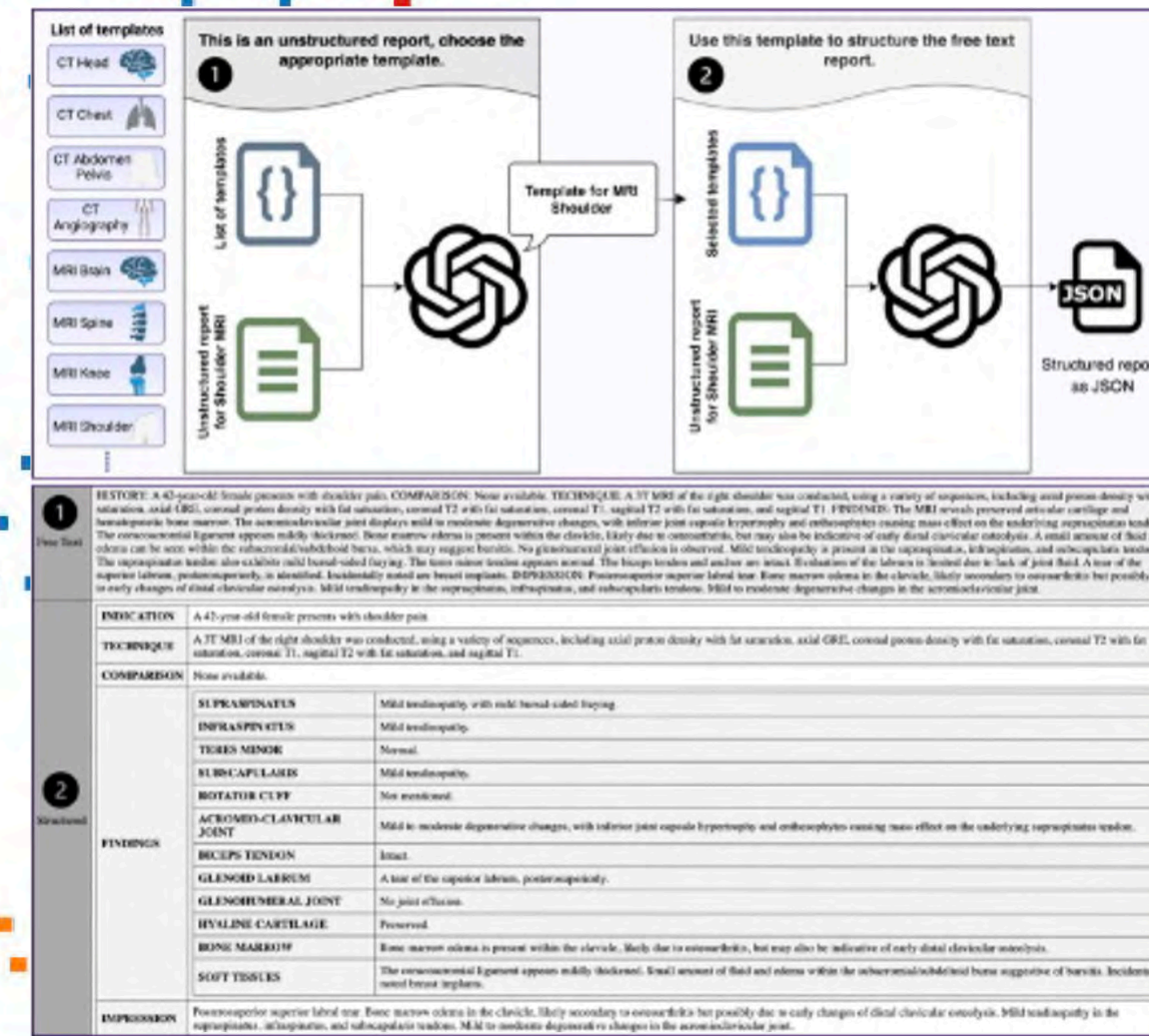
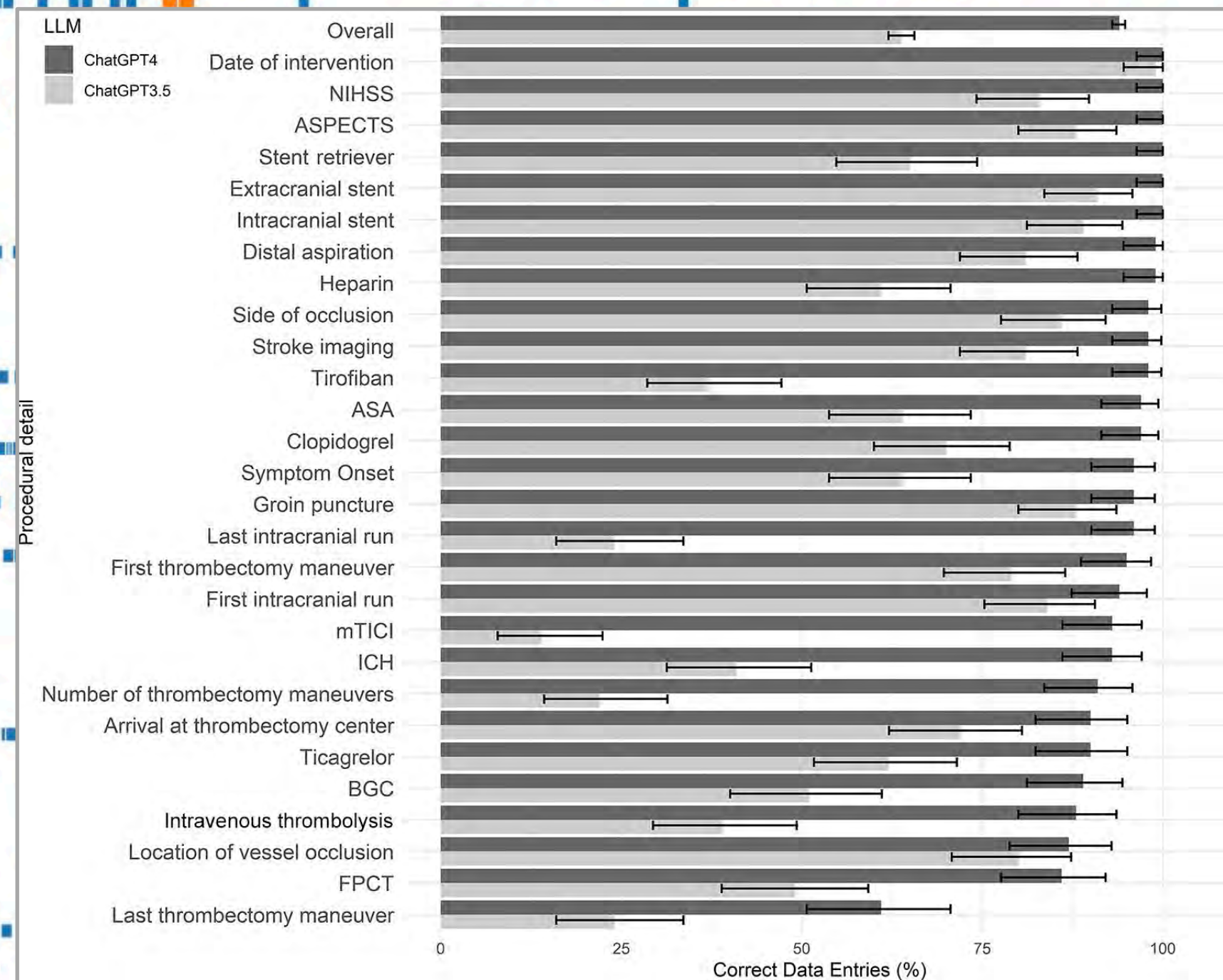
Learning from real life clinical data

Scaling to 100k patient evidence, capturing diversity and heterogeneity

Large scale real life data

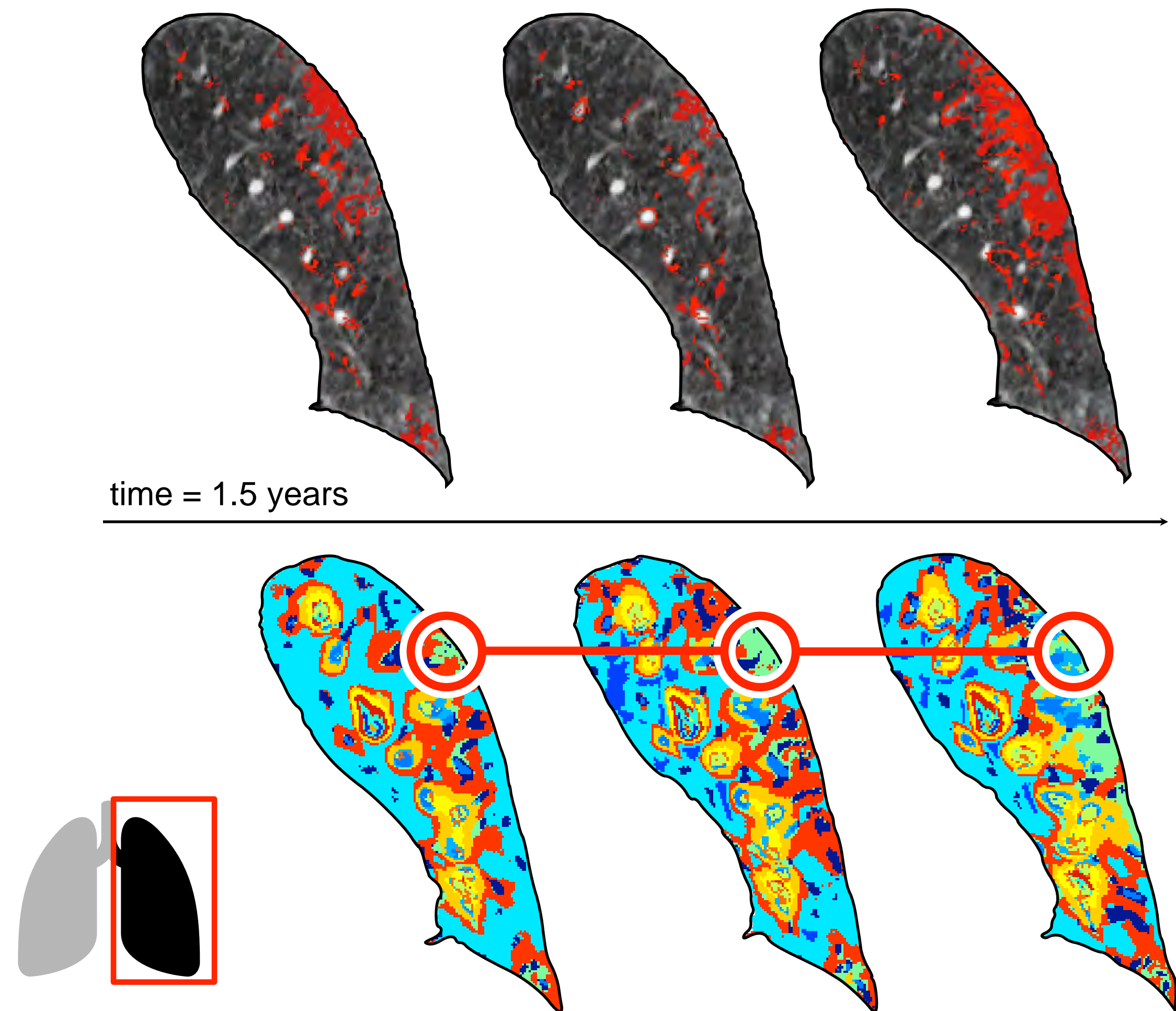
Scale ~100k patient trajectories
Ability to process unstructured data

event
■ LAB
■ CT
■ OP
■ DEATH



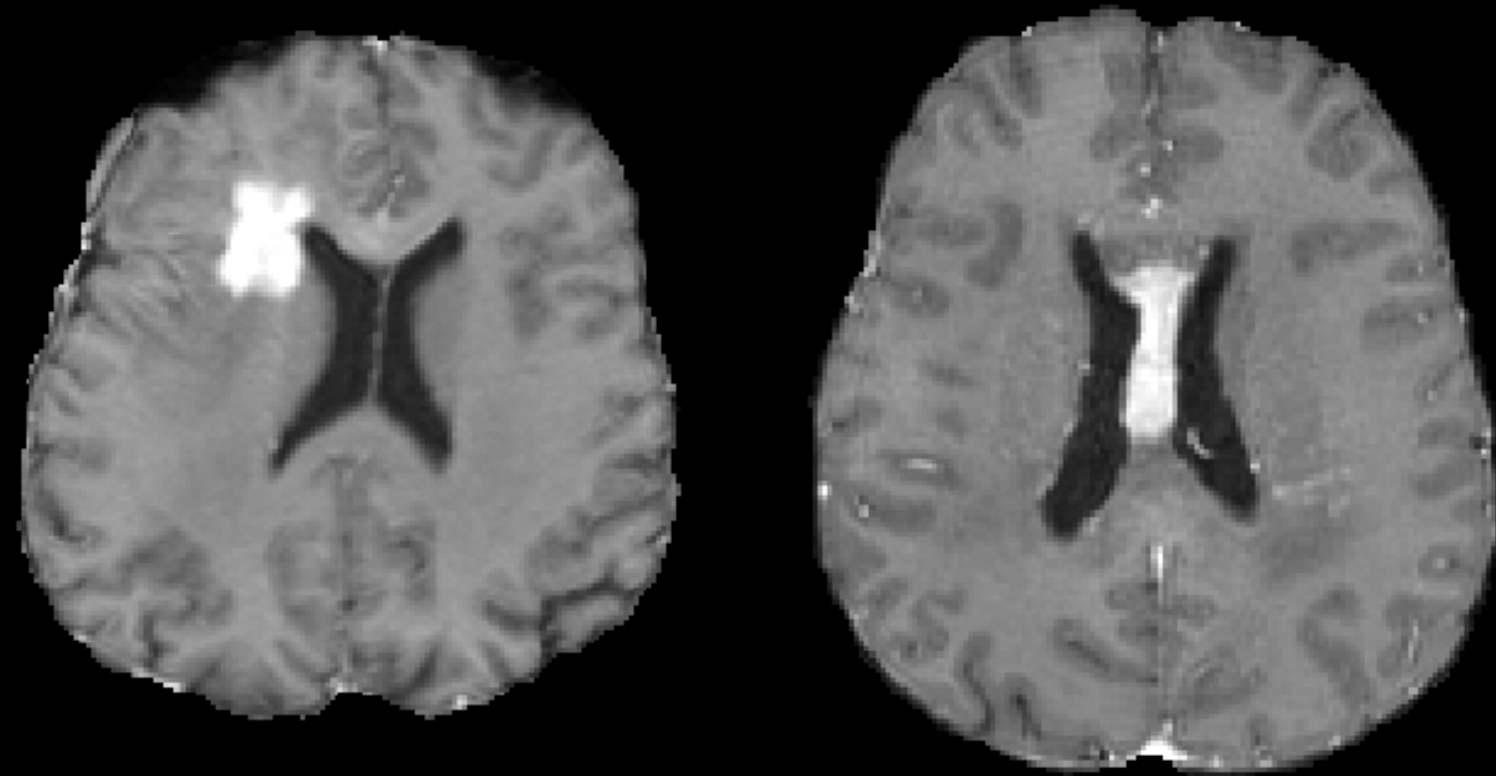
Lehnen et al. Radiology 2024, Adams Radiology 2023, Rahsepar Radiology 2023

Discover transition pathways in interstitial lung disease

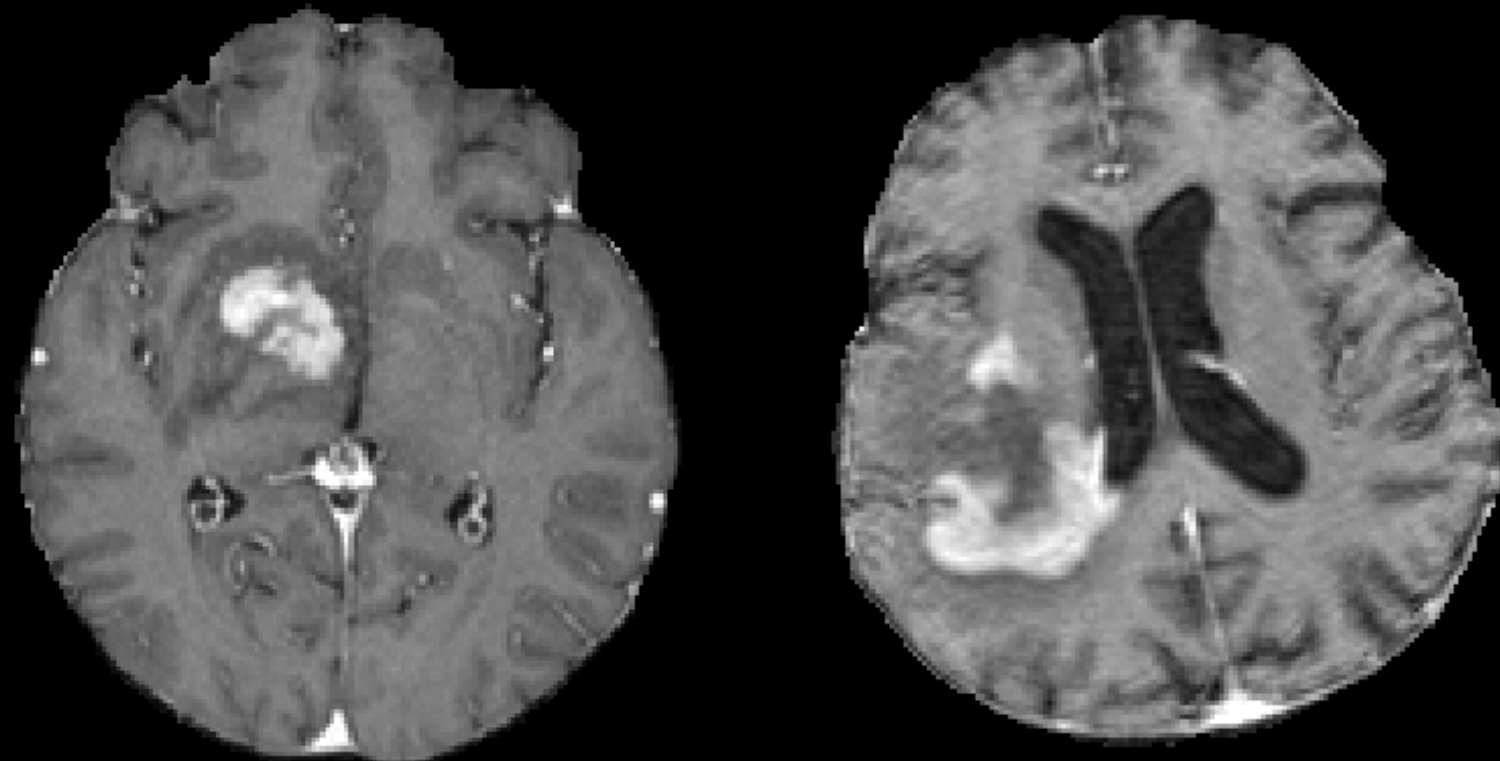


Vogl, Prosch, Ofner et al. MICCAI 2014, Pan et al. 2022 European Radiology

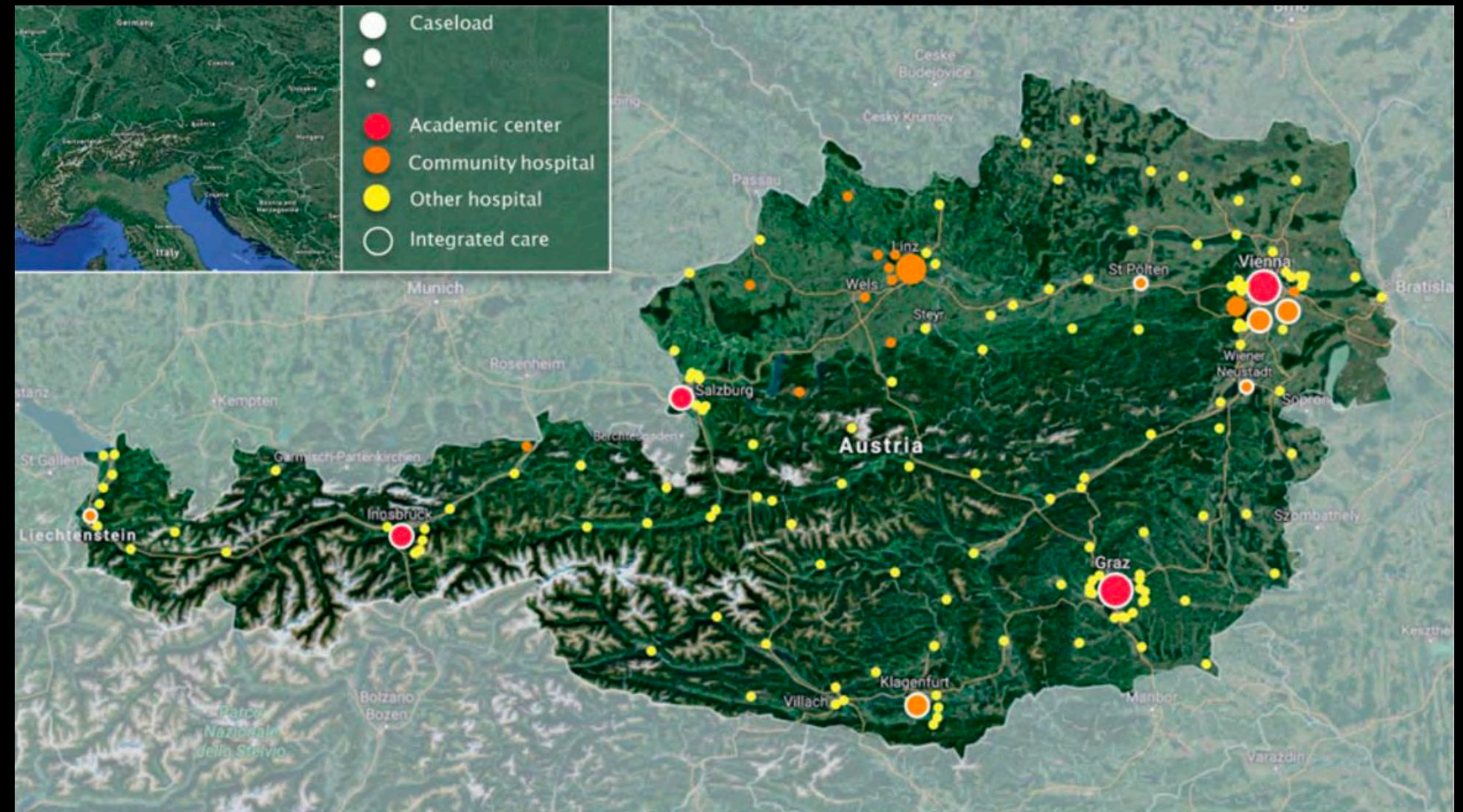
From bedside to bench: linking imaging and epigenomics



low risk: median survival 53.23 months



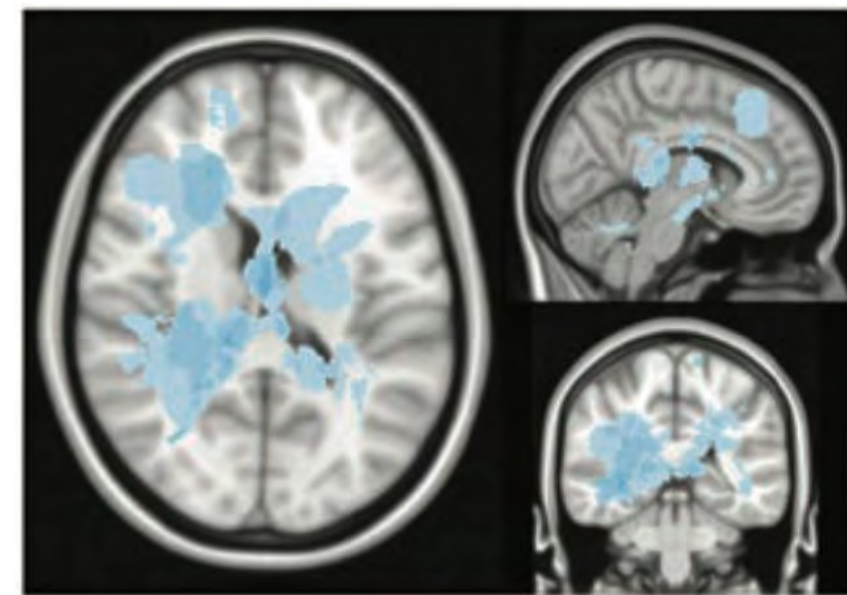
high risk: median survival 3,30 months



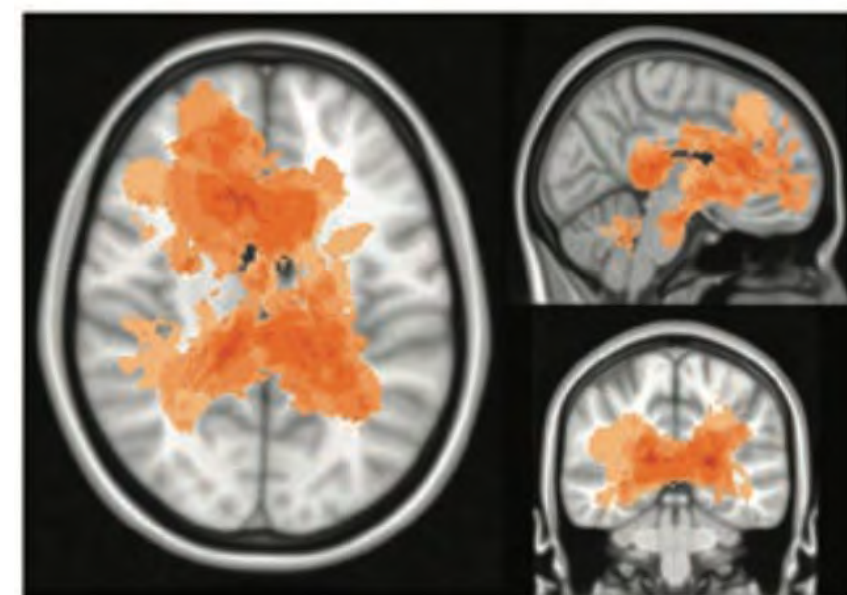
10 centers - 164 patients - ABTR Primary CNS Lymphoma Study
Validation: 1 center - South Korea

Nenning et al. 2023 *Neuro. Onc. Adv.* doi.org/10.1093/noajnl/vdad136

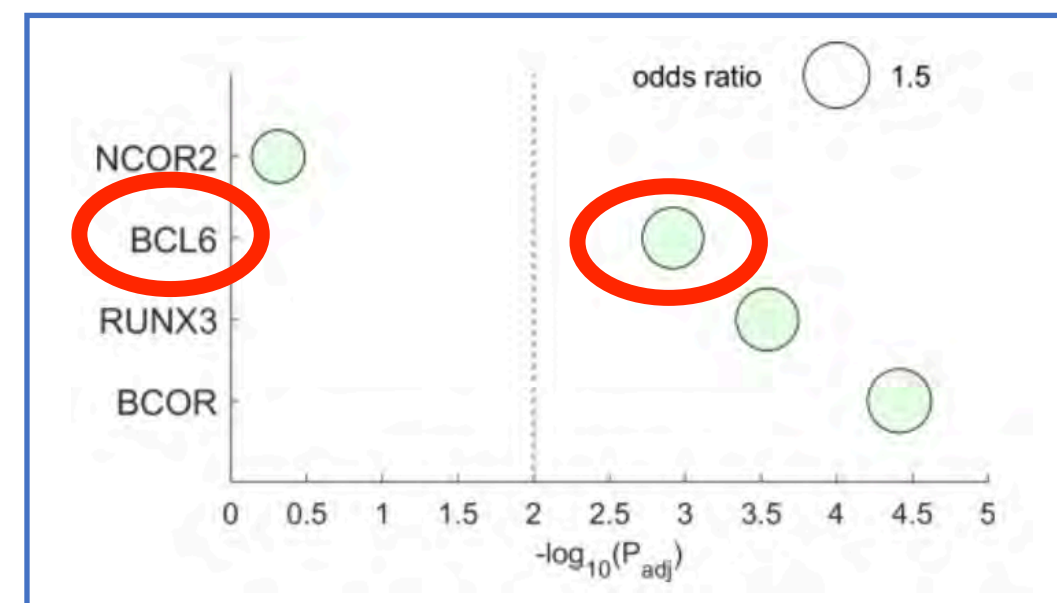
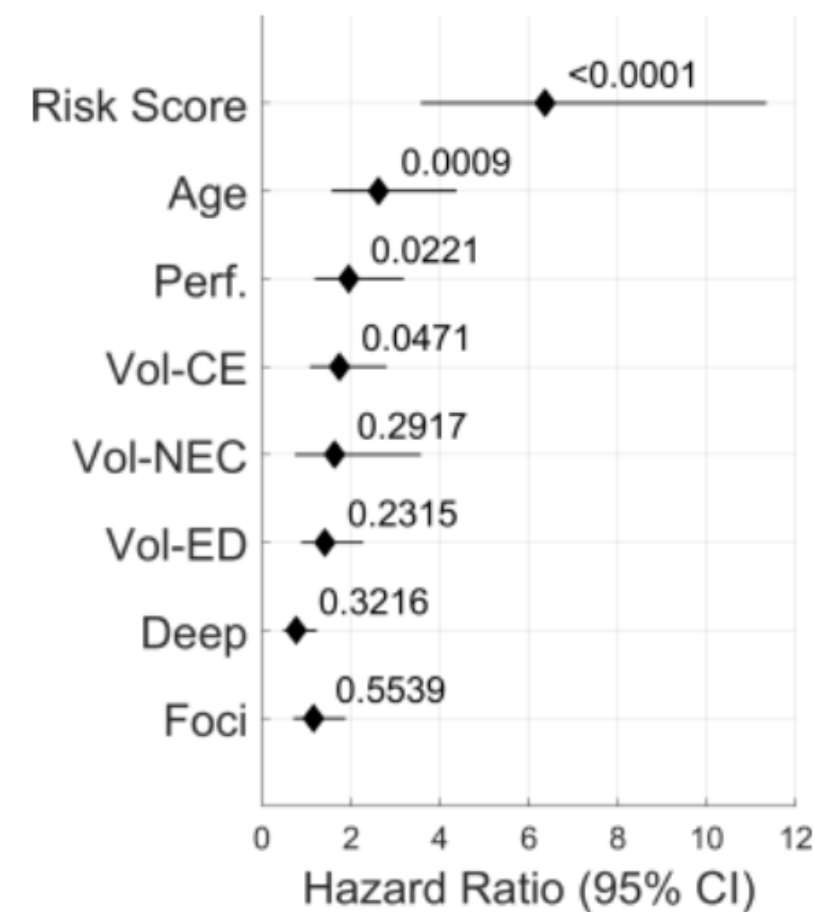
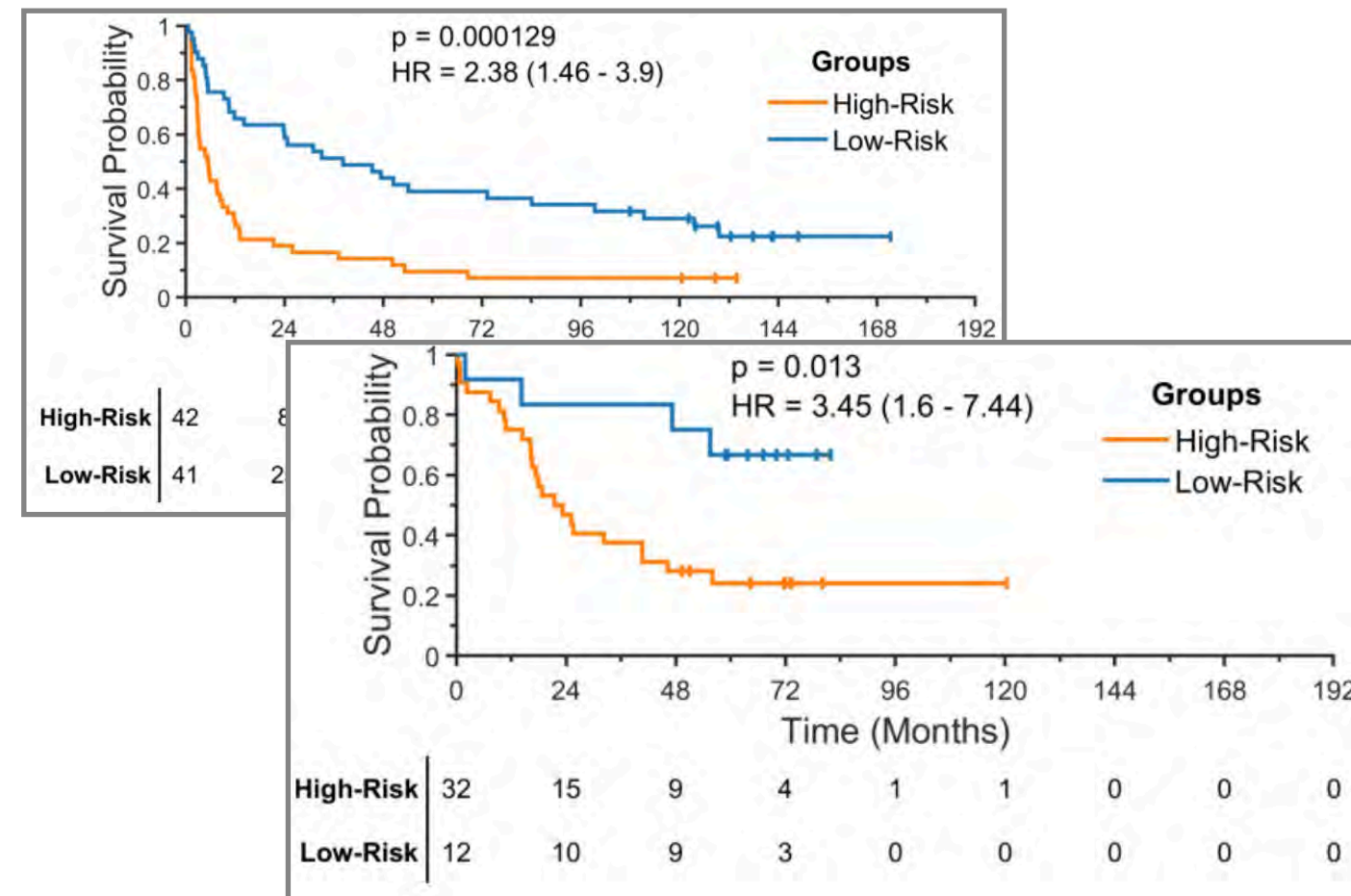
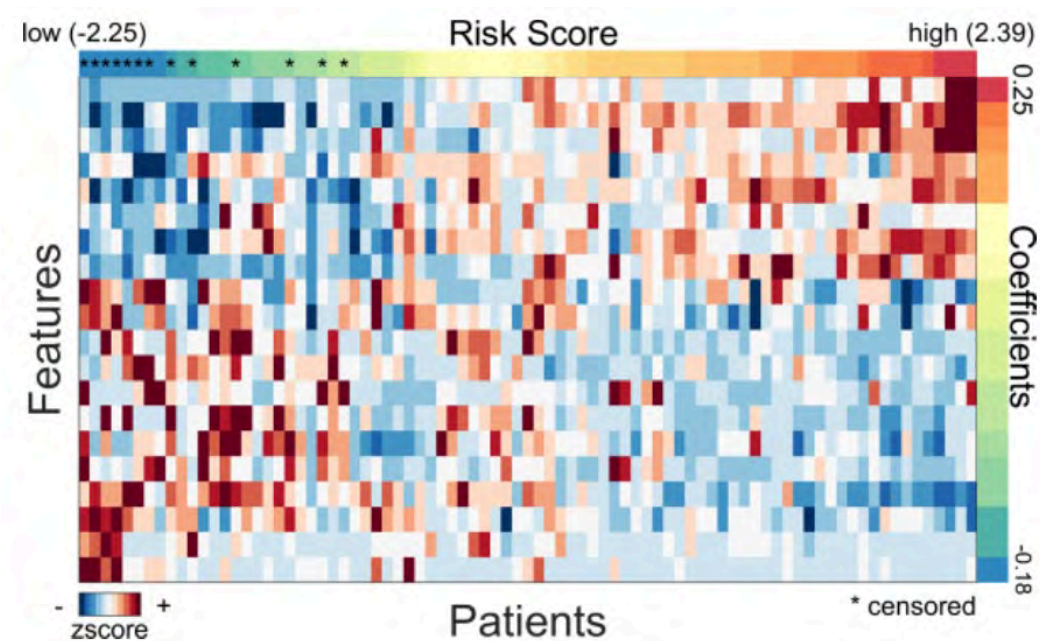
From prediction to target identification



Overlap (# low-risk) 0 10



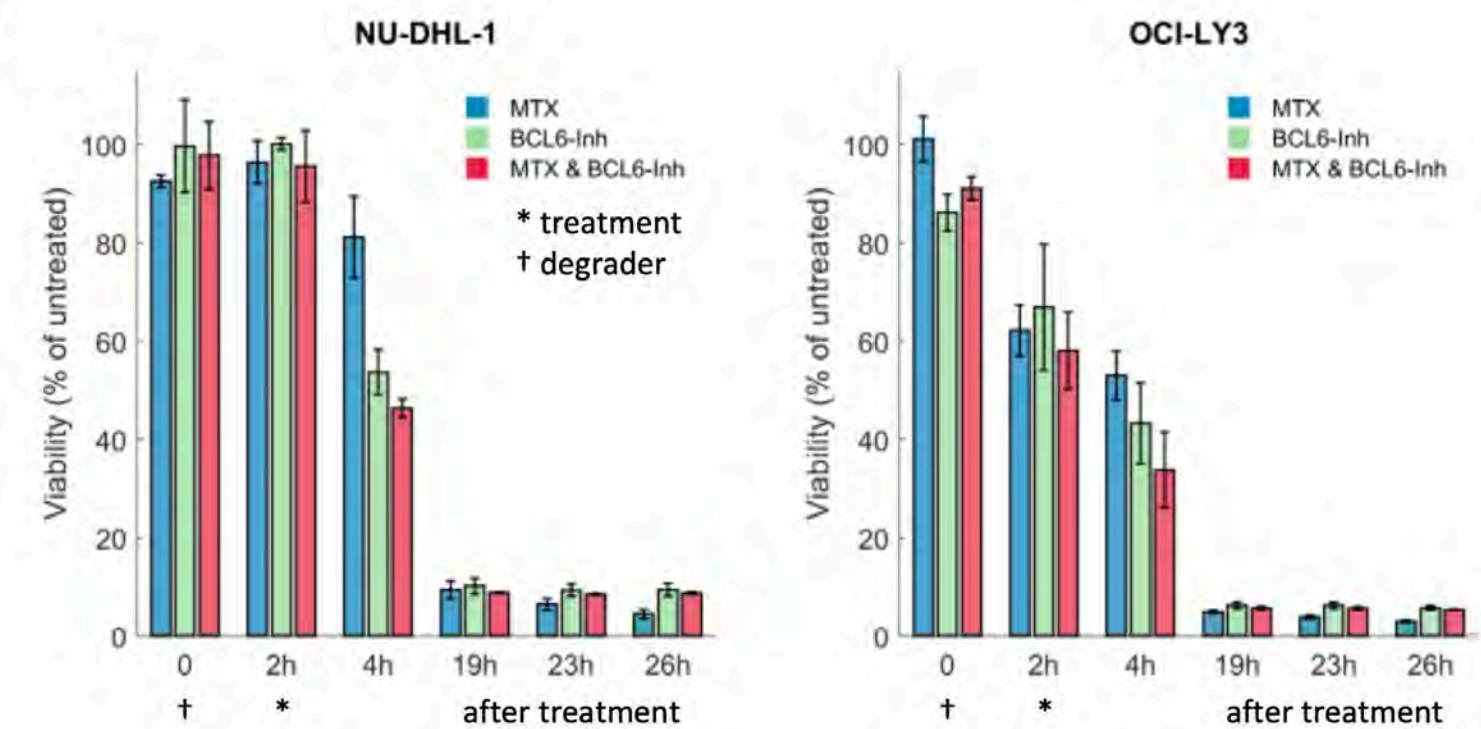
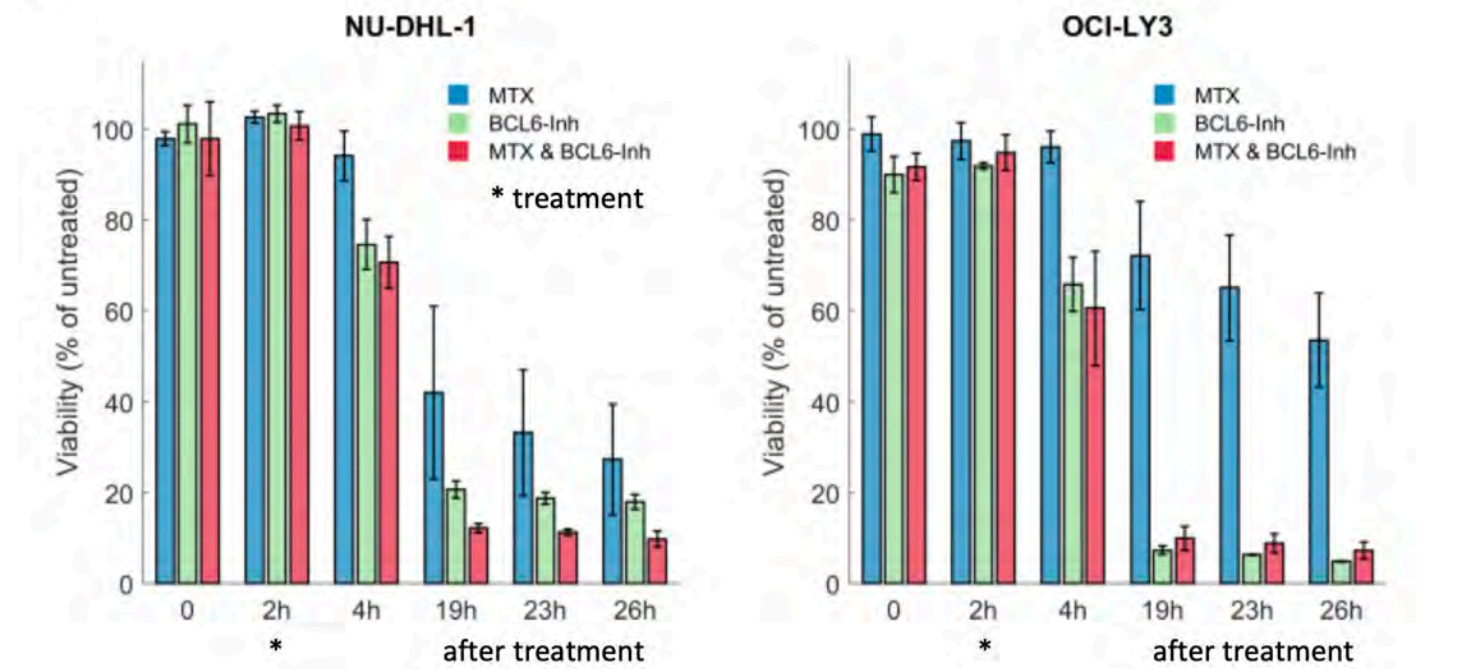
Overlap (# high-risk) 0 10



1. Predict outcome

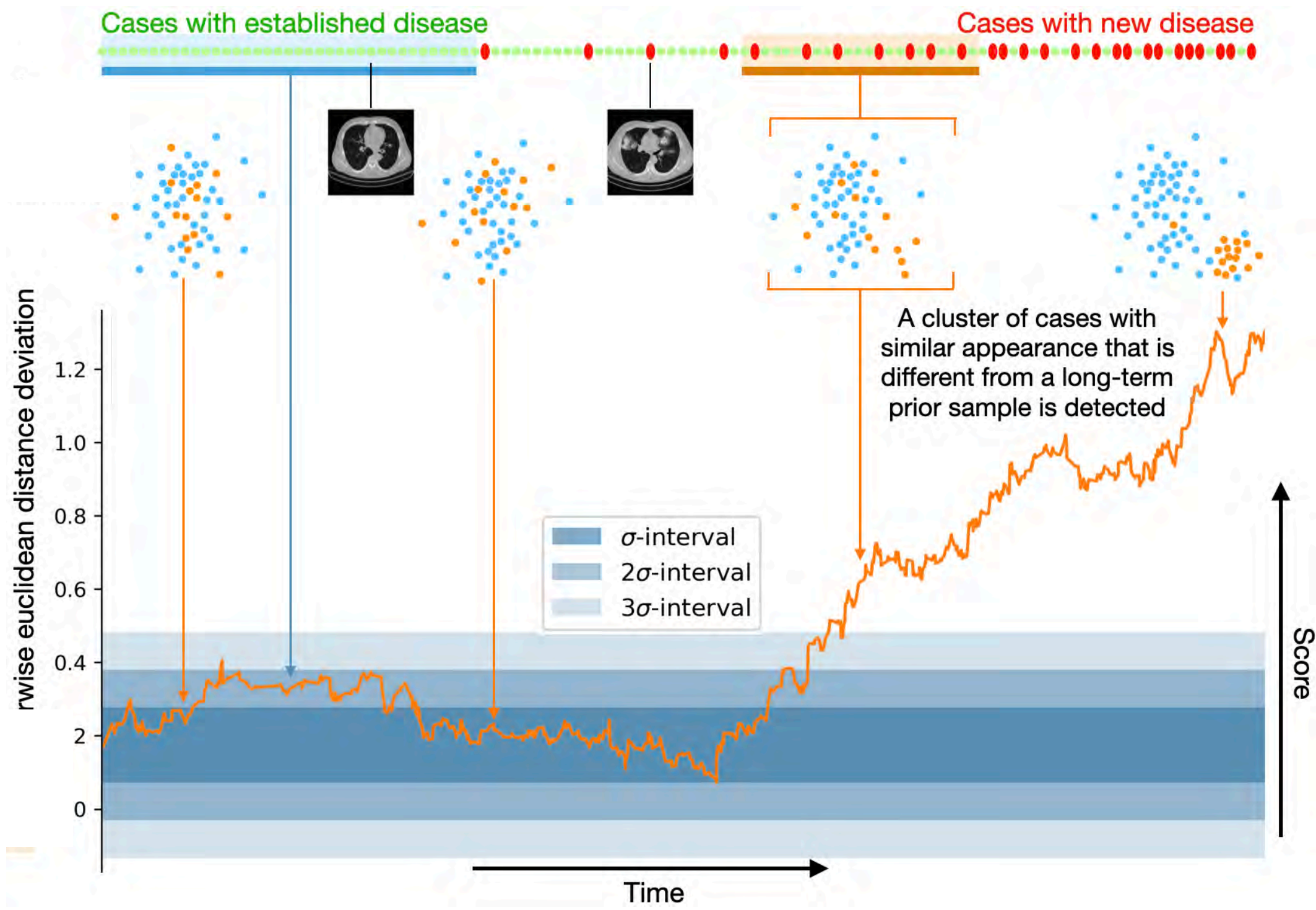
2. Identify target

3. Verify target in-vitro



Nenning et al. 2023 *Neuro. Onc. Adv.* doi.org/10.1093/noajnl/vdad136

Global surveillance for emerging diseases

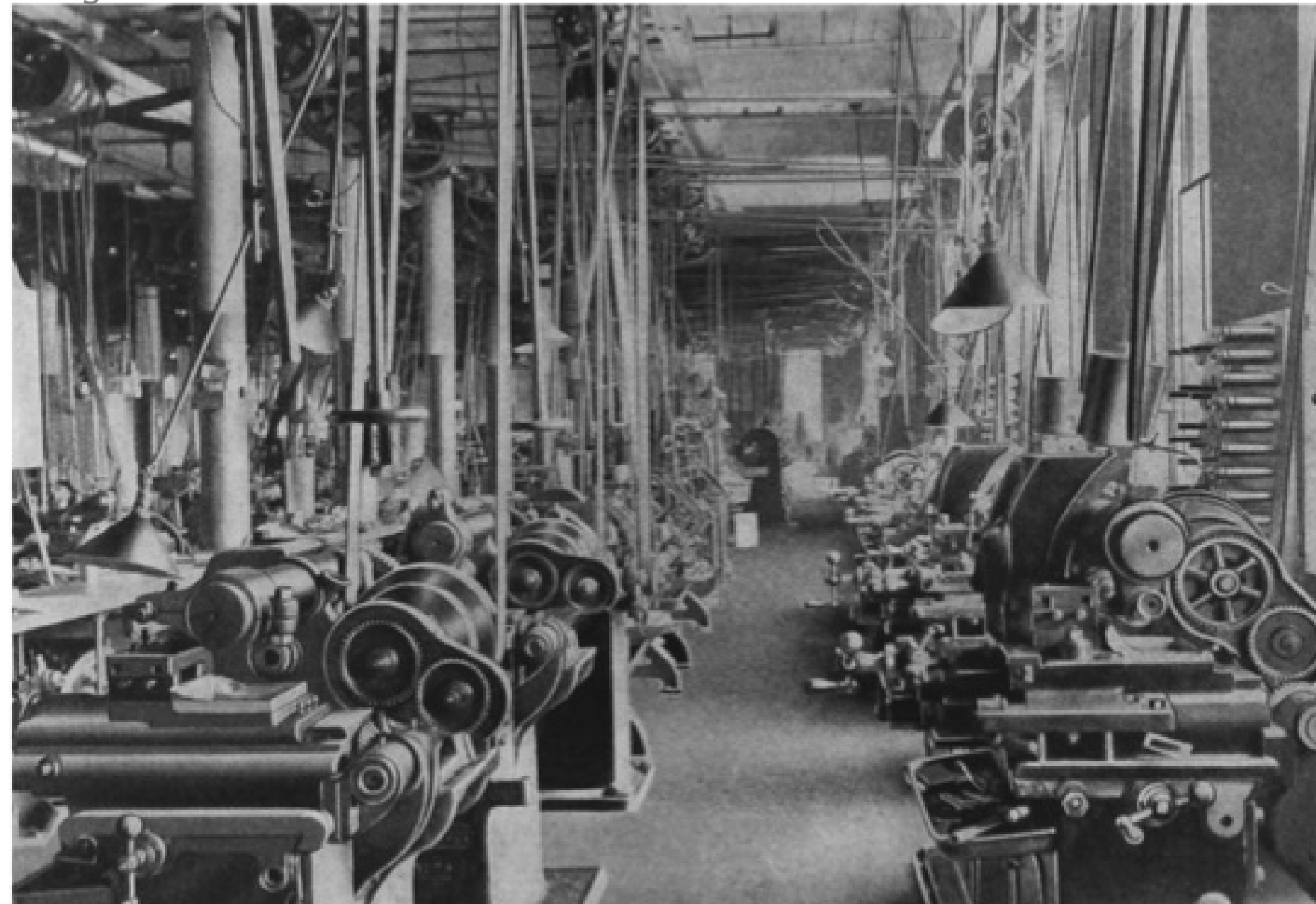


IAEA, WHO - ZODIAC Observatory

Straub et al. 2025 *Eur. Radiol.* doi.org/10.1007/s00330-024-11183-8 | Mitic et al. 2024 *MLMI* doi.org/10.1007/978-3-031-73290-4_14

Invention and adoption in industry: electricity

Image: Core77



Factory driven by steam power
Organization: along axis of power transmission

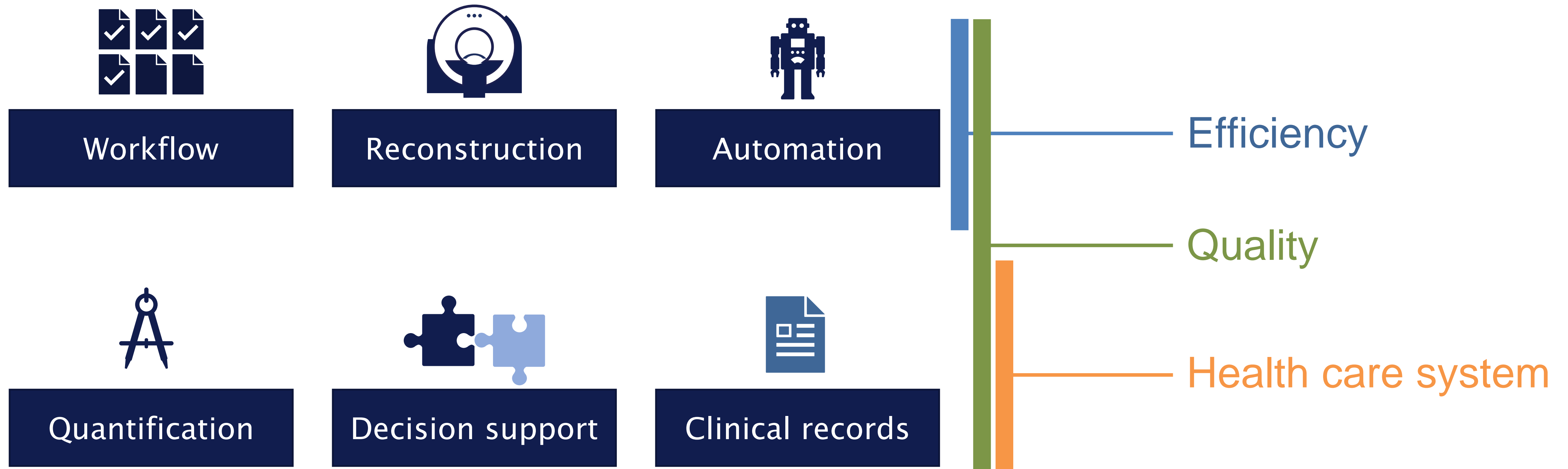
~ 40 years

Image: Getty



Factory driven by electricity
Organization: along production line

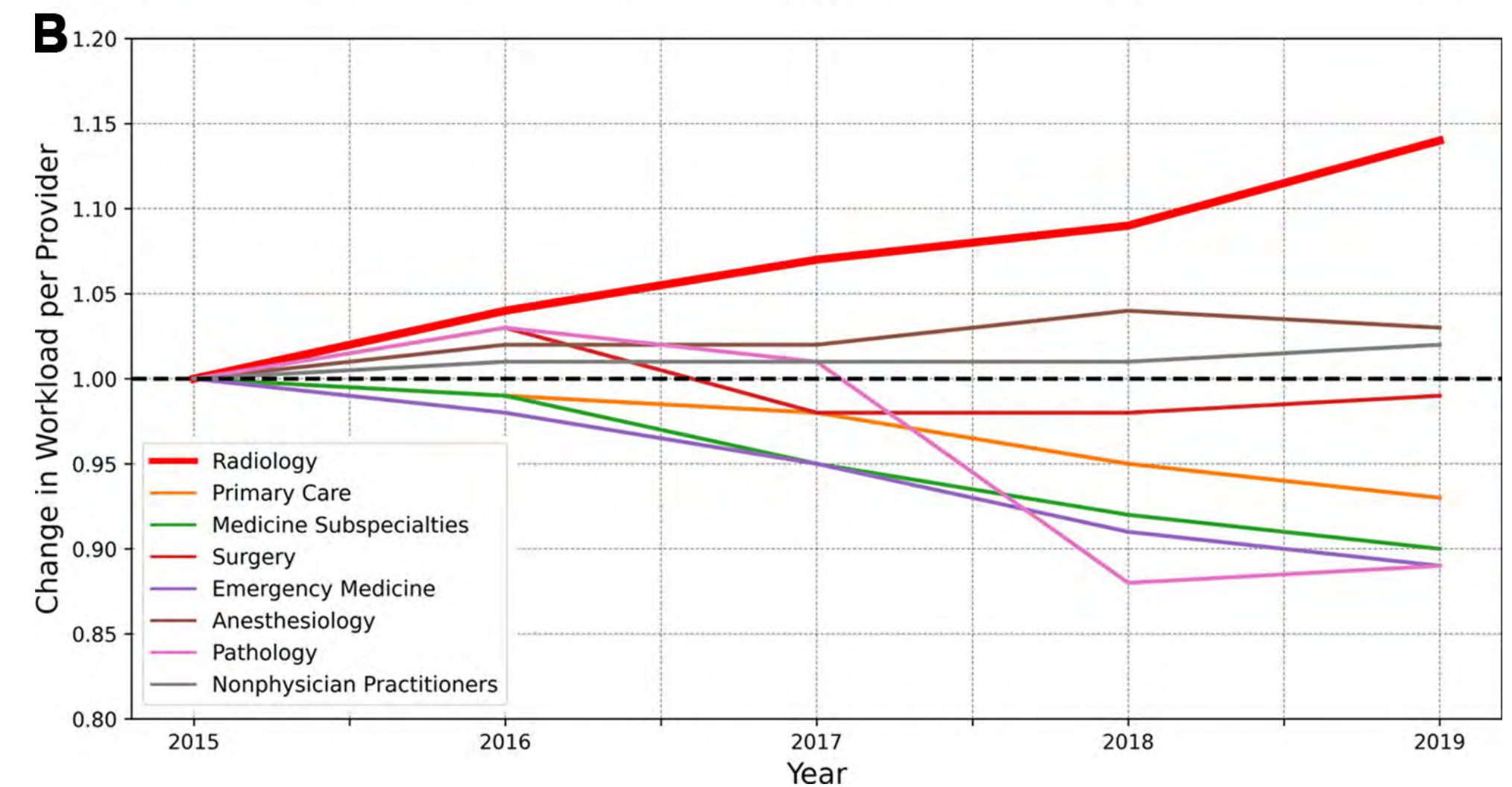
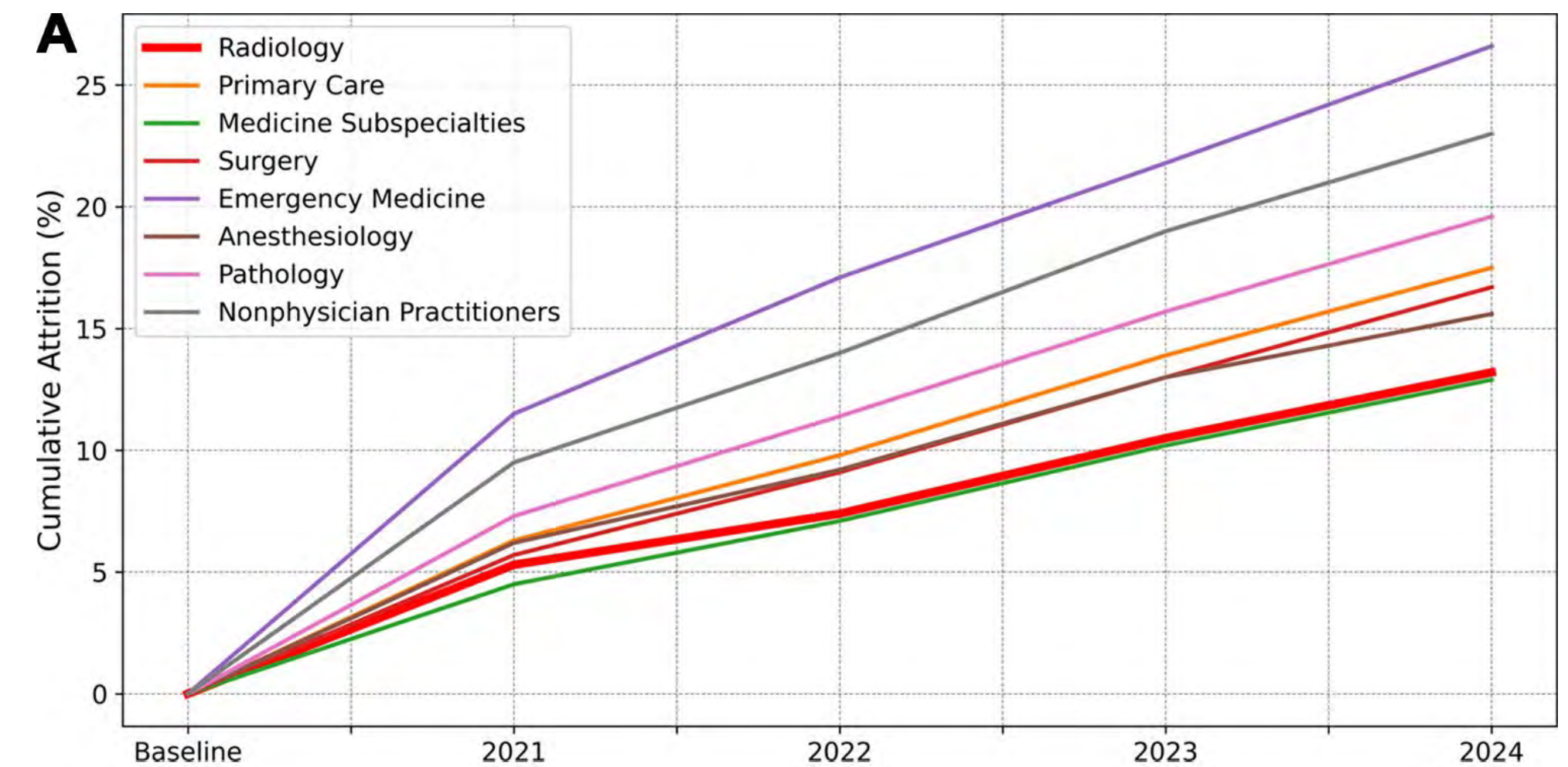
Drivers of AI adoption in clinical routine



Röhrich et al. EURA 2022, Langs Radiology 2023, Hricak & Langs Radiology 2024

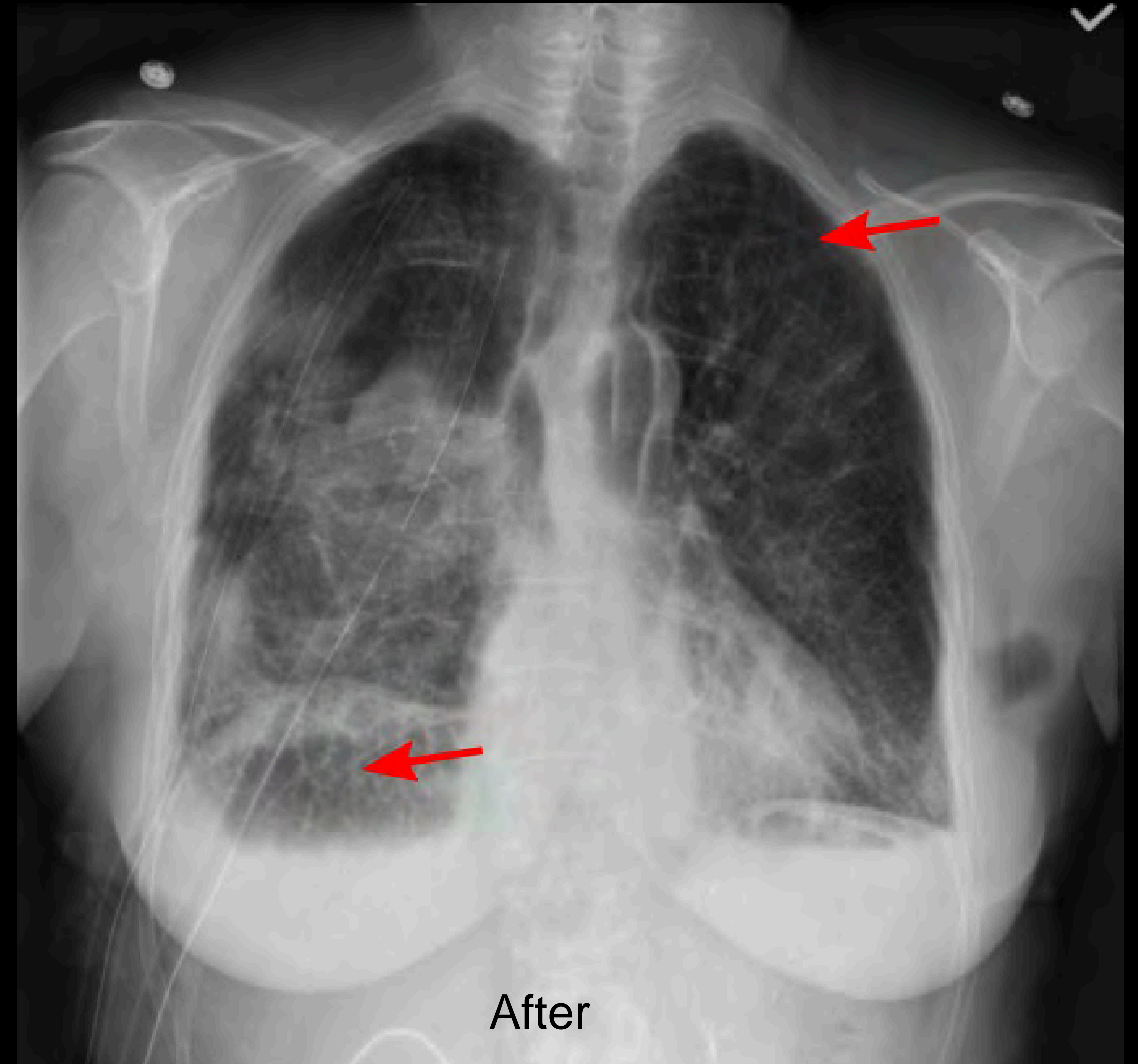
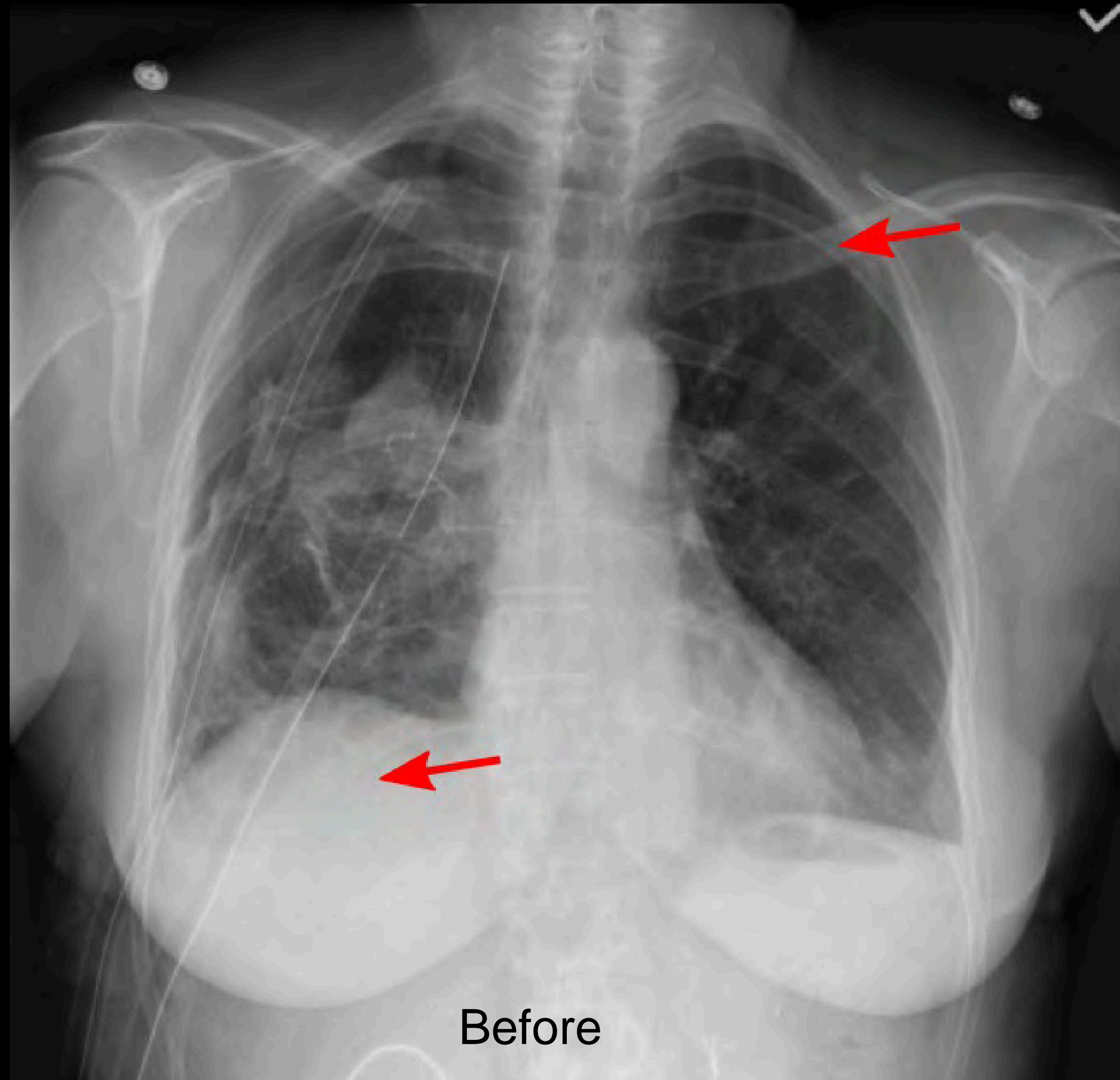
1. Efficiency - healthcare workforce gap

- Job postings *American College of Radiology* highest ever
- From 2010 to 2020:
 - DR trainees entering workforce increased by 2.5%
 - Adults over 65 increased by 34%
 - Workload increase by 55-143%

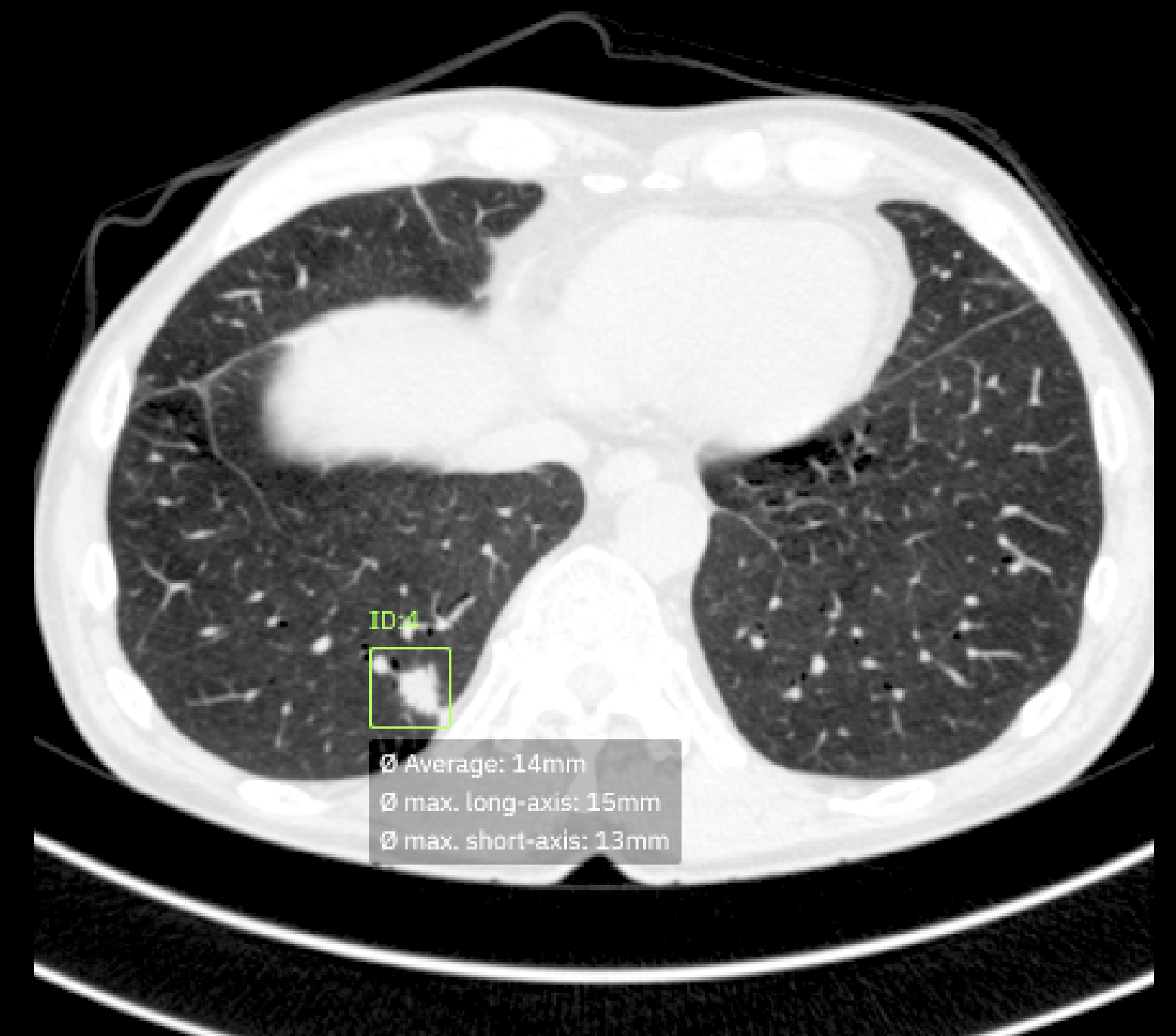
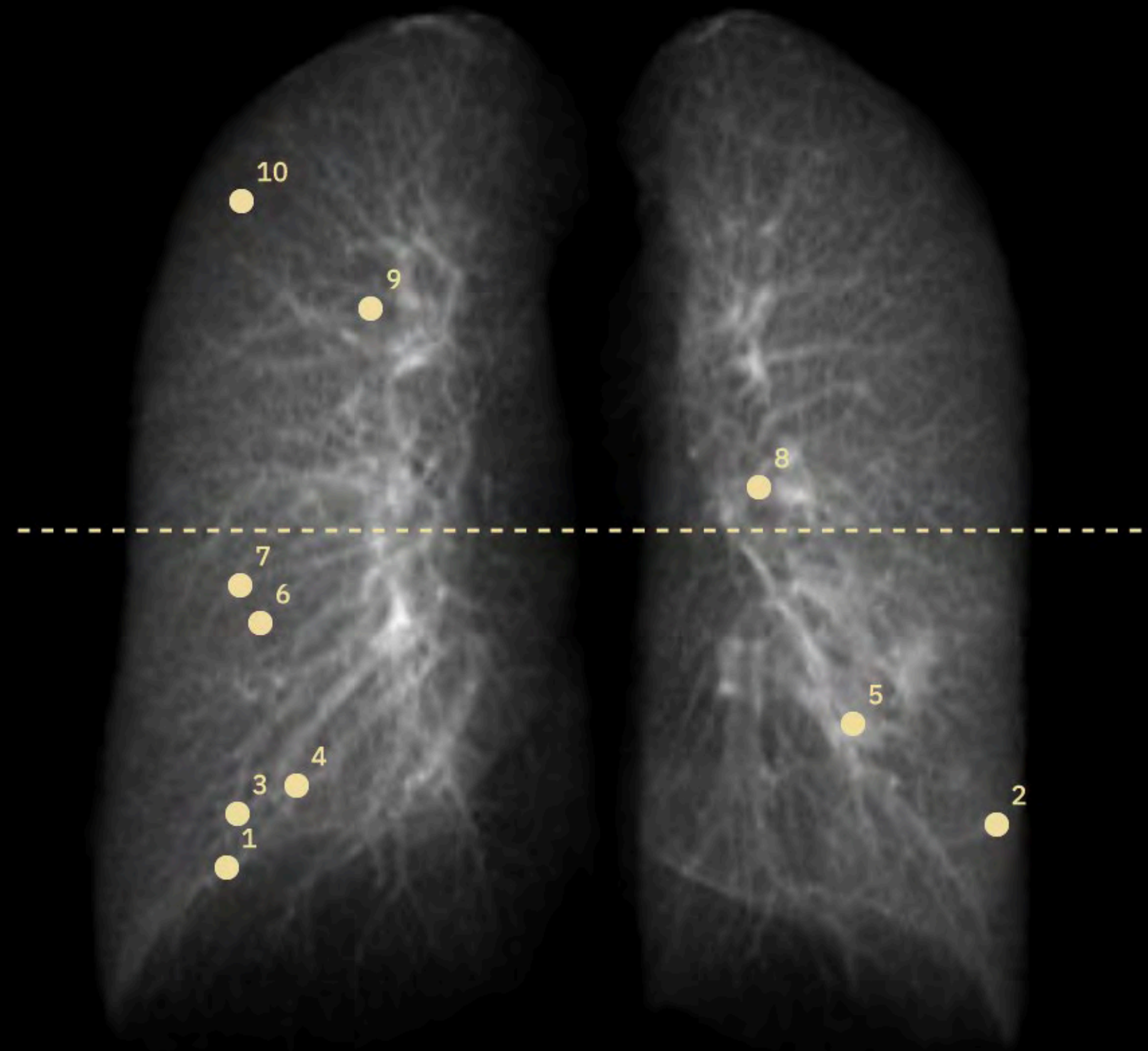
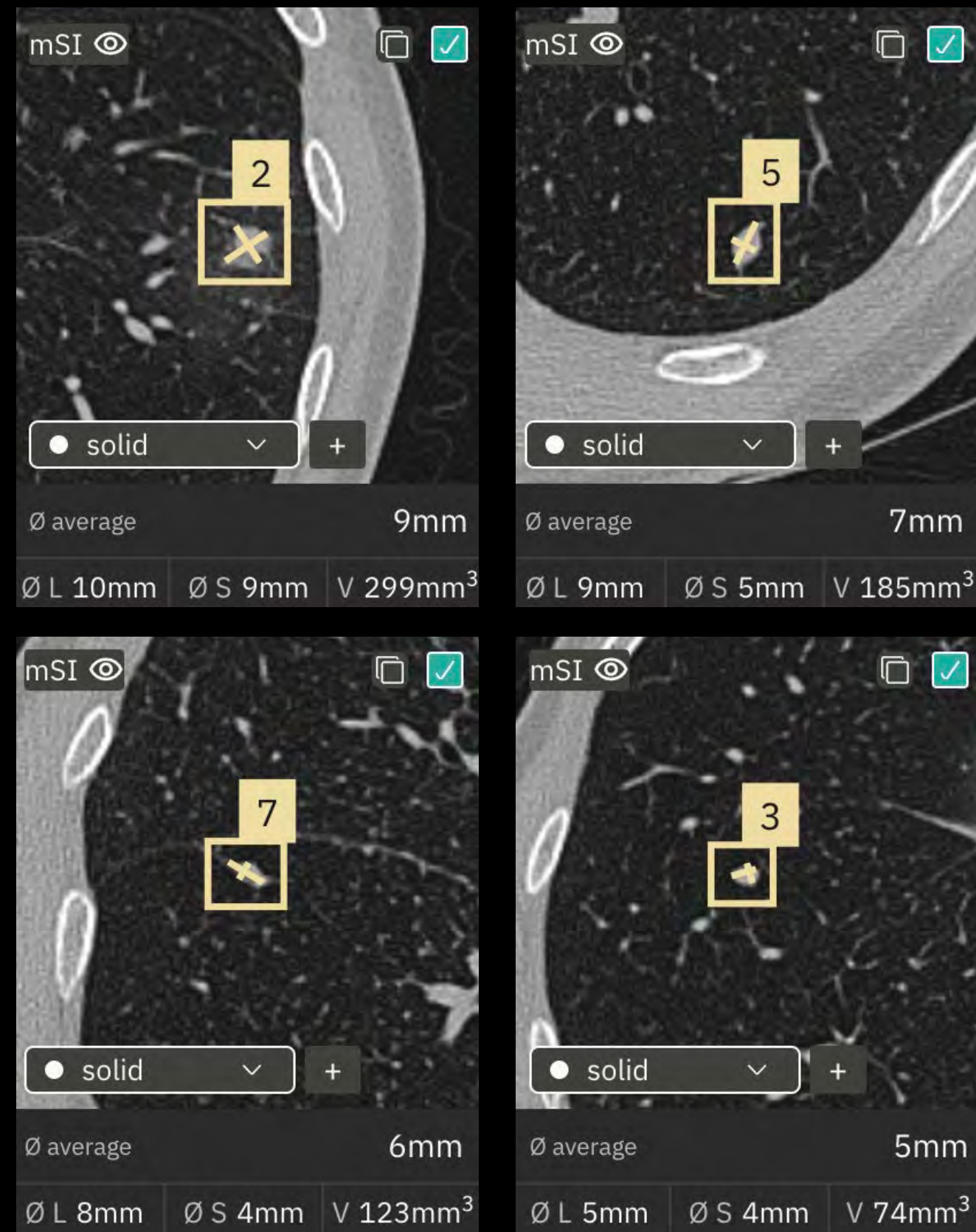


Workforce attrition in healthcare (US)

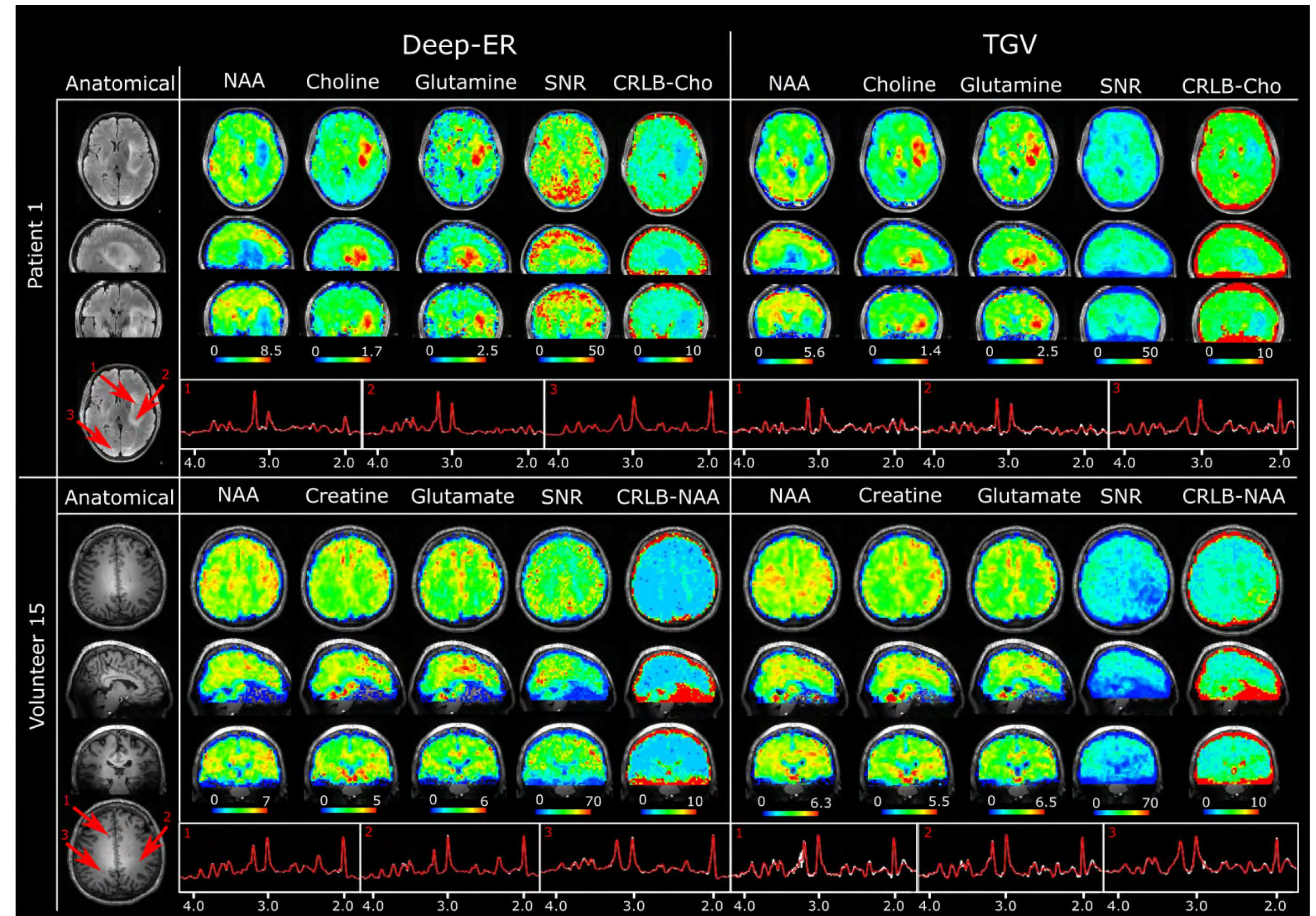
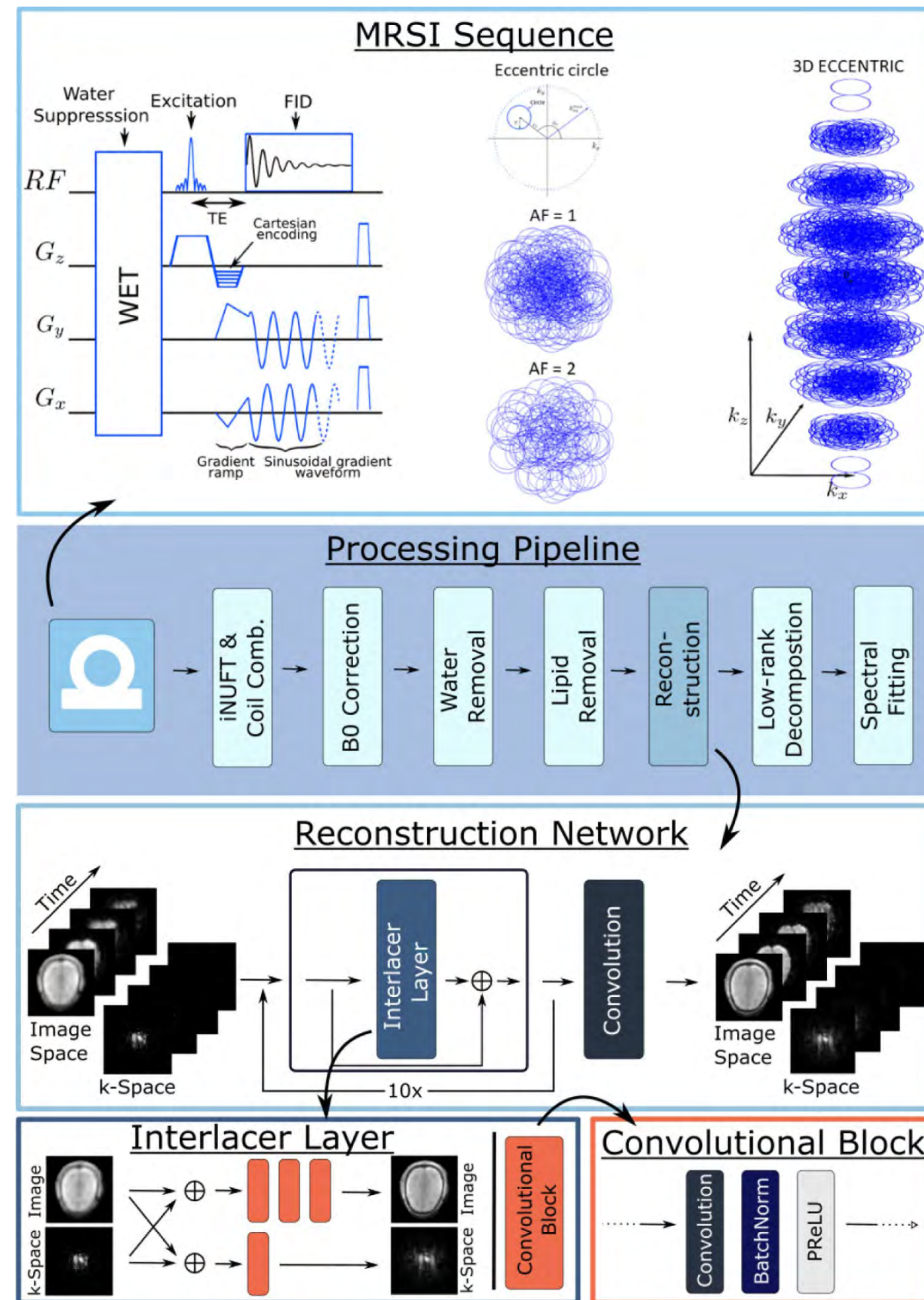
Enhancing images with AI



Automatic detection and risk prediction

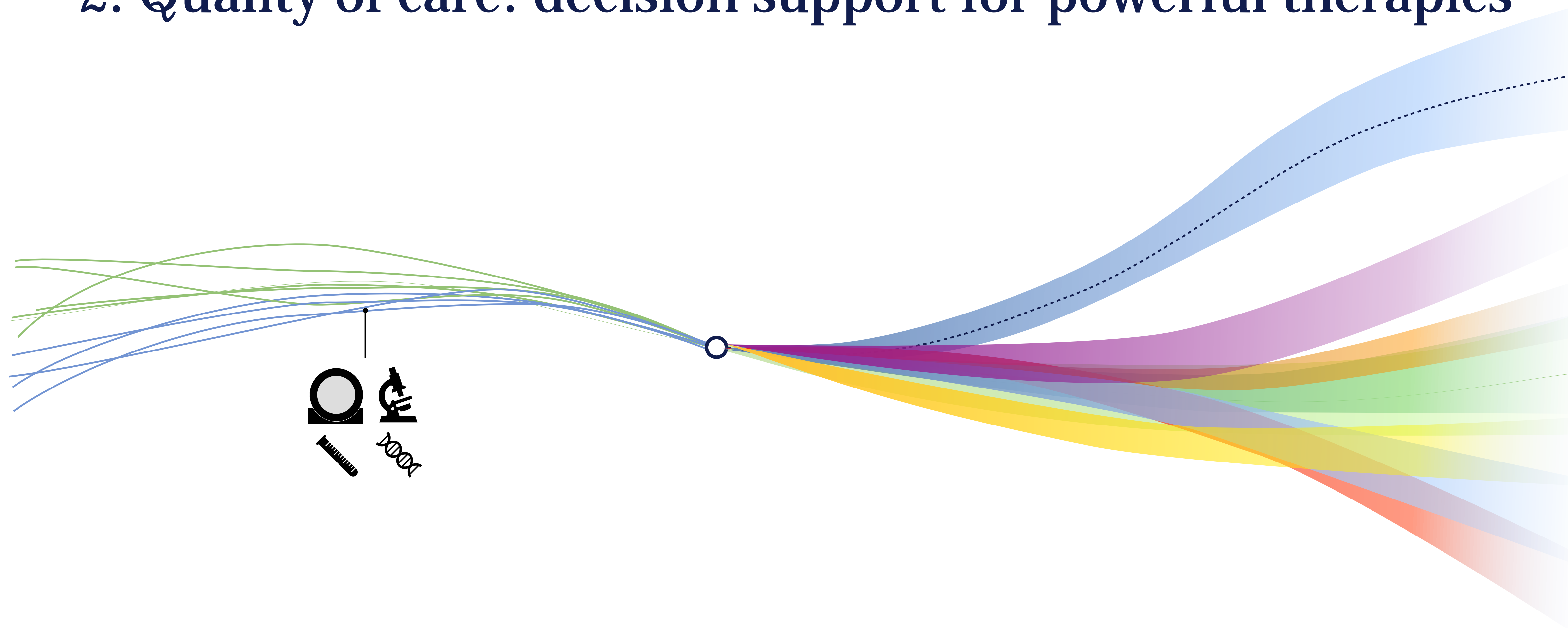


Reconstructing MR spectroscopy imaging

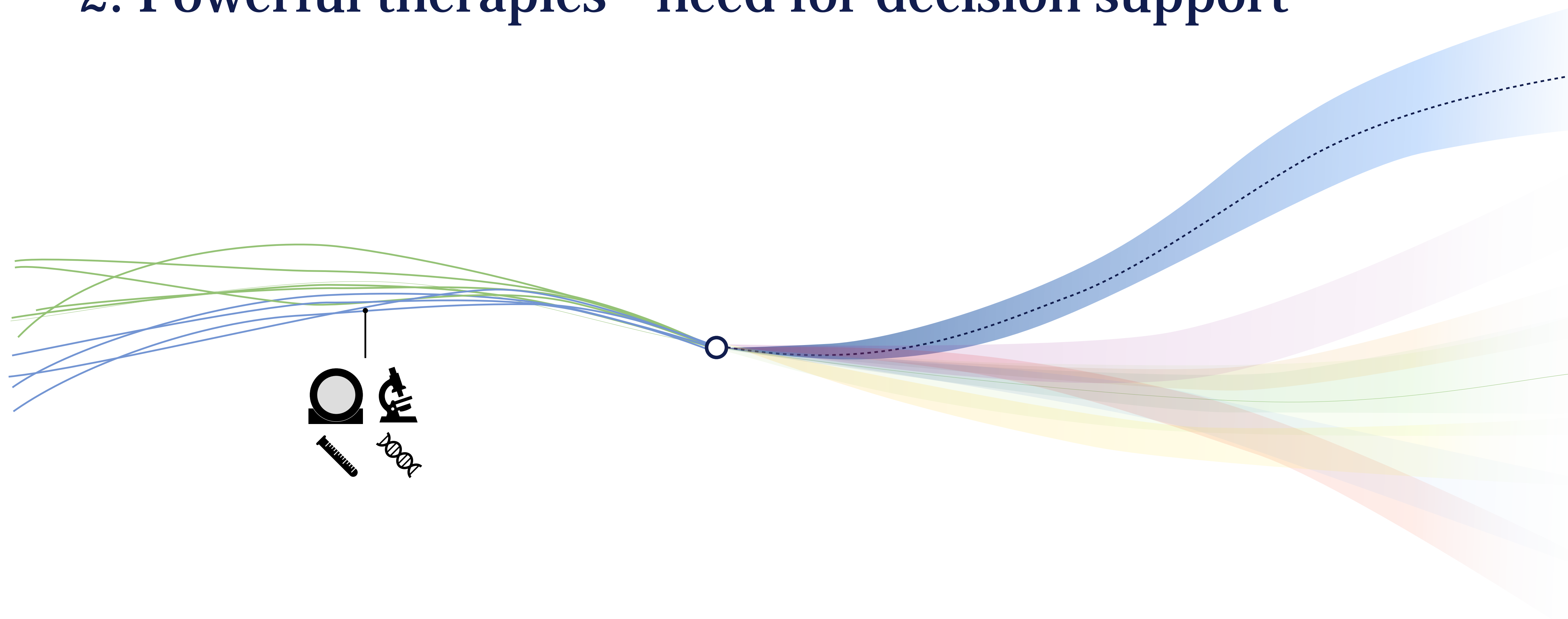


Weiser et al. 2025 <https://doi.org/10.1016/j.neuroimage.2025.121045>

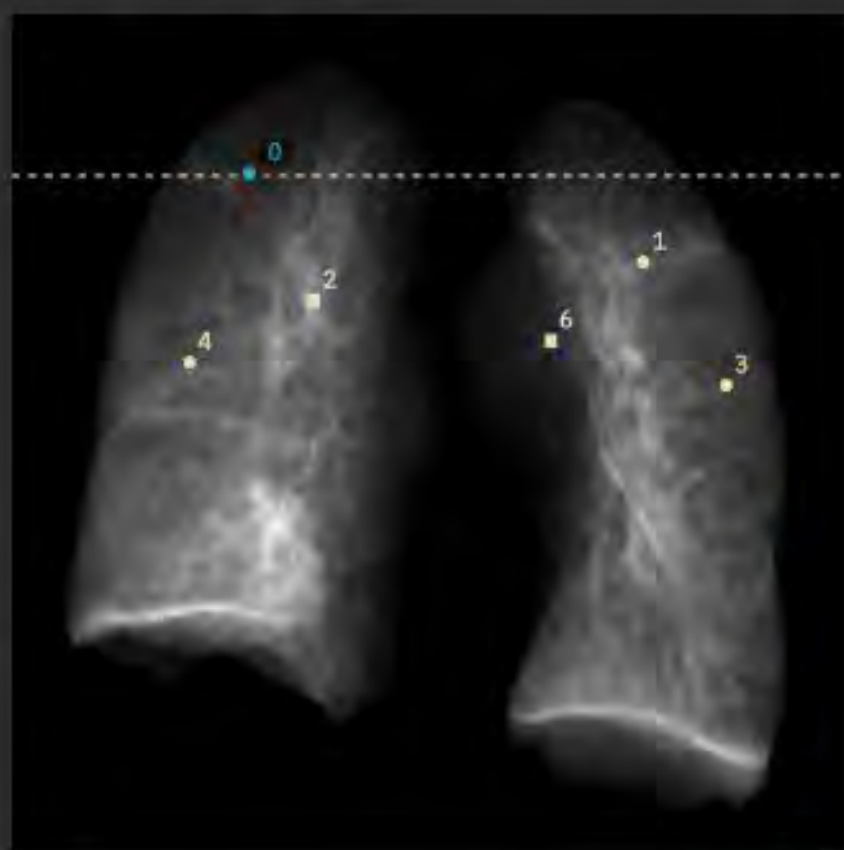
2. Quality of care: decision support for powerful therapies



2. Powerful therapies - need for decision support



Series description:
 Name: Toby FLENDERSON
 Age: 72 years
 Sex: M
 Slice: 30 of 150



Δ Cumulative volume	-576mm ³
04.03.2016	
Detected nodules	5
Cumulative volume	2328mm ³
04.04.2016	
Detected nodules	6
Cumulative volume	1752mm ³

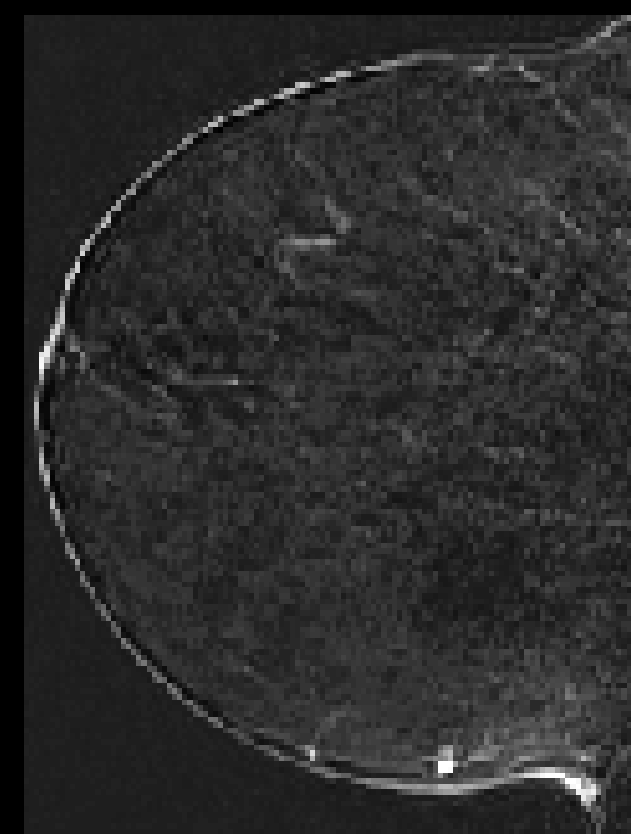
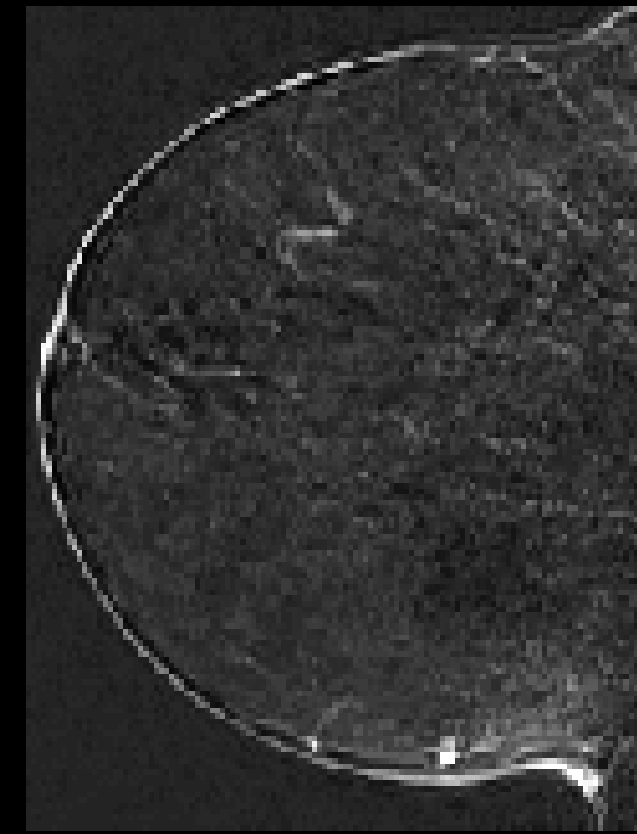
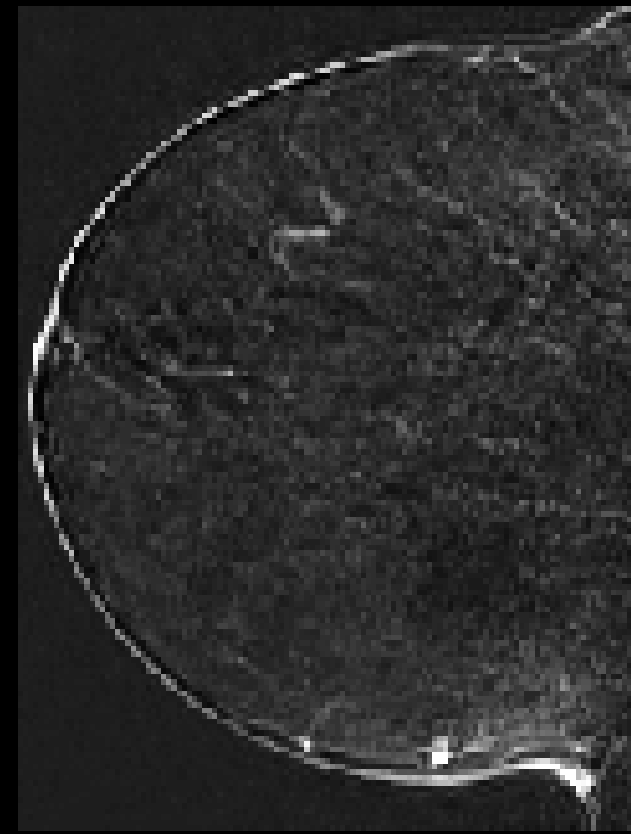
Guideline: Fleischner Guidelines | Sort by: volume

Max. long: 4mm | 30mm

Date	Volume (mm ³)	Average (mm)	Max. Long (mm)	Max. Short (mm)	Time to Double (d)
07.10.2014	2691	5	5	4	-53%
04.03.2016	2198	10	13	6	-139d
04.04.2016	1752	14	17	11	-18%
21.04.2015	2328	10	13	6	-726d
19.11.2015	1752	10	13	6	+6%
04.03.2016	2328	10	13	6	-1279d
04.04.2016	1752	14	17	11	-25%
04.04.2016	1024	14	17	11	-76d
04.04.2016	329	9	10	9	-1170d
04.04.2016	309	10	12	8	-188d
04.04.2016	275	9	10	7	-188d
04.04.2016	2390	21	23	20	-123d
04.04.2016	1472	18	19	16	-303d
04.04.2016	1295	19	21	17	-574d
04.04.2016	176	9	11	6	-11d
04.04.2016	233	7	8	5	-2709d
04.04.2016	221	7	8	5	-284d
04.04.2016	170	6	7	5	-97d
04.04.2016	137	6	7	6	-97d
04.04.2016	67	5	5	4	-318d
04.04.2016	91	5	5	4	484d
04.04.2016	80	5	5	4	-535d
04.04.2016	96	5	5	5	116d
04.04.2016	44	-	-	-	-

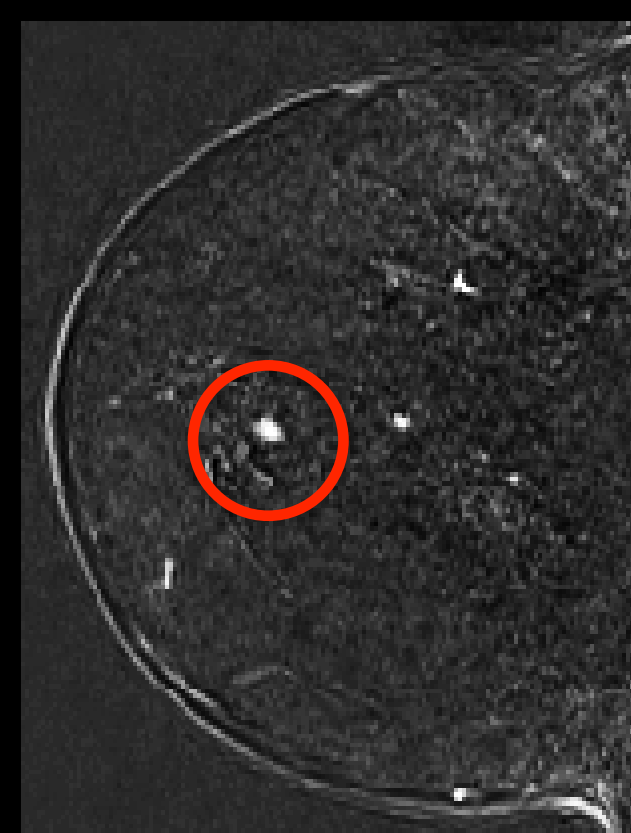
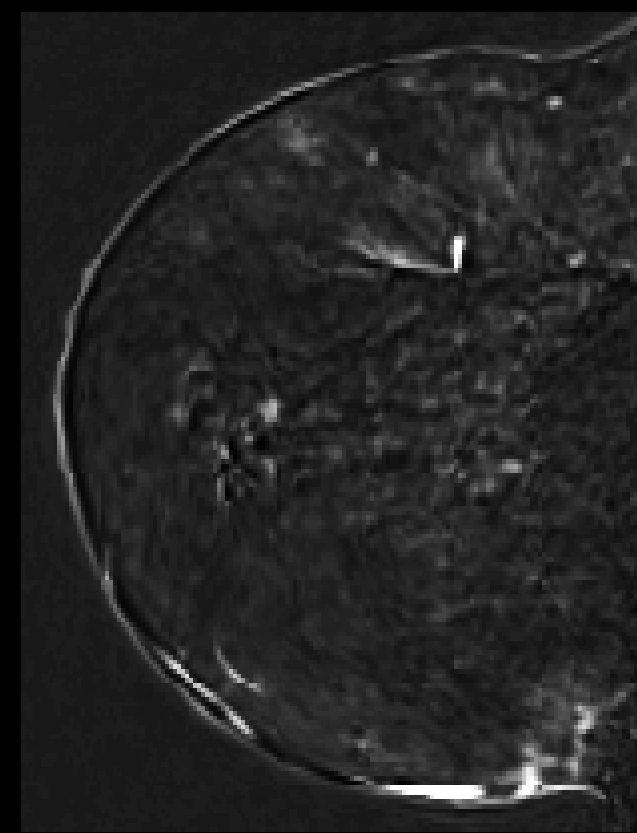
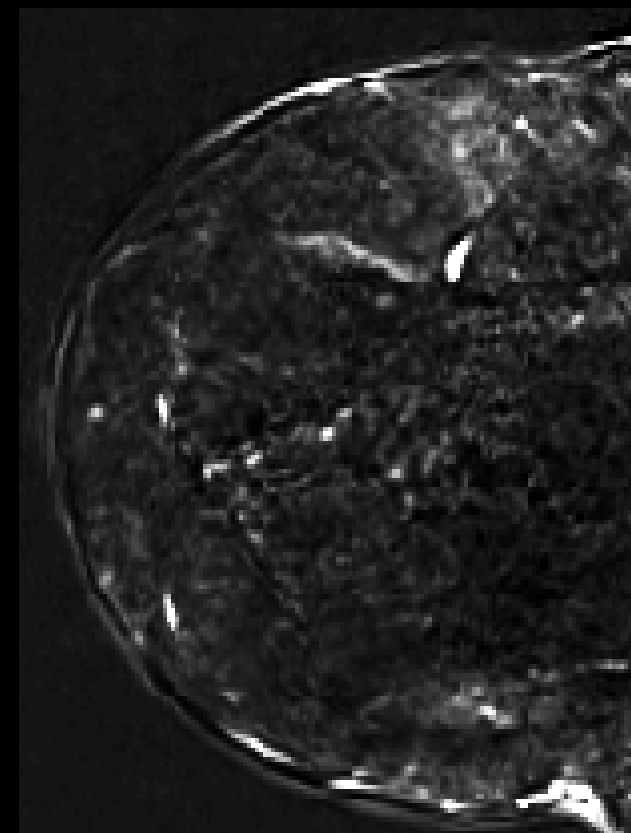
Prevention in breast cancer: early disease signatures

Subject 1



No lesion

Subject 2

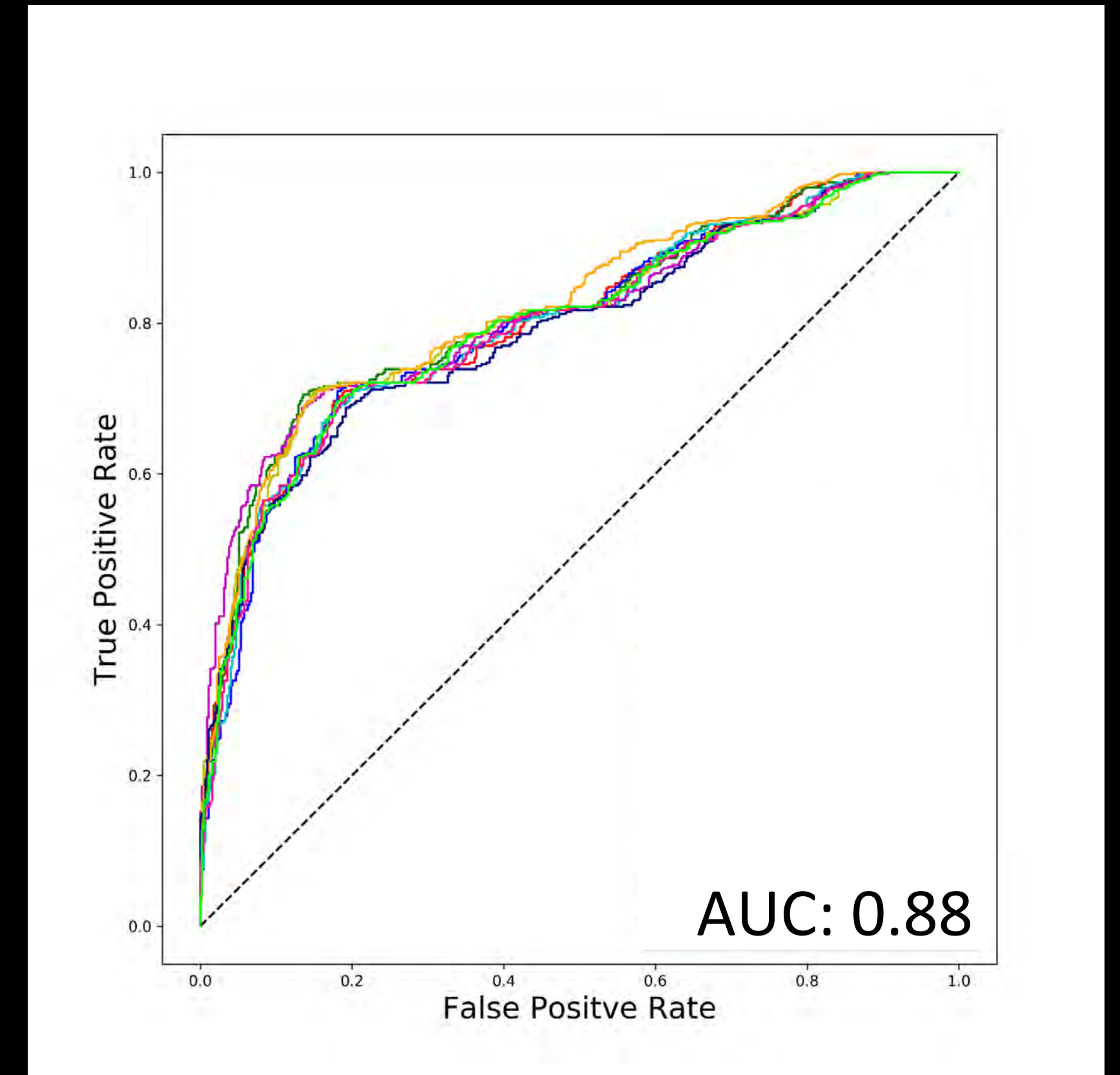


Lesion

Previous visit

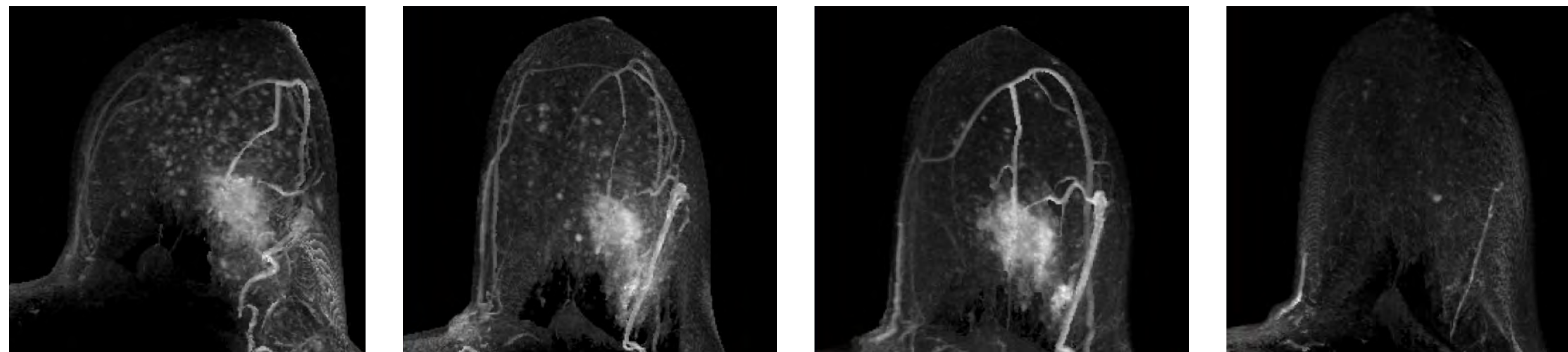
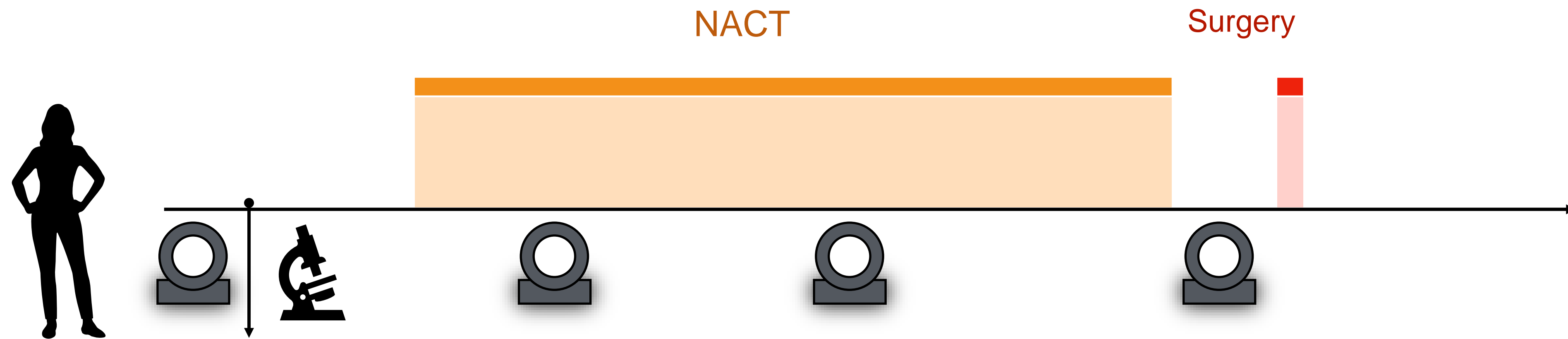
Now

Next visit



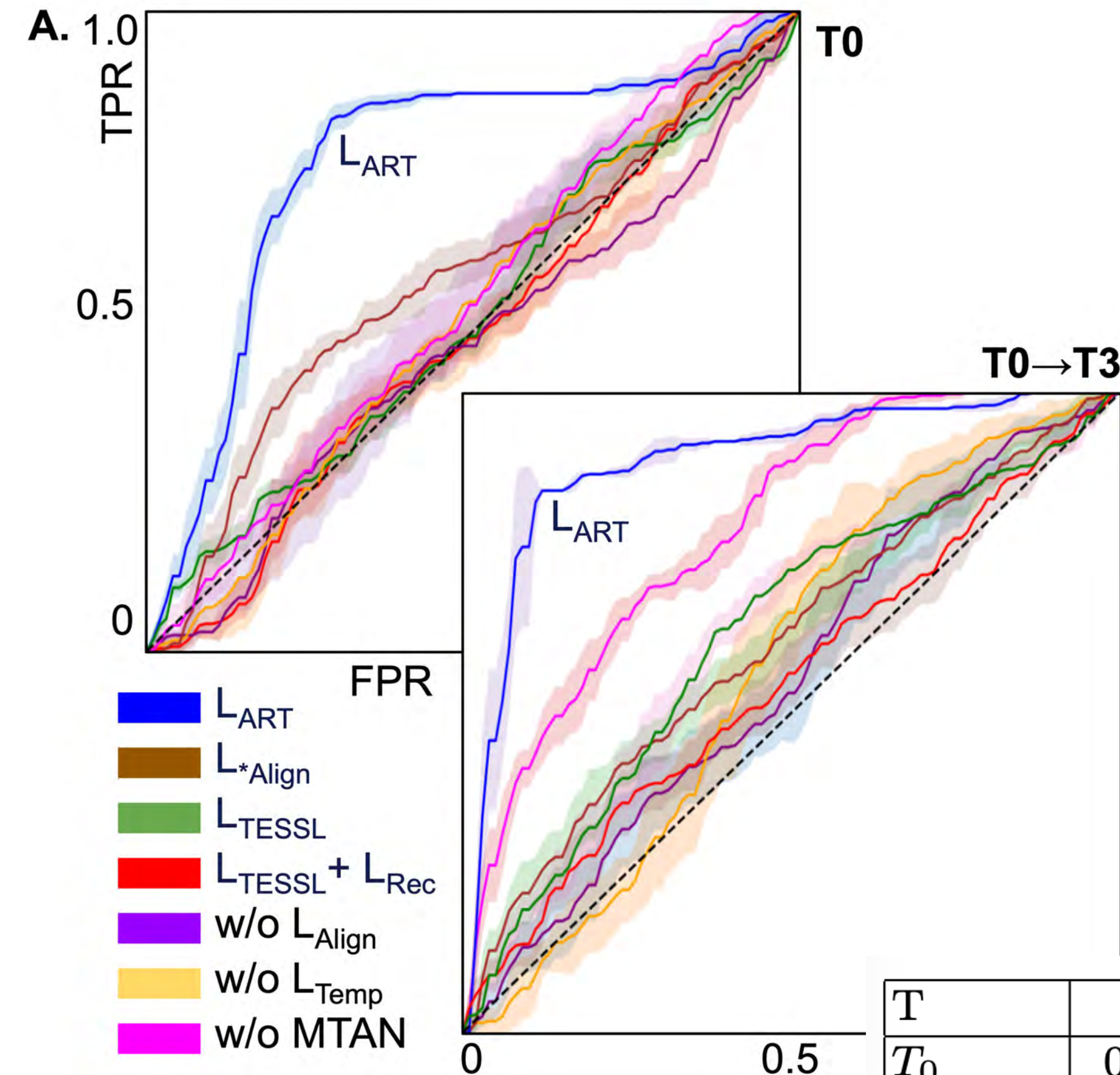
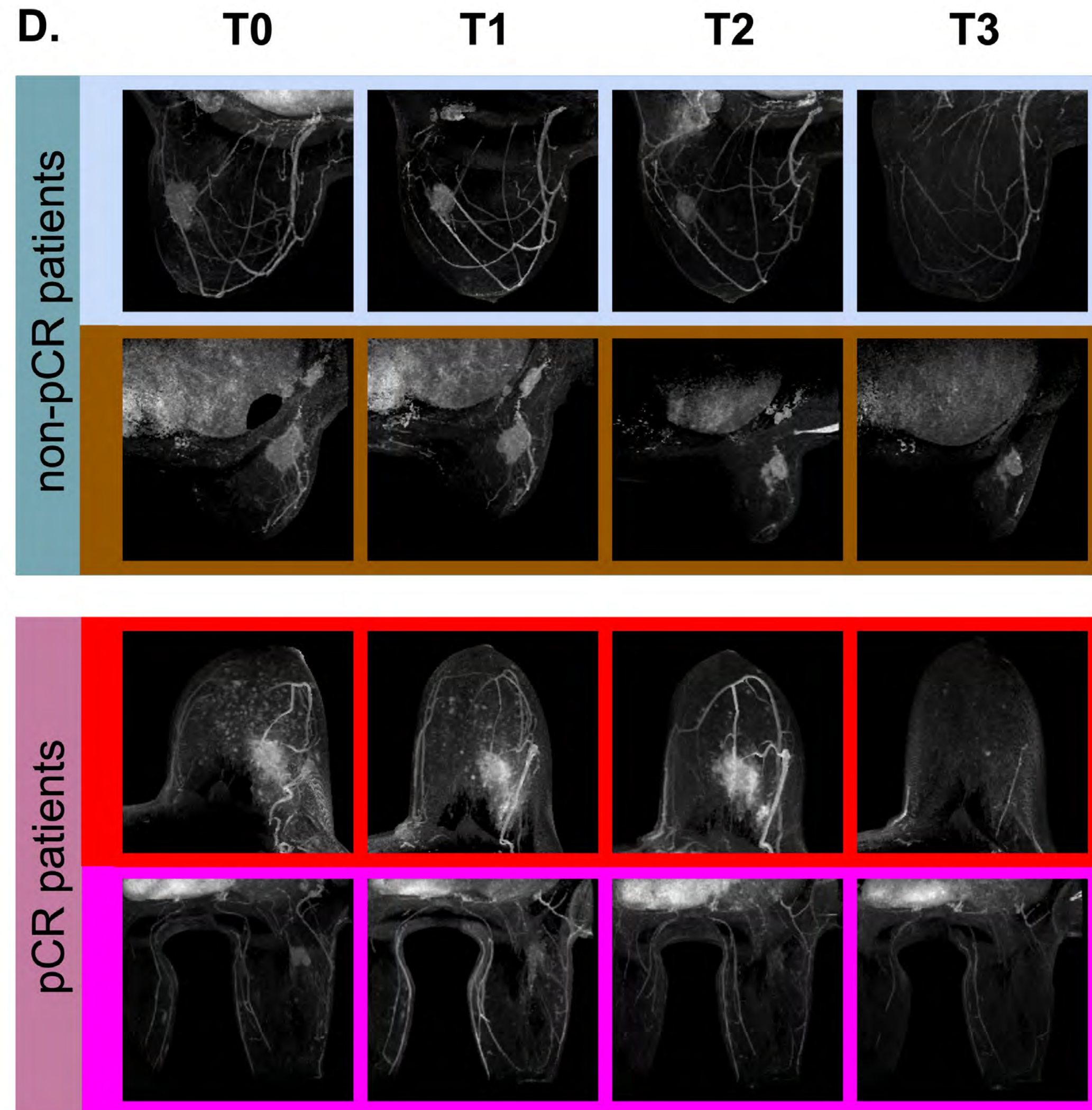
Burger, Helbich, Langs et al EURE 2023

Predicting response to Neoadjuvant Chemotherapy in BC



WWTF - PREDICTOME Janickova et al. MICCAI 2025 <https://arxiv.org/abs/2509.14872>

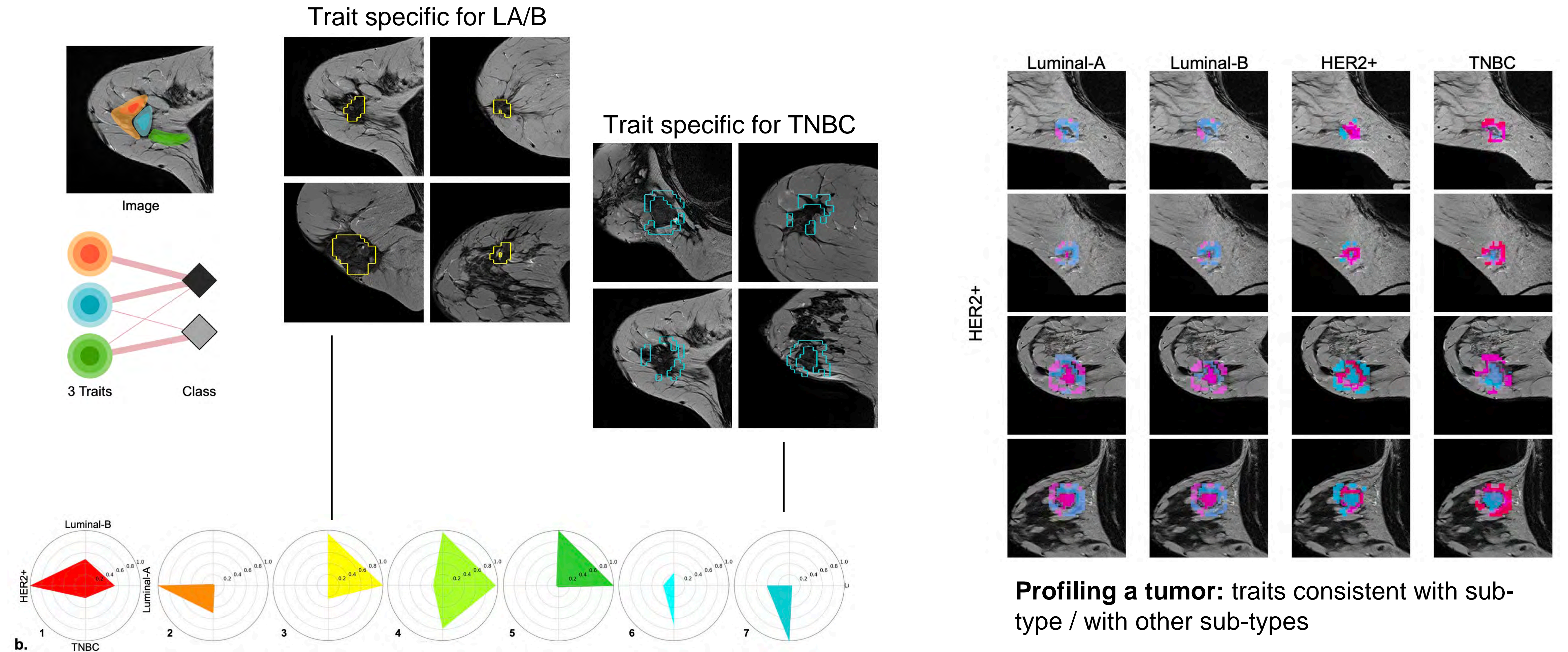
Predicting response to Neoadjuvant Chemotherapy in BC



T	PPV	NPV
T_0	$0.603 \pm .02$	$0.852 \pm .01$
$T_0 + T_1$	$0.721 \pm .02$	$0.883 \pm .02$
$T_0 \rightarrow T_3$	$0.772 \pm .02$	$0.920 \pm .01$

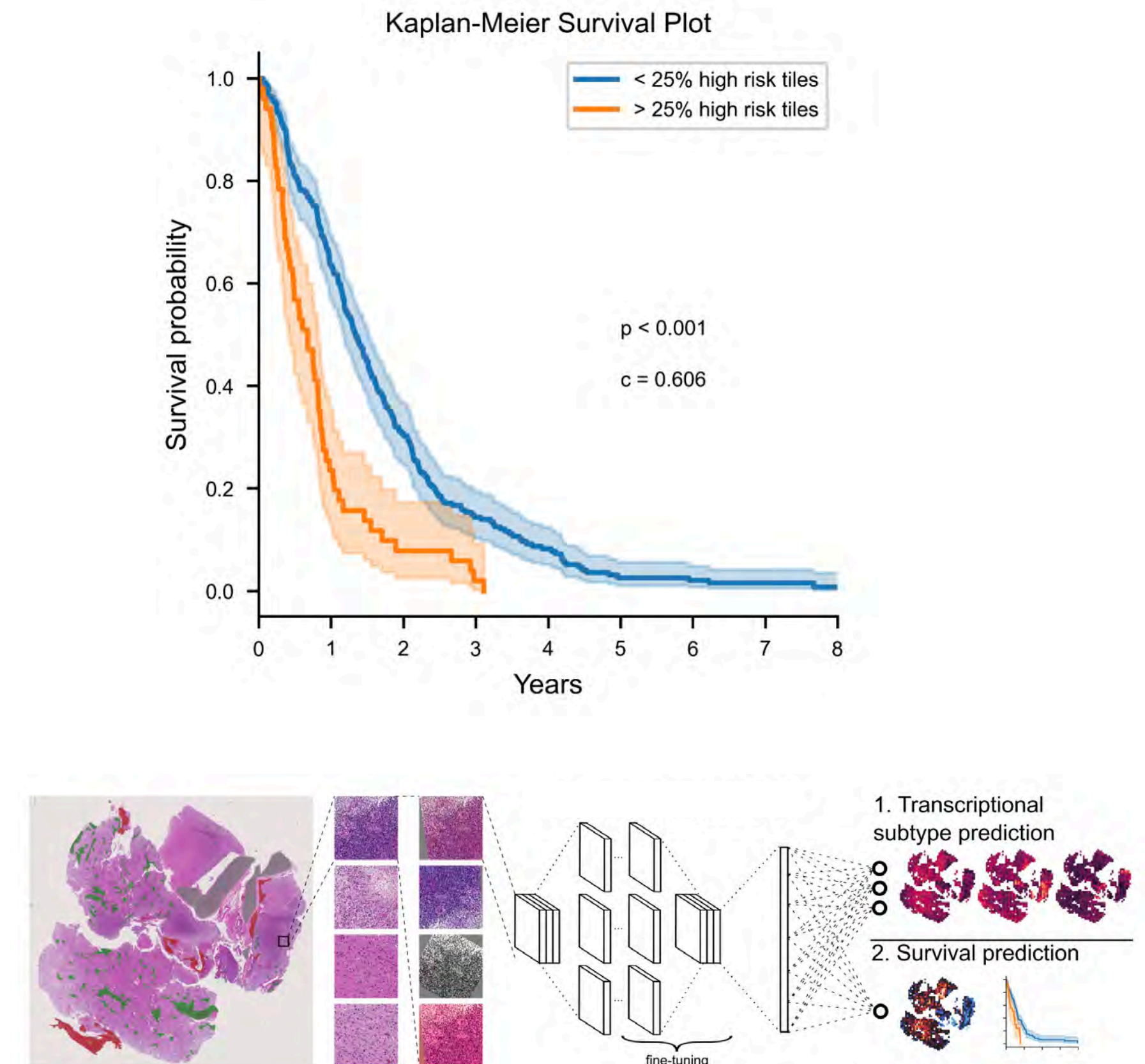
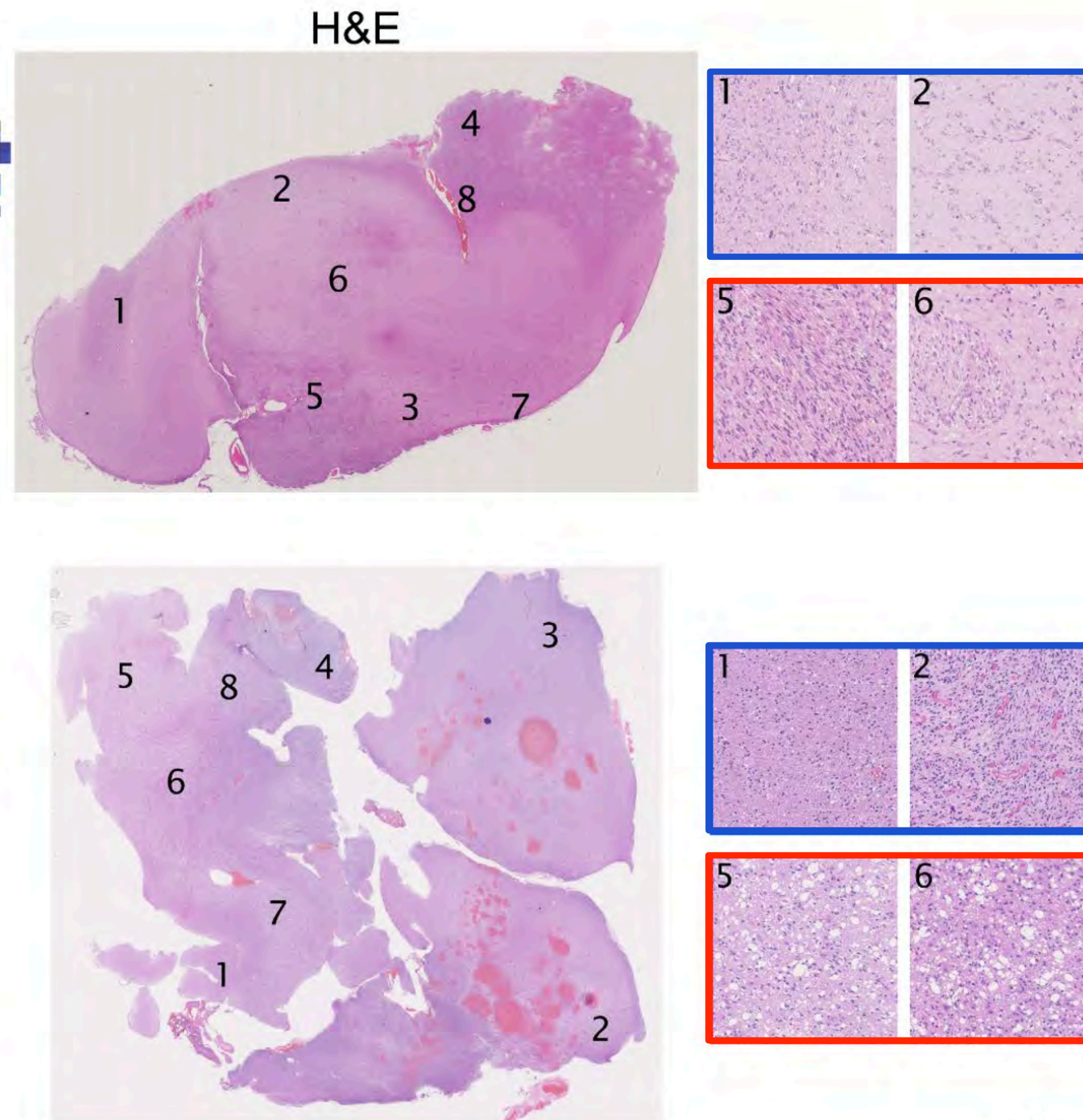
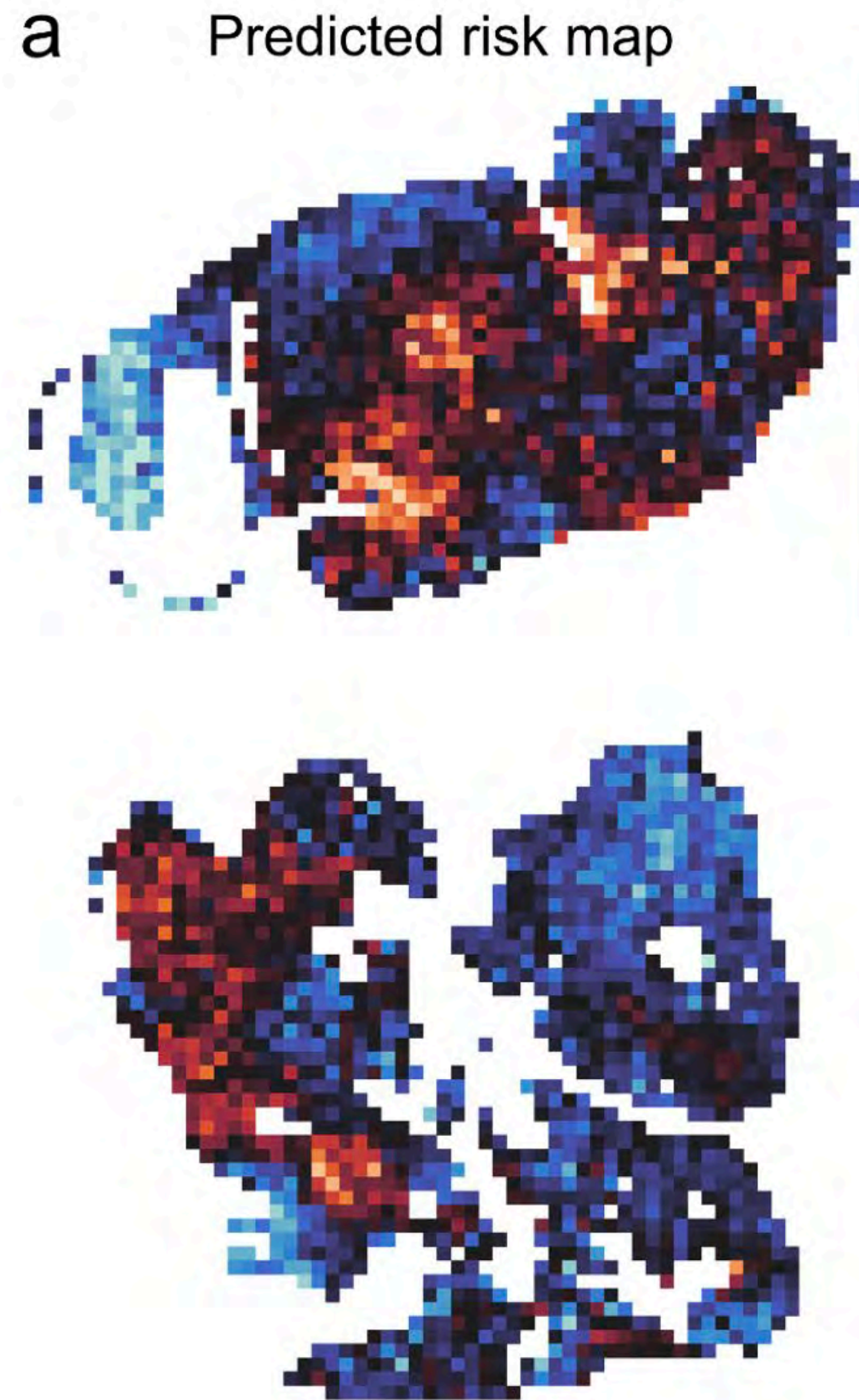
WWTF - PREDICTOME Janickova et al. MICCAI 2025 <https://arxiv.org/abs/2509.14872>

Link imaging traits to molecular tumor sub-types



Funded by **WWTF** www.cir.meduniwien.ac.at/predictome

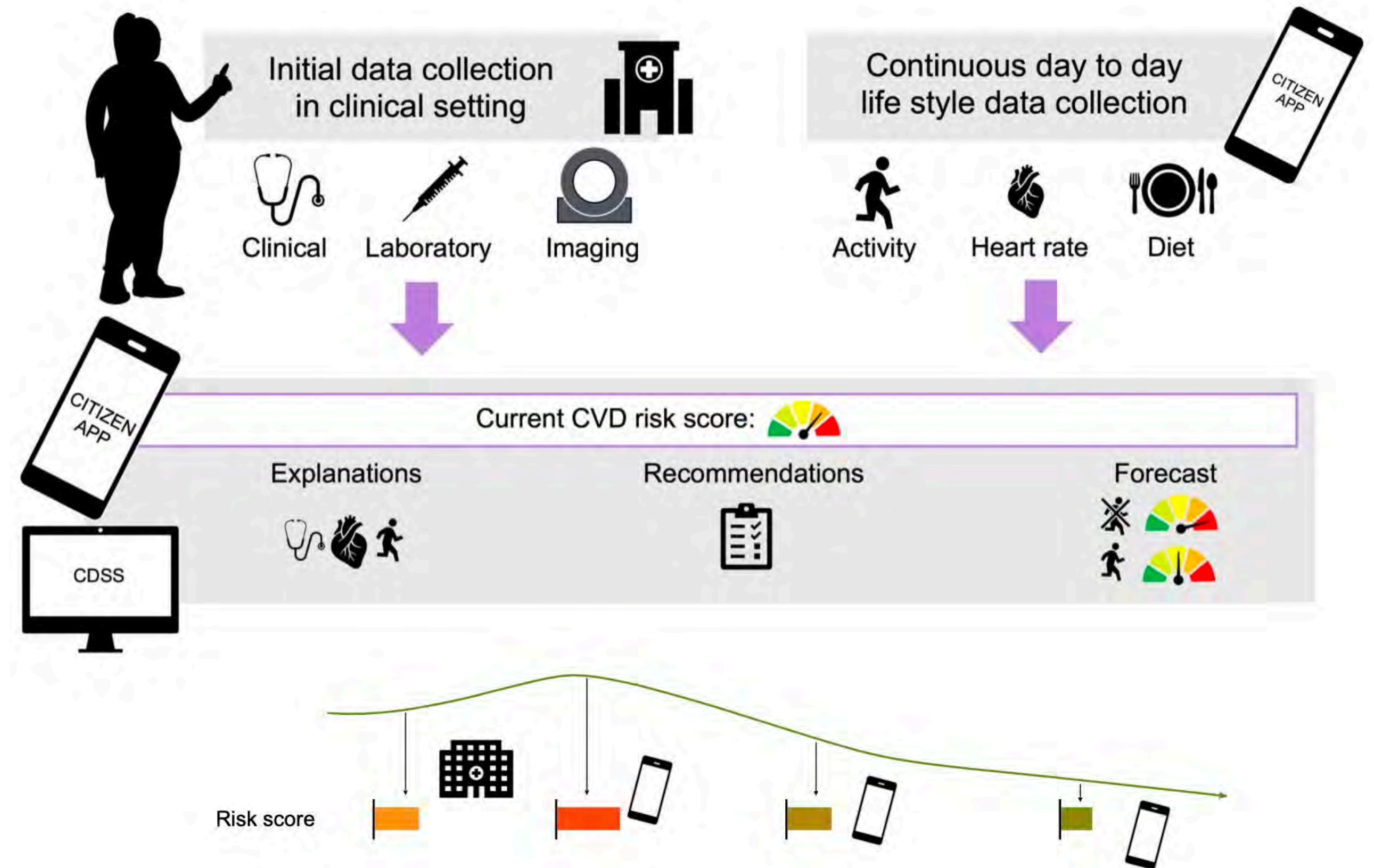
Linking imaging data to molecular information and risk



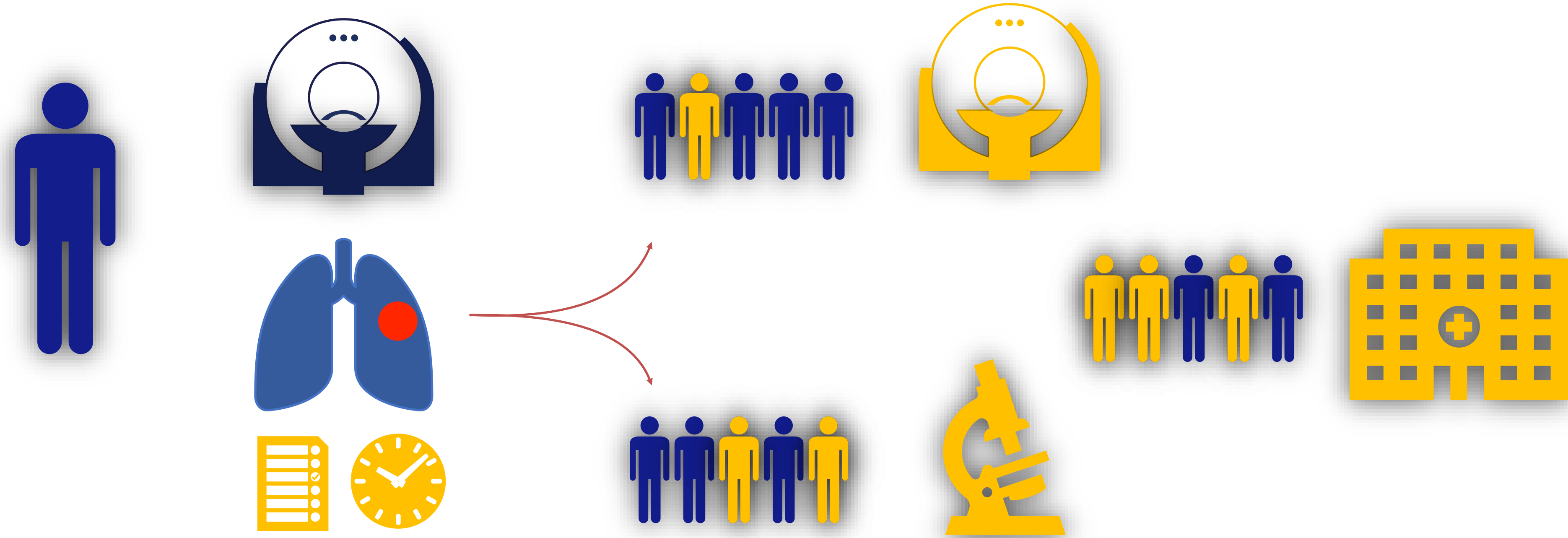
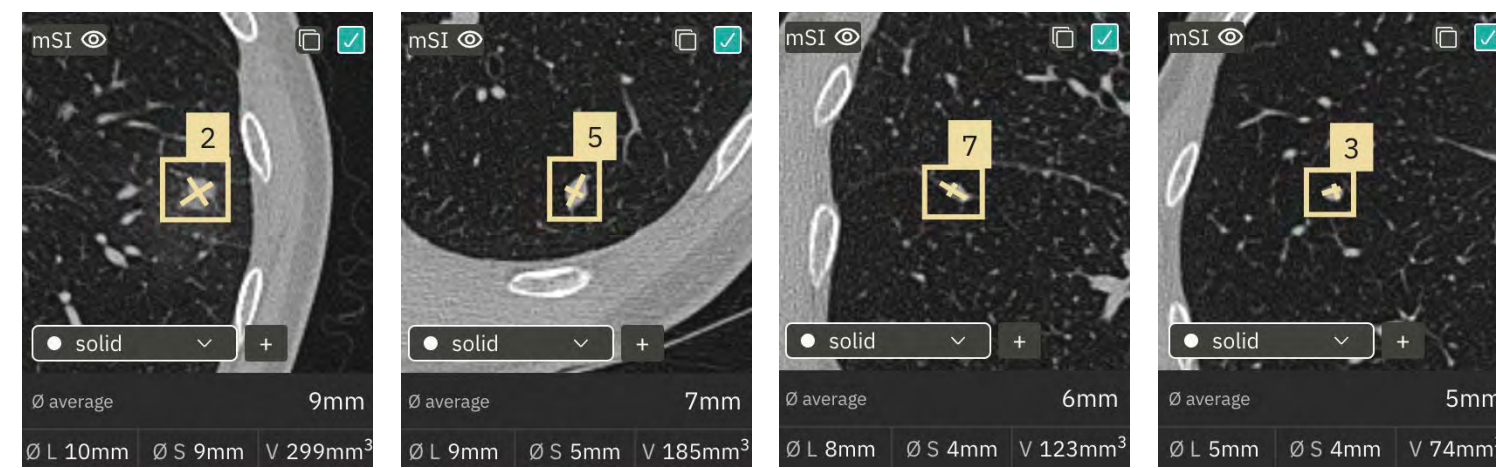
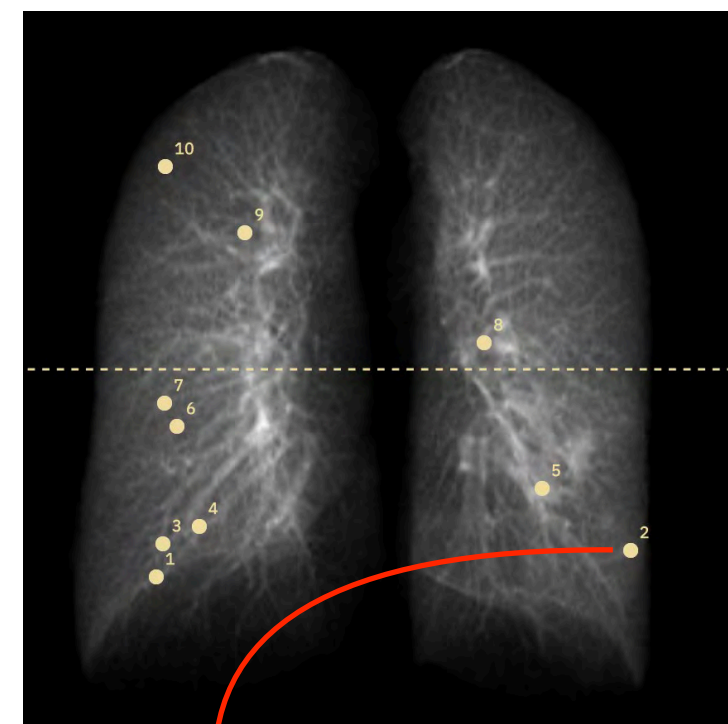
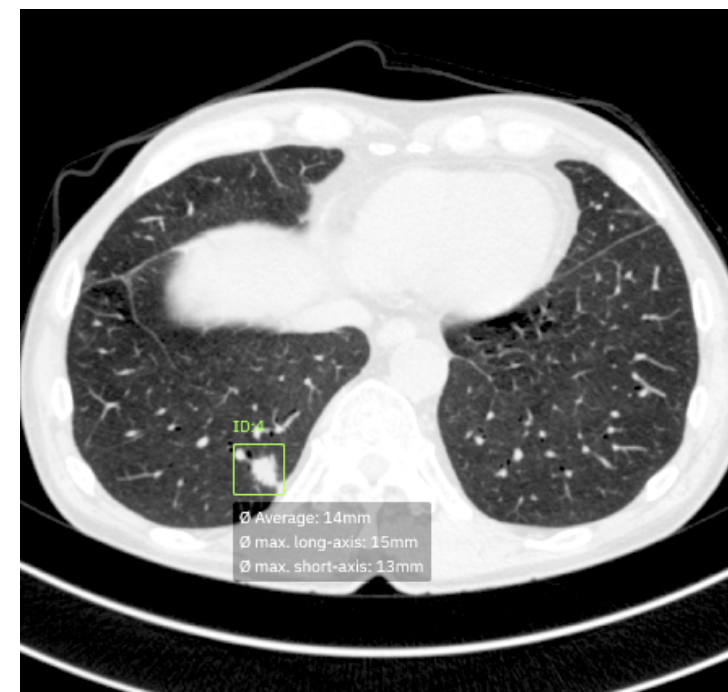
Outside the hospital: informing patients

AI for the Prediction of Obesity-Related Vascular Diseases

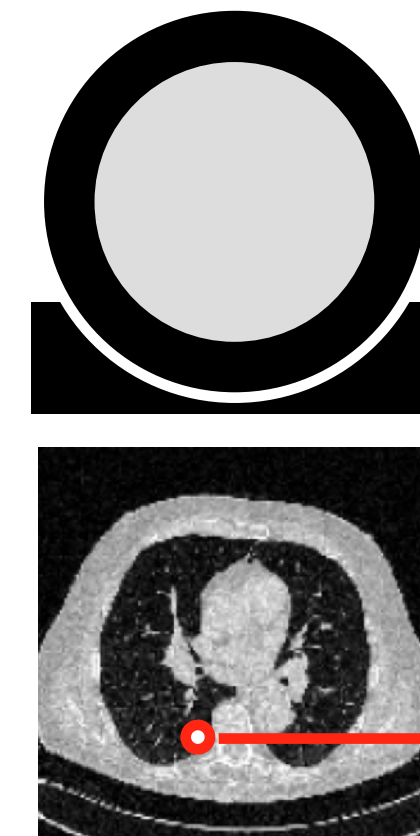
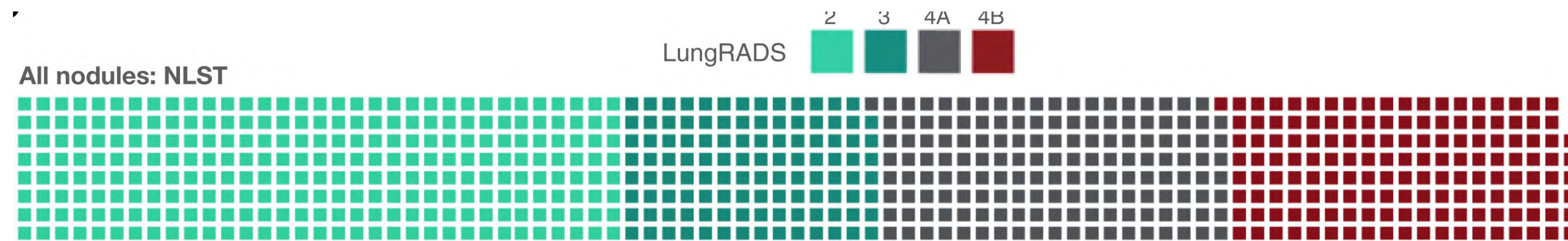
Integrating the influence of diet and lifestyle in real-time monitoring



3. Patient pathways - actionable information



Impact of malignancy prediction to augment guidelines



Follow-up
Biopsy
No action

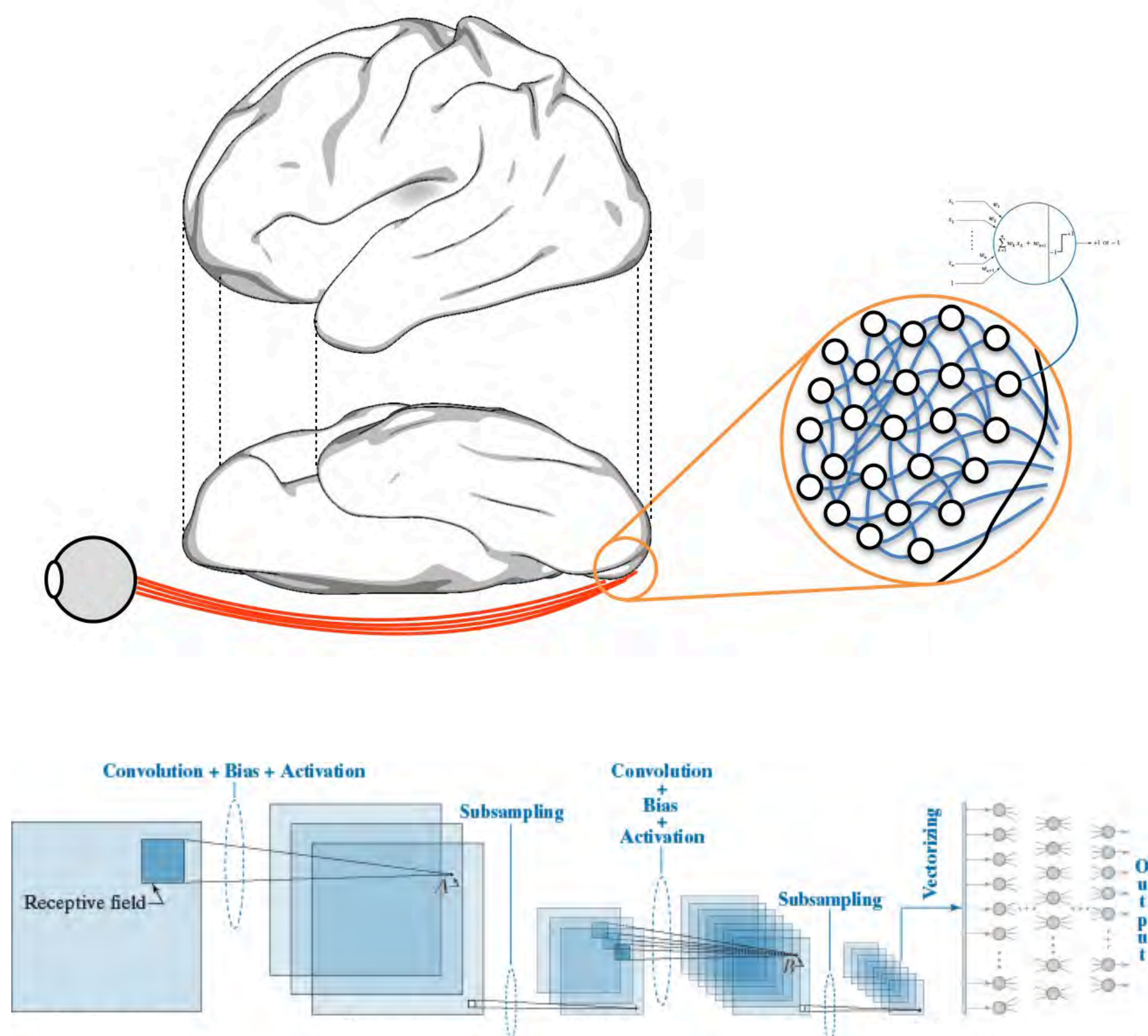
Impact on downstream management
Augmenting guidelines with AI scoring
mSI: eliminate diagnostic delay of
interval scan in 33-45%, and reducing
false positives

Adams et al. JACR 2022

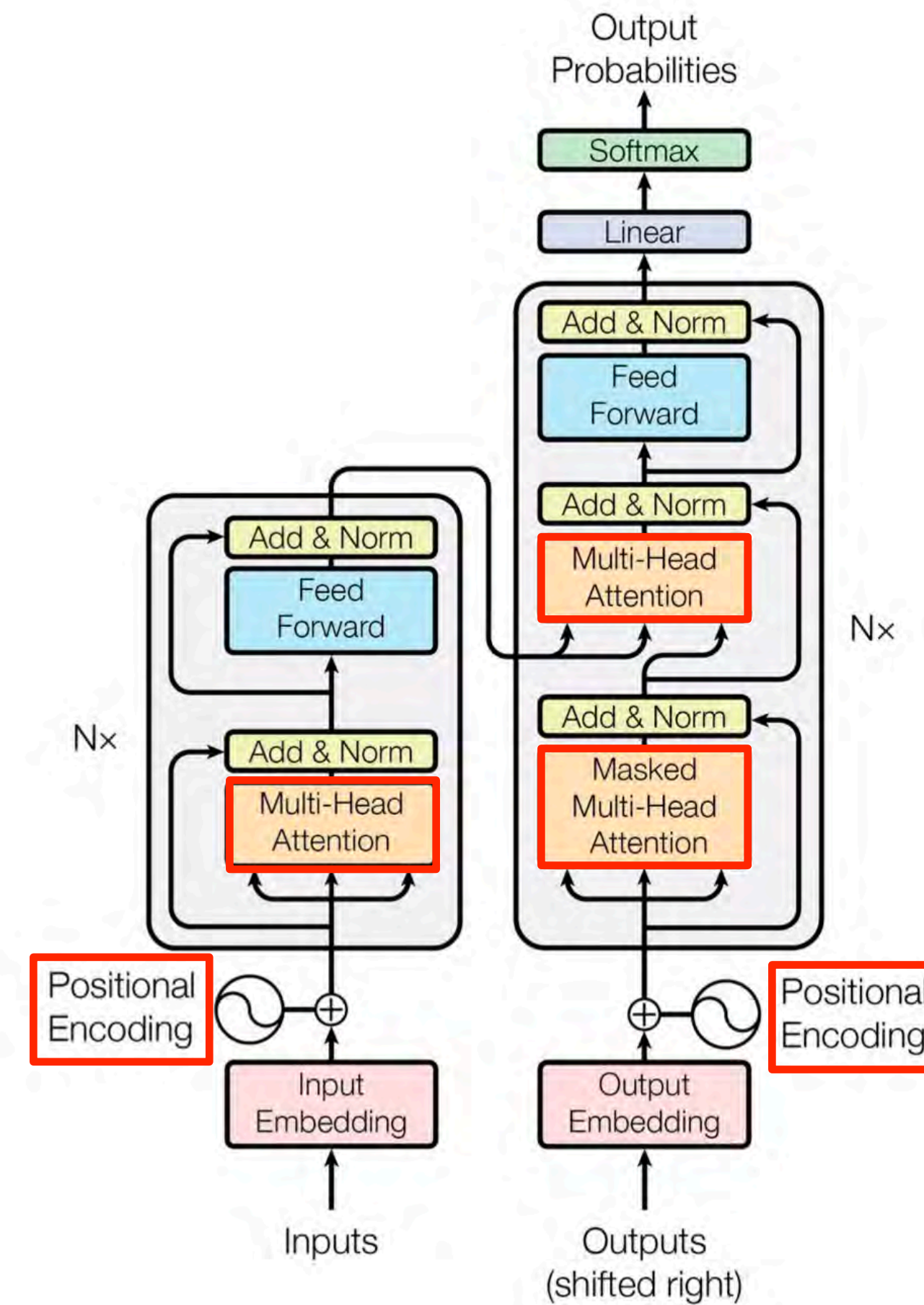
Agentic AI

Deeper integration, increased robustness & unknowns

Transformers: learning to translate large sequences



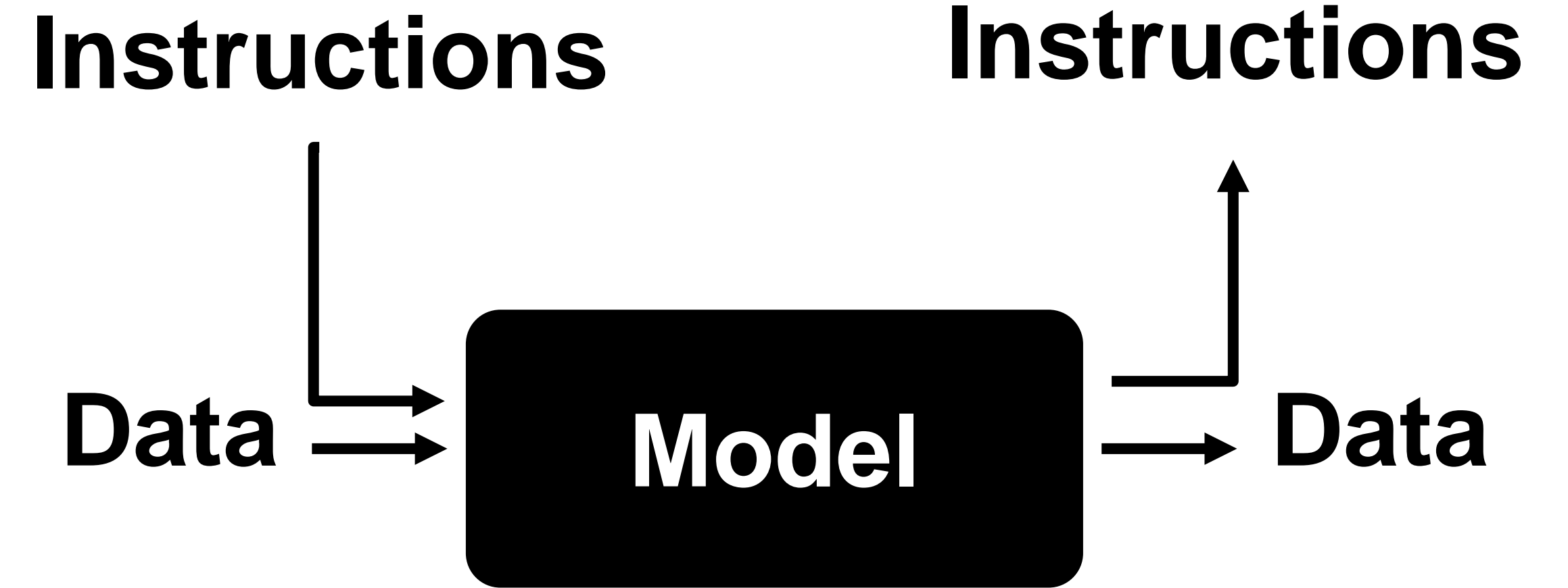
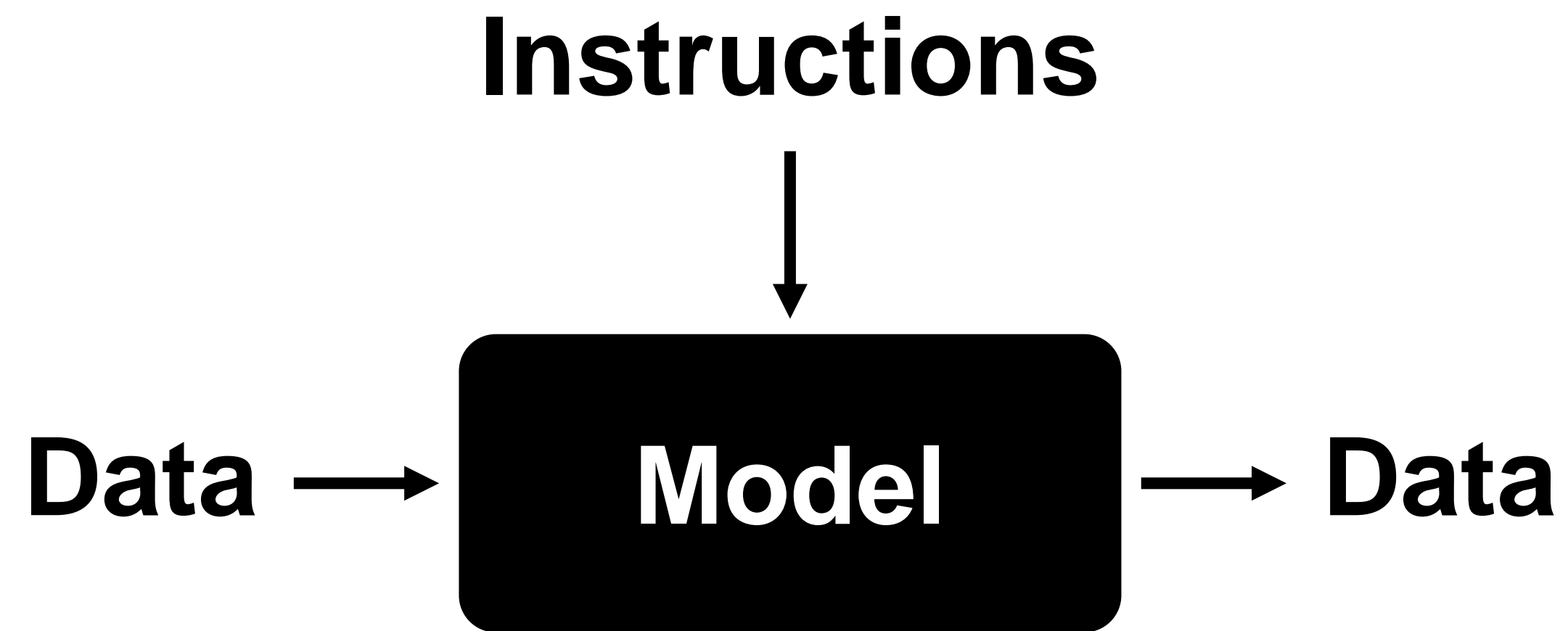
Convolutional neural networks



Transformers

Ashish Vaswani et al. *Attention is all you need* 2017 NeurIPS - 198.960 citations

Agents - what's new?



AI Agents in Healthcare

- Multi-agent systems for healthcare (MASH) deliver workflows
- Natural language communication, planning, coordination, and reasoning
- Enabled by robust input/output in the form of language

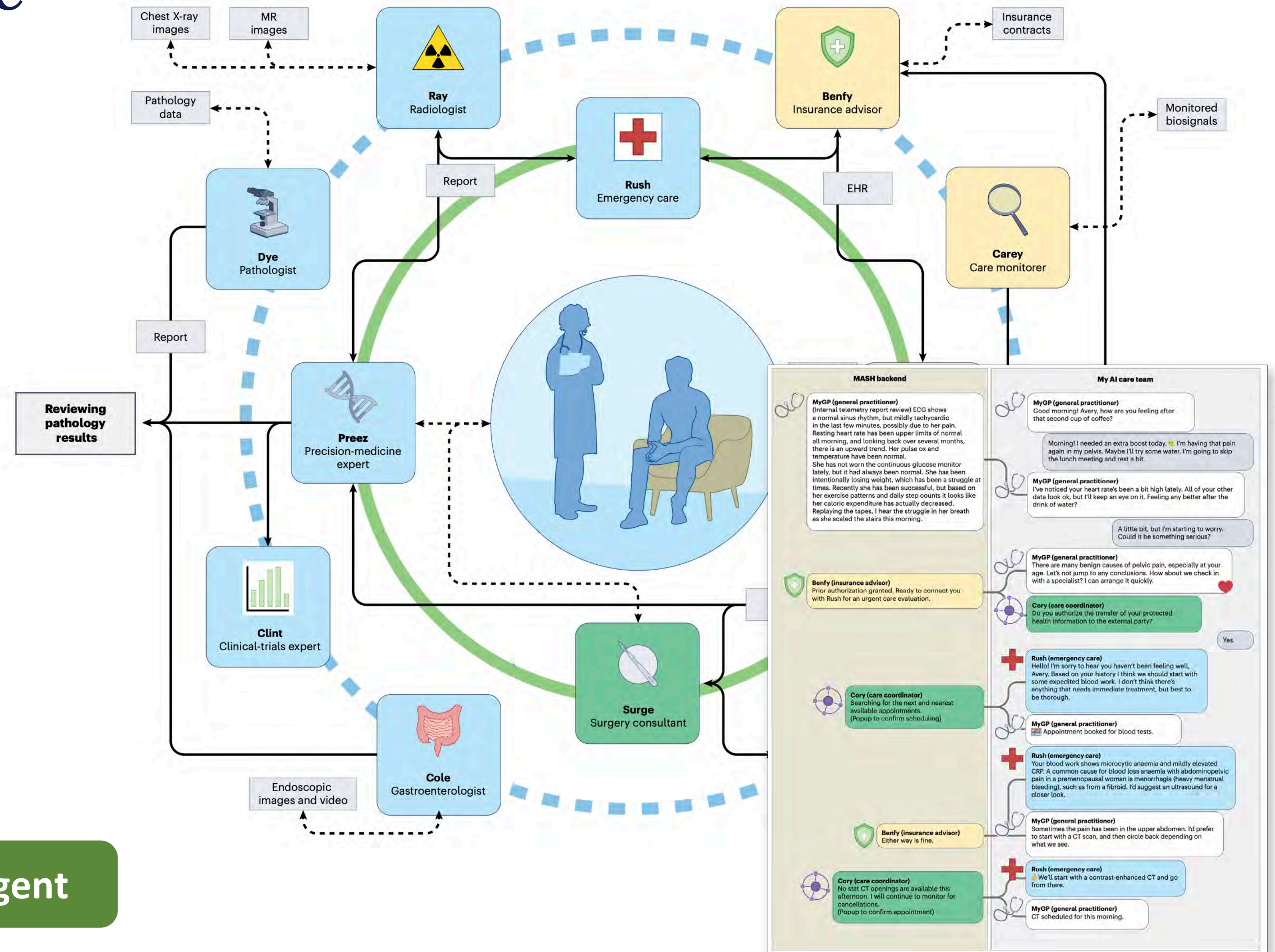
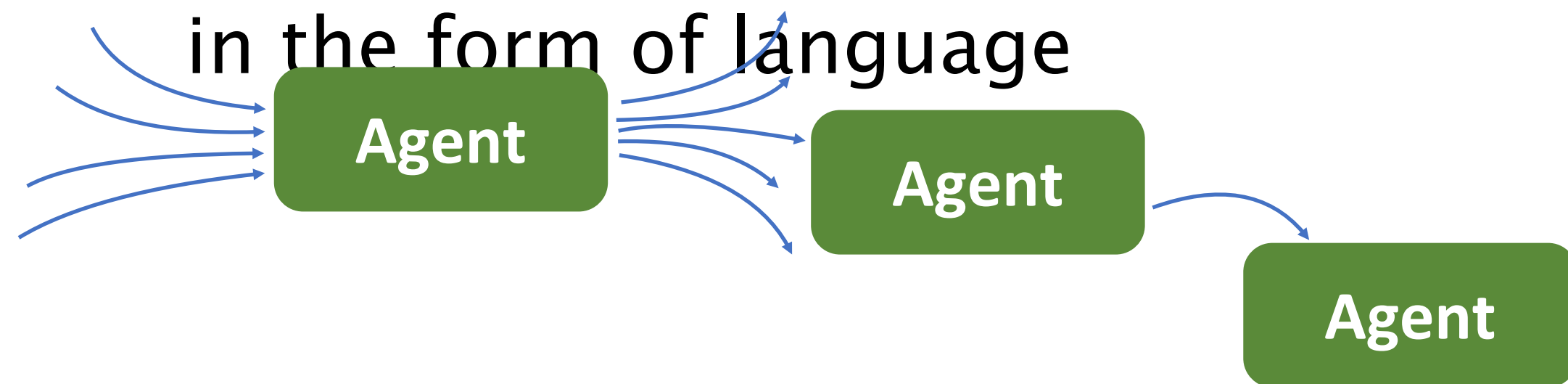
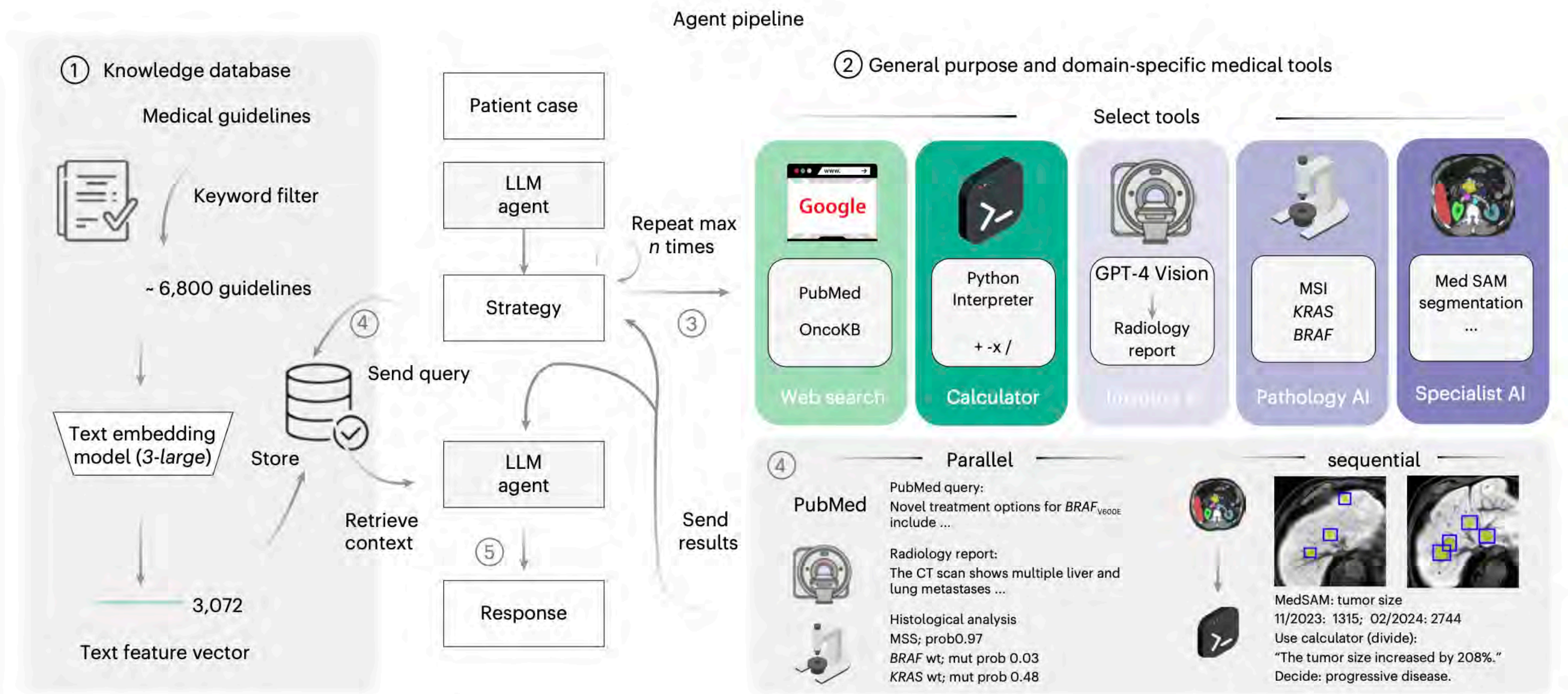


Figure from Moritz et al. April 2025 *Nature* doi.org/10.1038/s41551-025-01363-2

Agents: LLMs give instructions to other LLMs

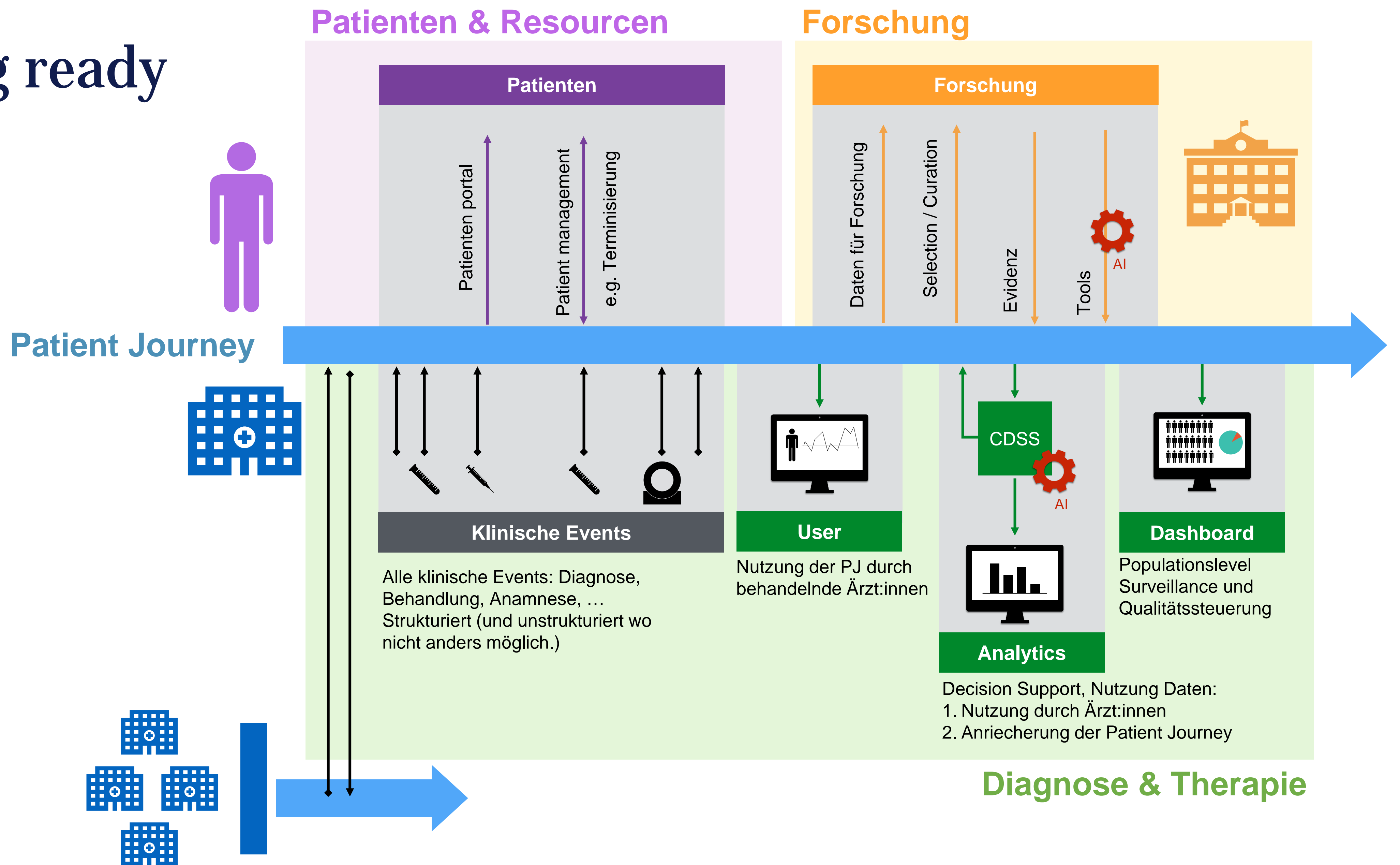


Ferber et al. Jun 2025 Nature Cancer

Implementation

Strategy & Capacity Building

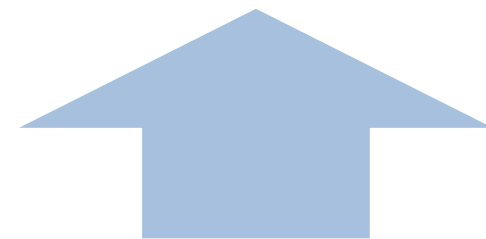
Getting ready



Strategy - capacity building - streamlining

Capacity to scale

- ▶ **Streamlined processes** for purchasing: security, data protection, compliance checks
- ▶ **Exploit repeats** (platforms, platforms, platforms)
- ▶ **AI governance** processes ~ post-market surveillance

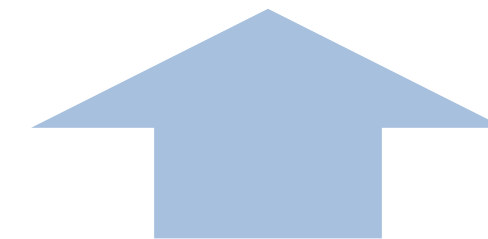


Who

- ▶ Purchasing, IT, data protection officer
- ▶ Management

Capacity to identify usecases

- ▶ **Identify target usecases** don't be too narrow, but have KPIs
- ▶ **Bridge disciplines** (integrated diagnostics, MTB, ...)
- ▶ **Levels of use:** patient, department, institution

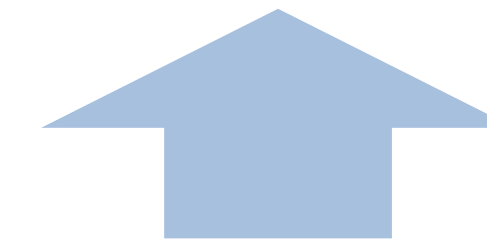


Who

- ▶ Domain experts & specialty leads
- ▶ Established interdisciplinary groups

Capacity to change workflows

- ▶ **Review workflows** and test alternatives
- ▶ **Test, iterate** and collaborate with vendors
- ▶ **Have evaluation measures**

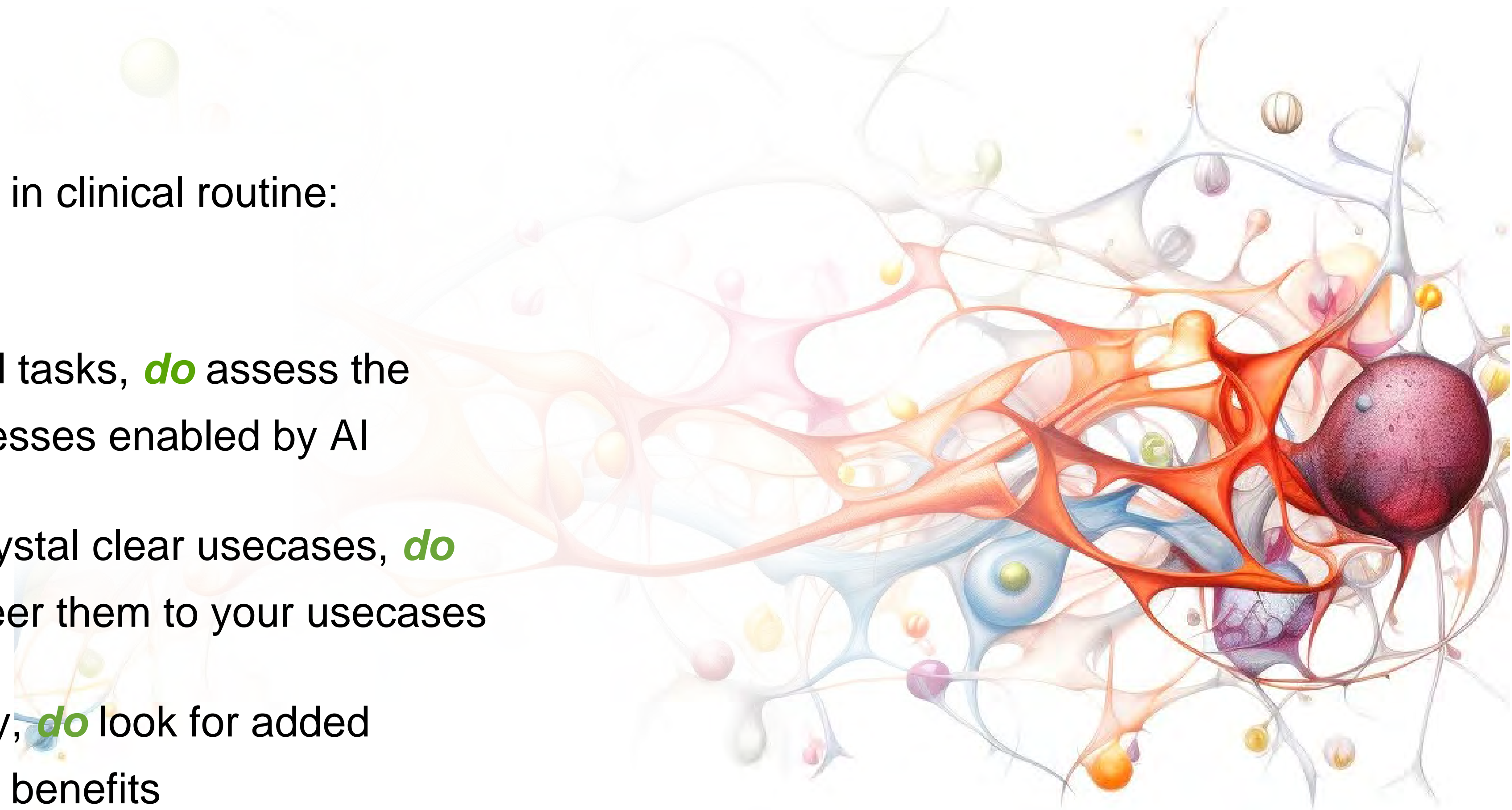


Who

- ▶ Specialty management
- ▶ Hospital management

Summary

- Artificial Intelligence is arriving in clinical routine: *gradually and then suddenly**
- **Don't** only automate individual tasks, **do** assess the feasibility of entirely new processes enabled by AI
- **Don't** wait for products and crystal clear usecases, **do** work with manufacturers to steer them to your usecases
- **Don't** get hung-up in efficiency, **do** look for added patient value and system level benefits



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*Ernest Hemingway, *The Sun Also Rises*