

DeepRadiology

REDEFINING MEDICAL PRACTICE THROUGH ARTIFICIAL INTELLIGENCE™

Something amazing is coming.....

KEEP ME UPDATED

CAREERS

Tartalom

- Miért én vagyok az előadó?
- Mit nevezek mesterséges intelligenciának?
- Dotcomlufi vagy valóság?
- Jogi-etikai-szabályozási dilemmák
- „Sehallselát Dömötör buta volt, mint hat ökör”
avagy AI-ember interakció
- Tipikus AI hibák




Viktor Gál MD., Ph.D.

Presenter

Disclosure

Radiologist/researcher specialized in MSK and neuro MRI (Semmelweis University & Research Centre for Natural Sciences)

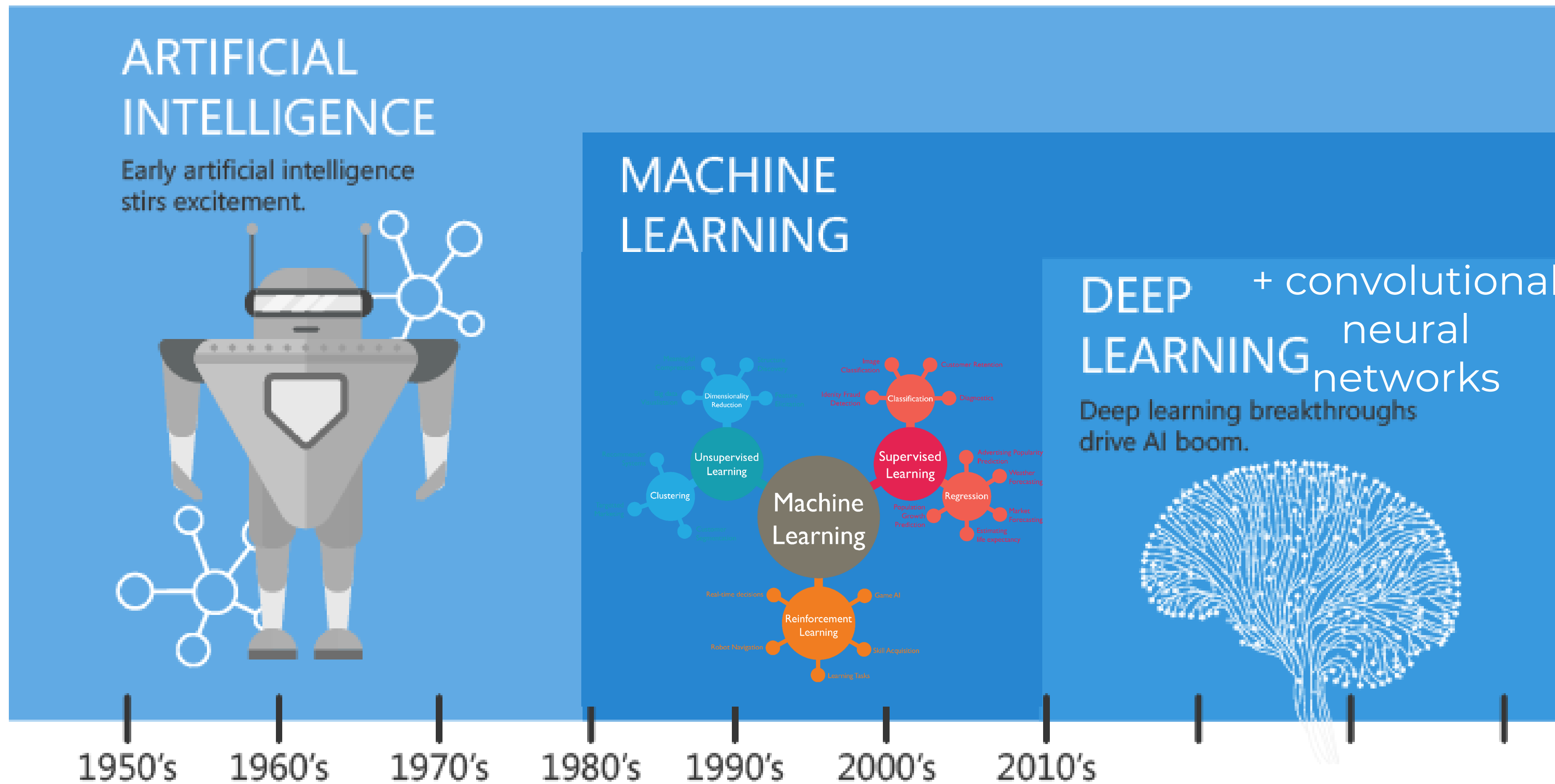
CEO of  startup, MSK MRI diagnostics automation
ORTHOPRED

Afib, owner/user of a

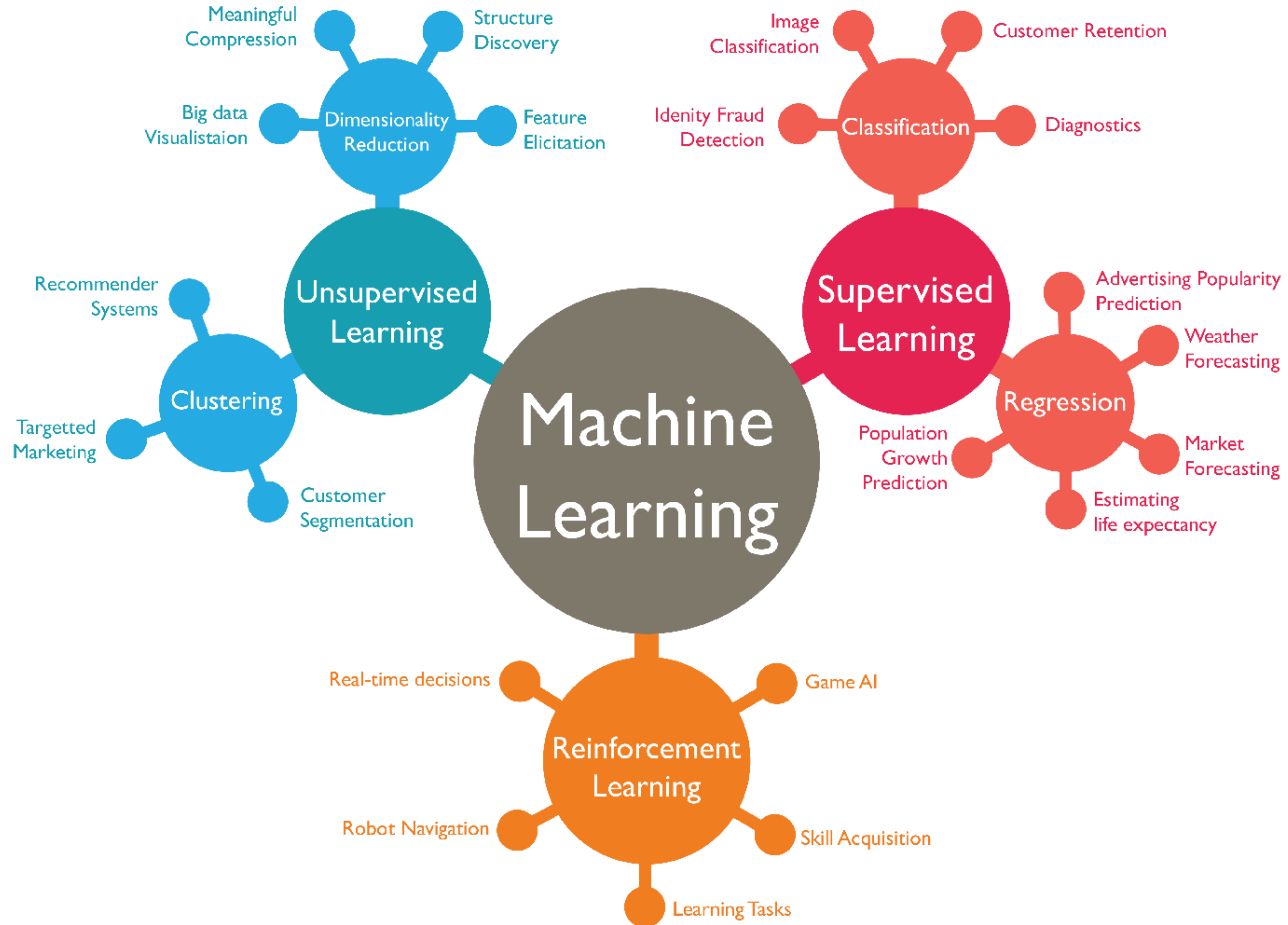


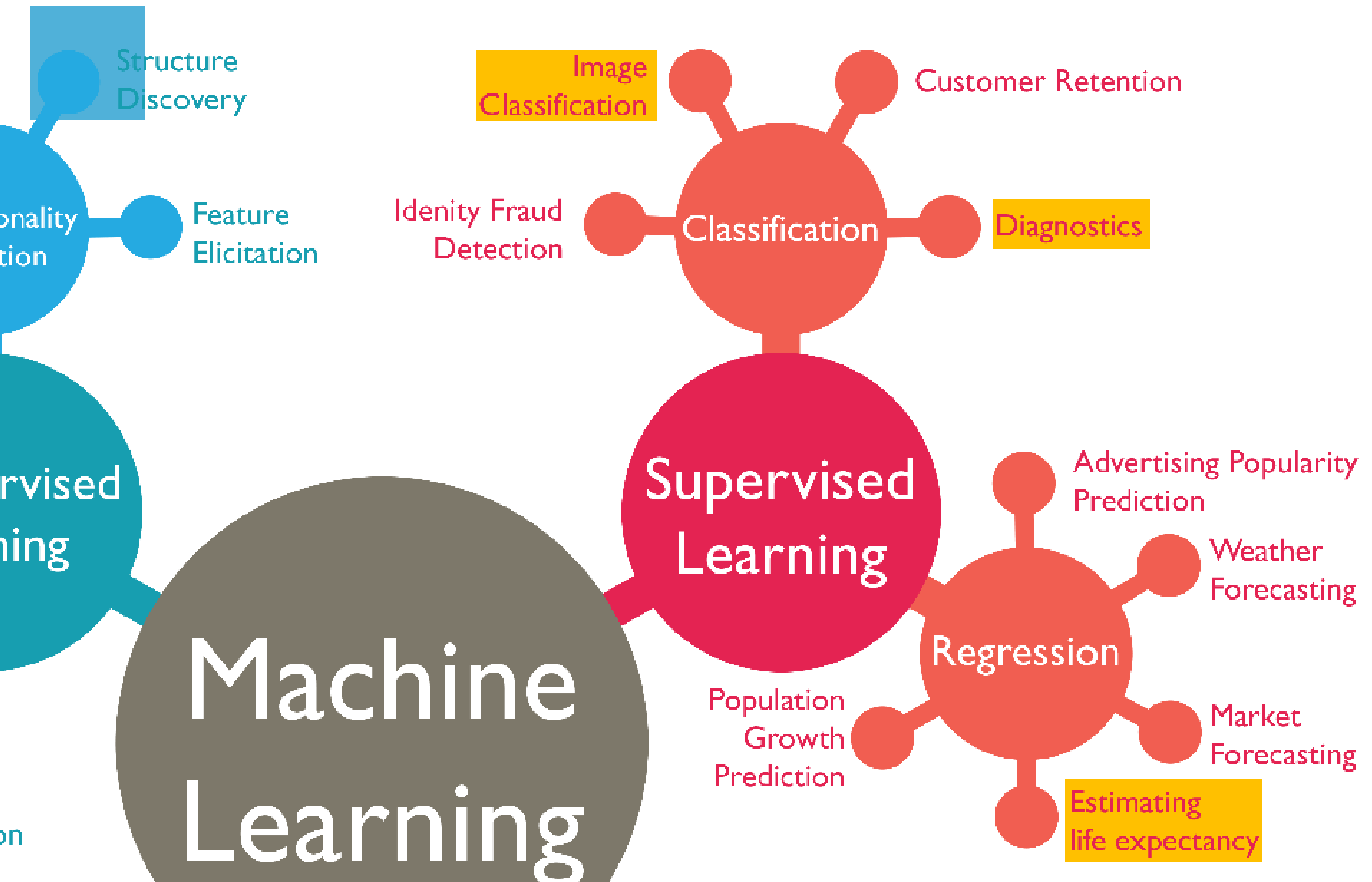


ARTIFICIAL INTELLIGENCE: DEFINITION?

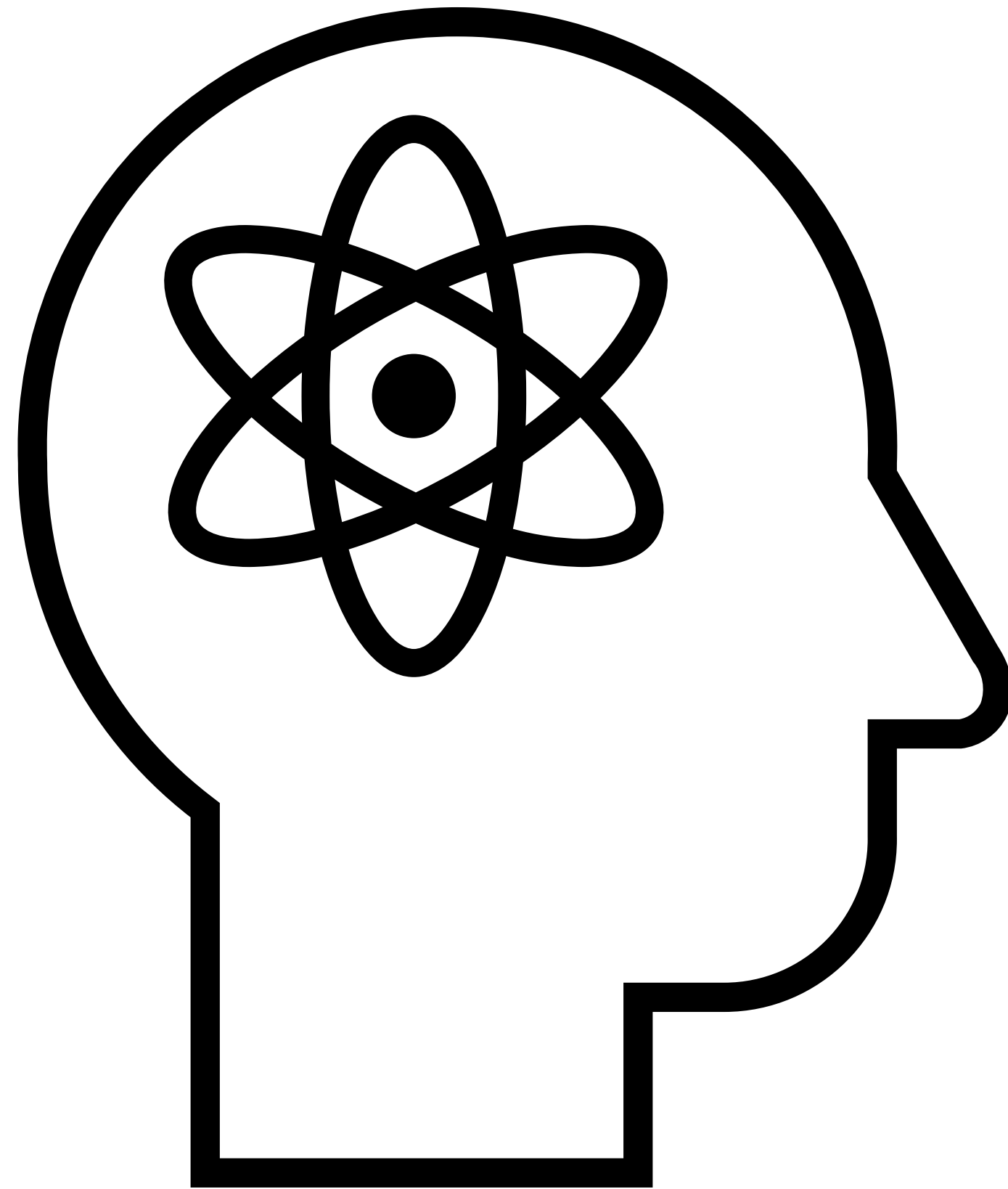


Since an early flush of optimism in the 1950's, smaller subsets of artificial intelligence - first machine learning, then deep learning, a subset of machine learning - have created ever larger disruptions.





IS IT REVOLUTIONARY? REQUIREMENTS?



IS REVOLUTIONARY? REQUIREMENTS?



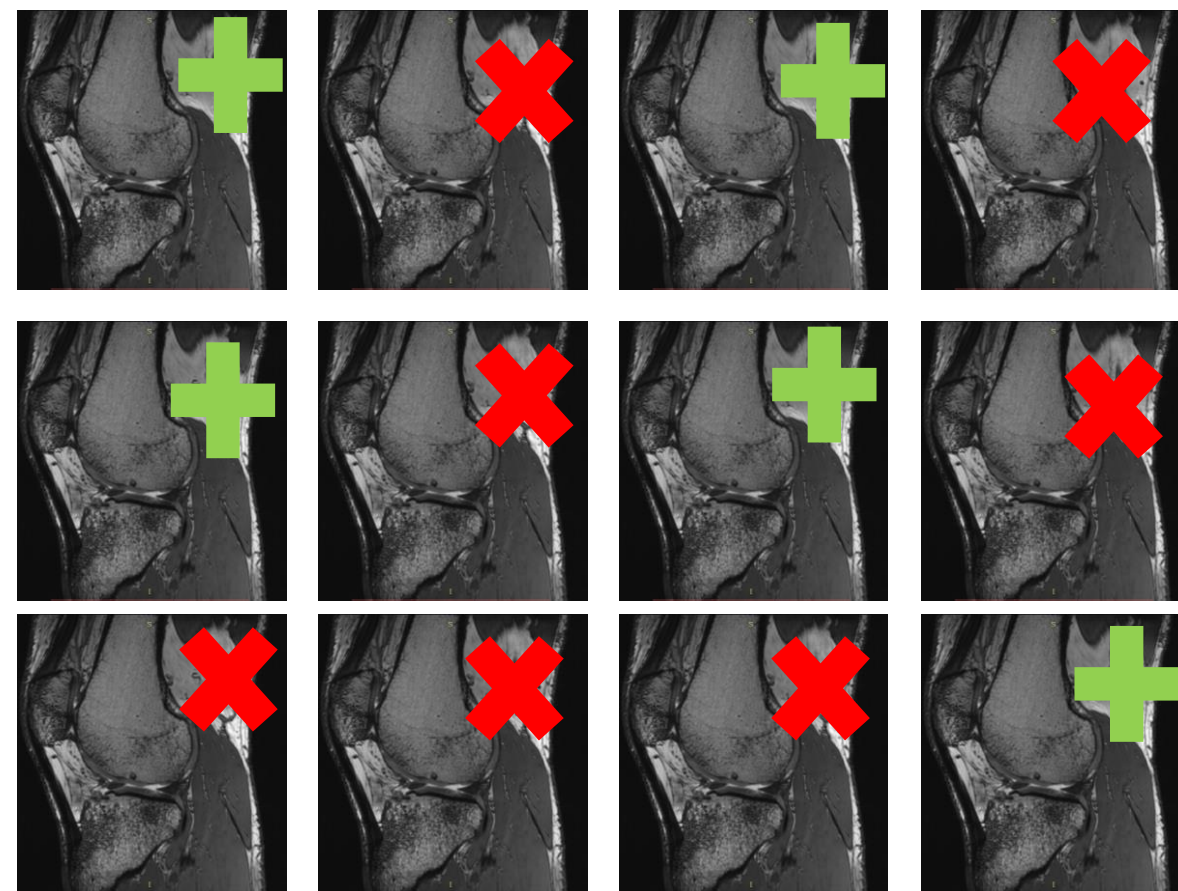
Digitalization



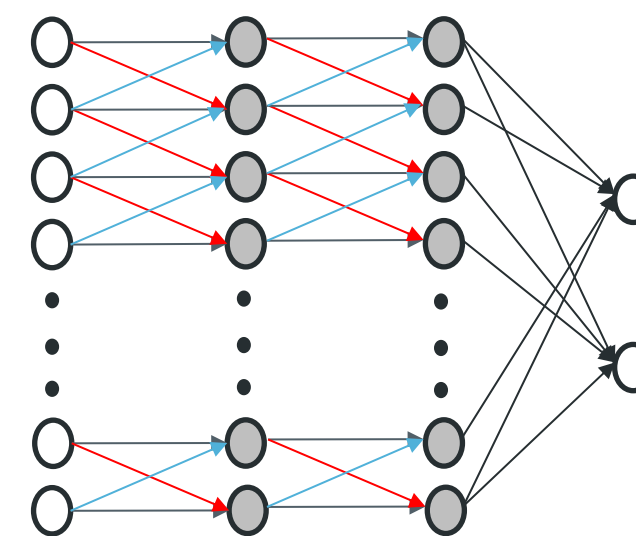
Hardware



Big database



Standard annotation

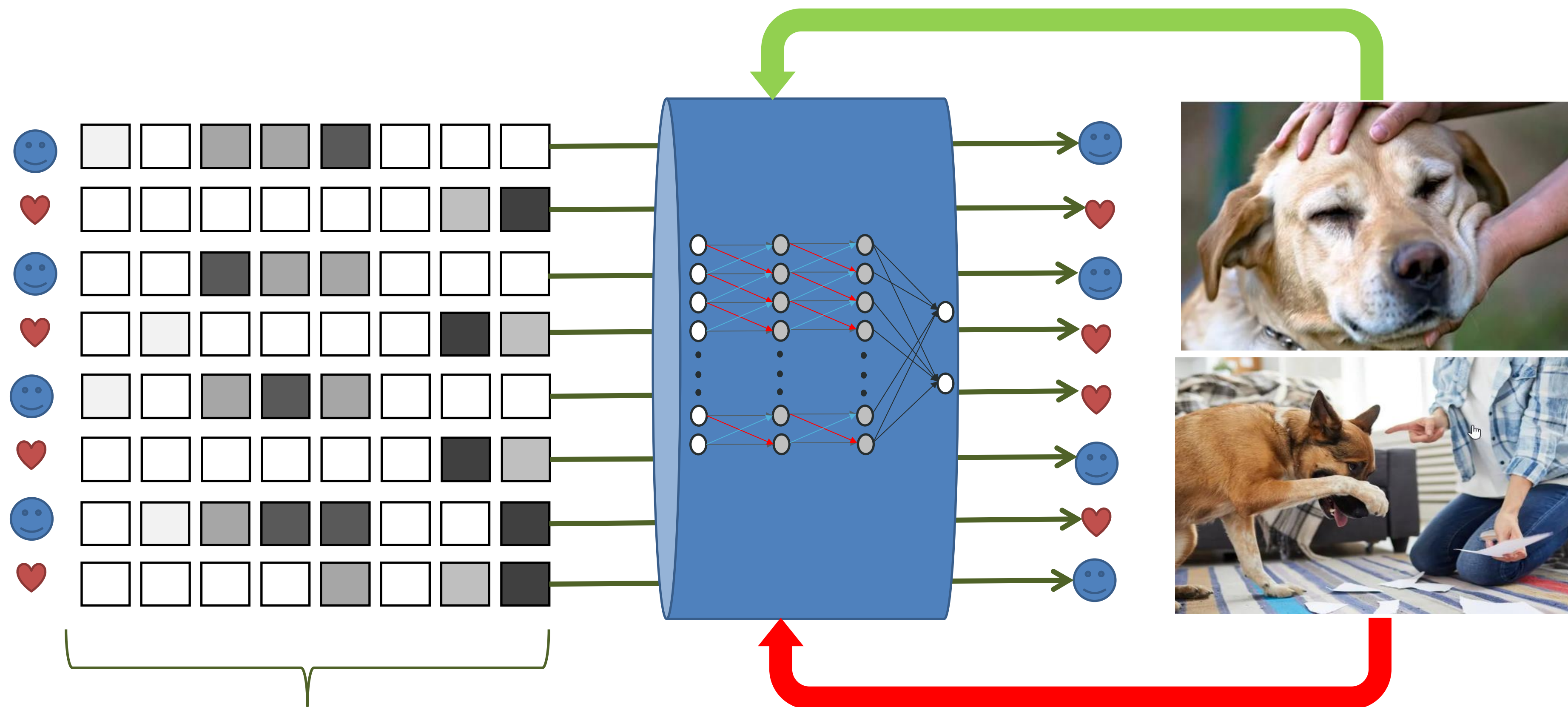


„Off the shelf” algorithms



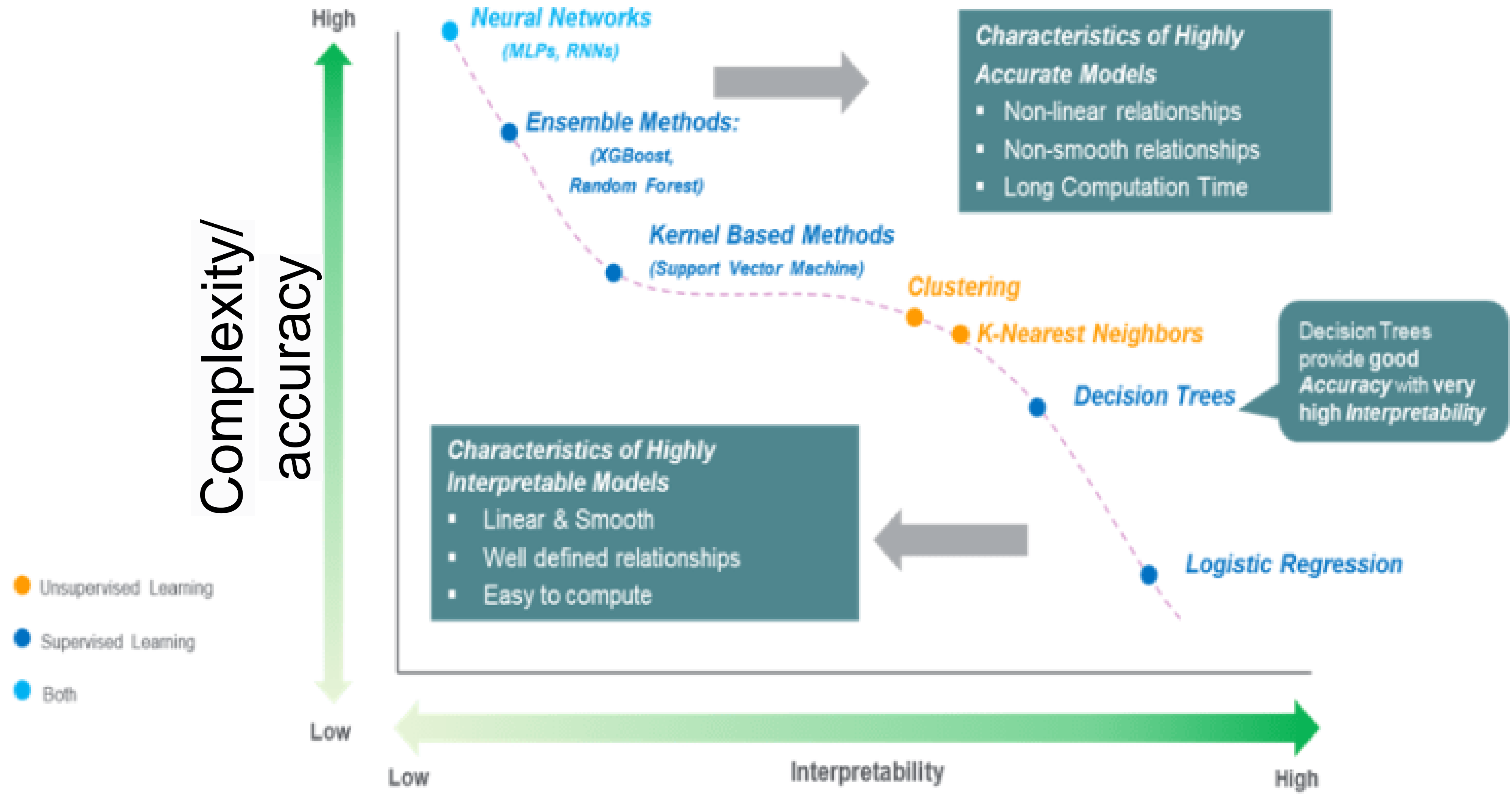
Standard protocol

SUPERVISED LEARNING OF NEURAL NETWORKS: CLASSIFICATION



Features: **raw** vs. „hand made”

PERFORMANCE / INTERPRETABILITY



VERY BRIEF History OF Artificial Intelligence

1956 Dartmouth Conference: birth of the definiton/notion of AI

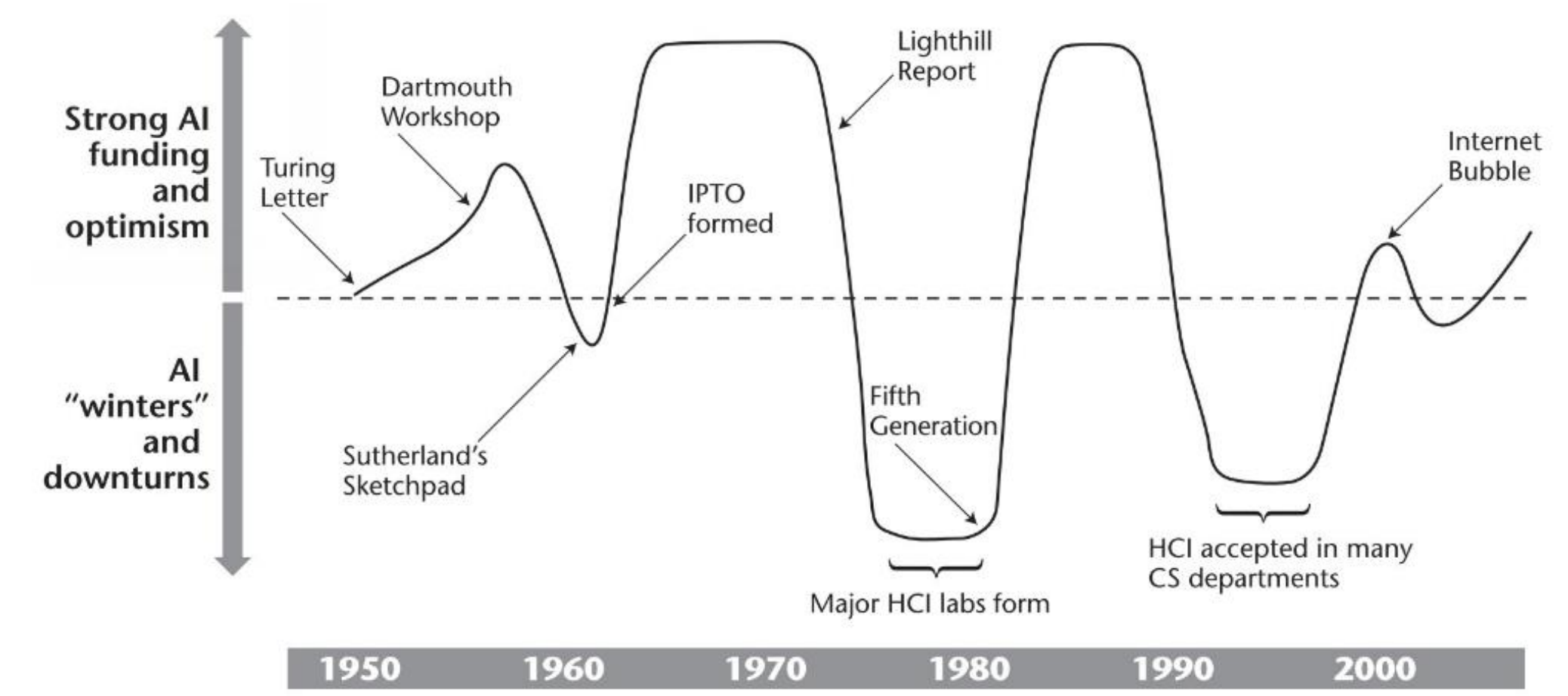
1974-2011 3x „AI winter” period with hype in between

1998 Yann LeCun (Facebook) Convolutional Neural Network (hand-written postal code reading)

AlexNet (2012) Geoffrey Hinton(Google): ImageNet contest winner(15.4 vs 26.2% error rate)

GoogLeNet (2015)

Microsoft ResNet (2015) 3.6% error rate (better than human)



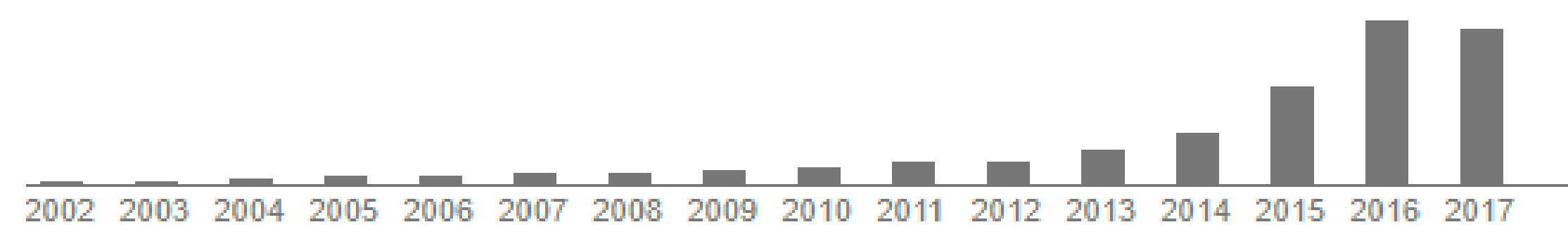
Grudin, J. (2009) AI and HCI: Two fields divided by a common focus.

Frankish, K. & Ramsey, W.M. (2017) The Cambridge Handbook of Artificial Intelligence.

Non-deep learning

Deep learning+ convolutional neural networks

Cited by 9584



Big data problems

Big Data Analysis, exploration of statistical correlations beyond human capacity

HR problems/reliability

- **Risk of misdiagnosis** ↓
- **Precision** ↑
- **Speed** ↑
- Exam planning, triage, report acceleration, second opinion, screening

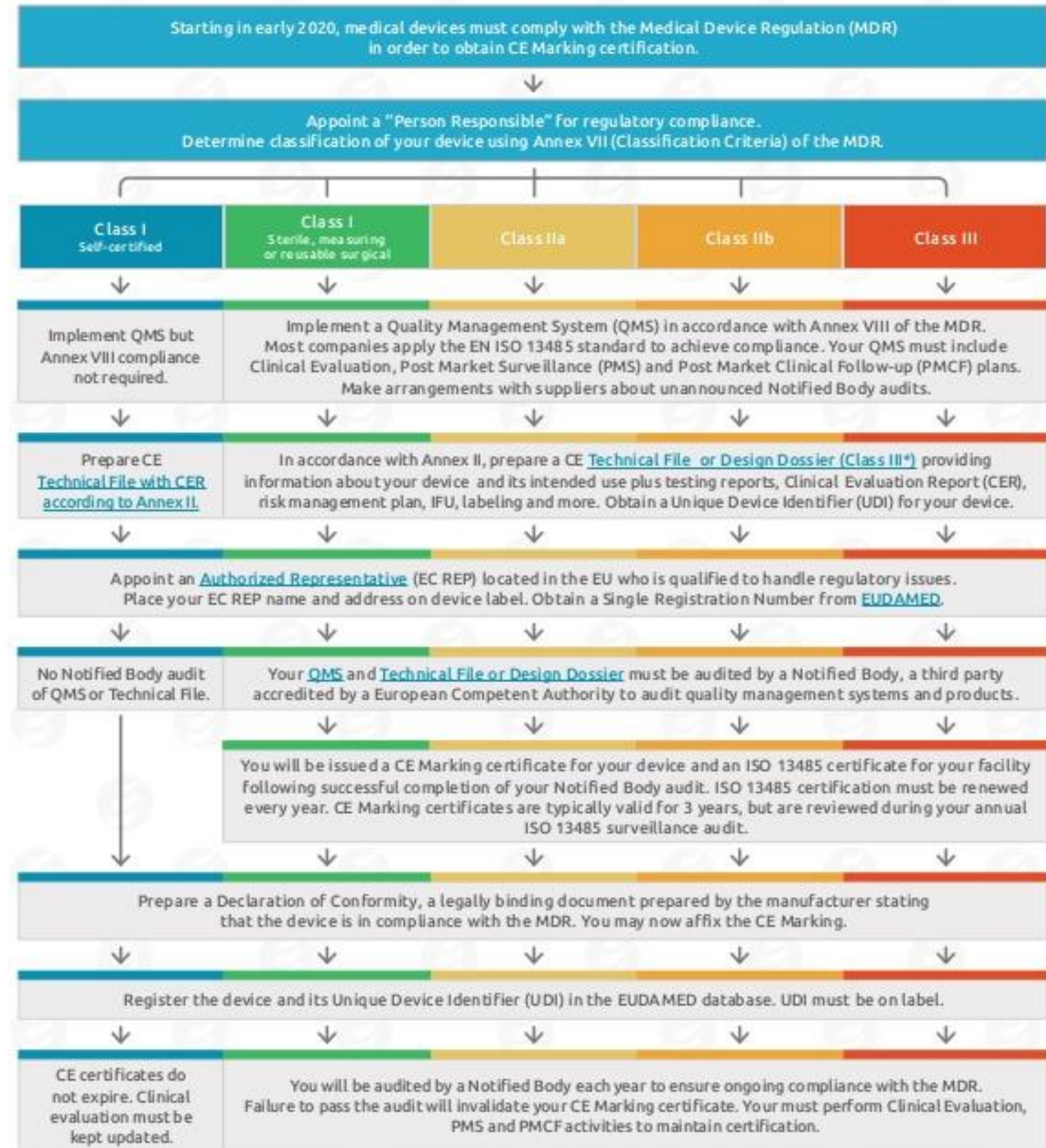
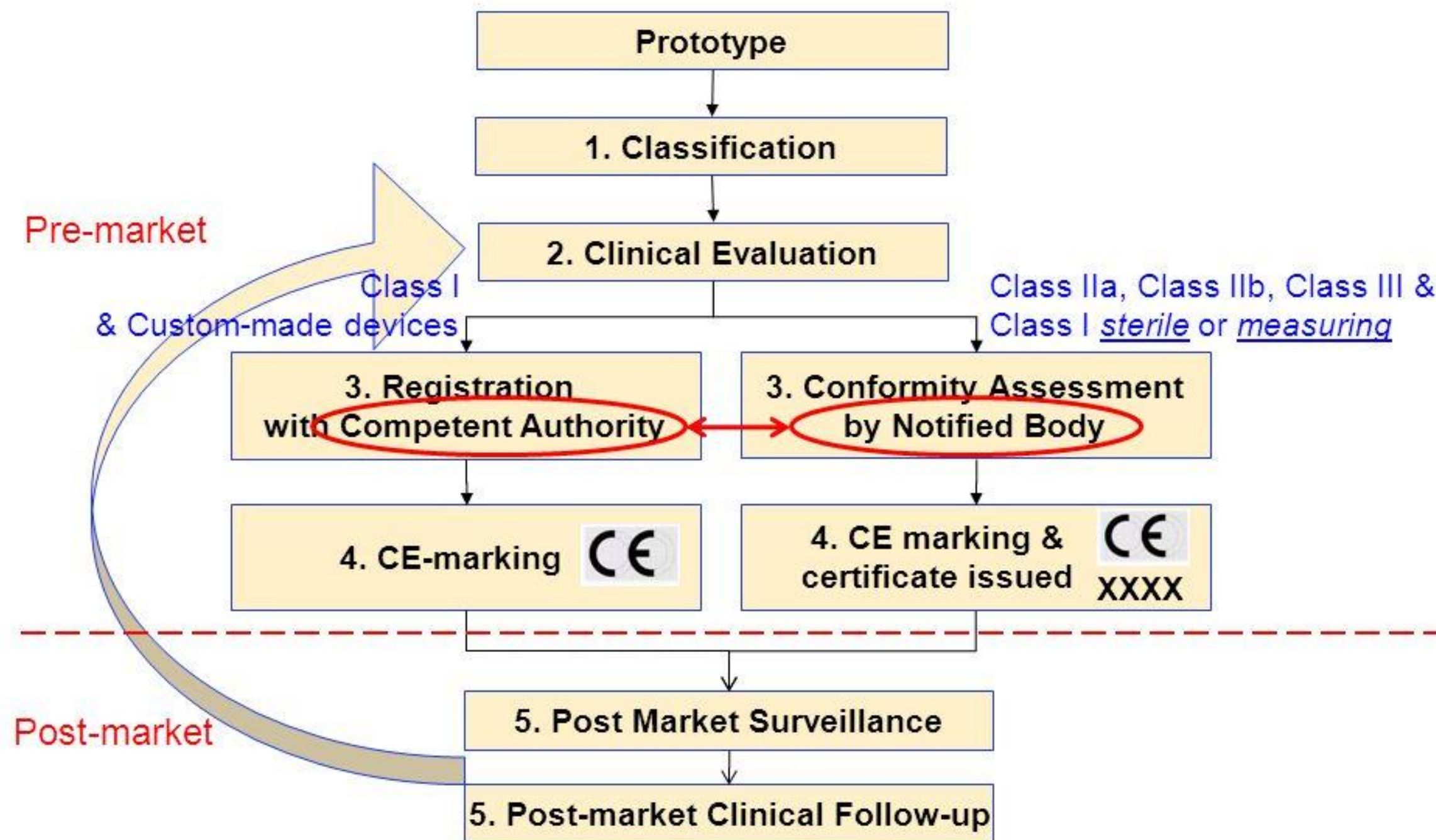
CE: REGULATORY PROC



Europe The Regulatory Process for Medical Devices

MDR Process
Effective early 2020

Lifecycle of Medical Device



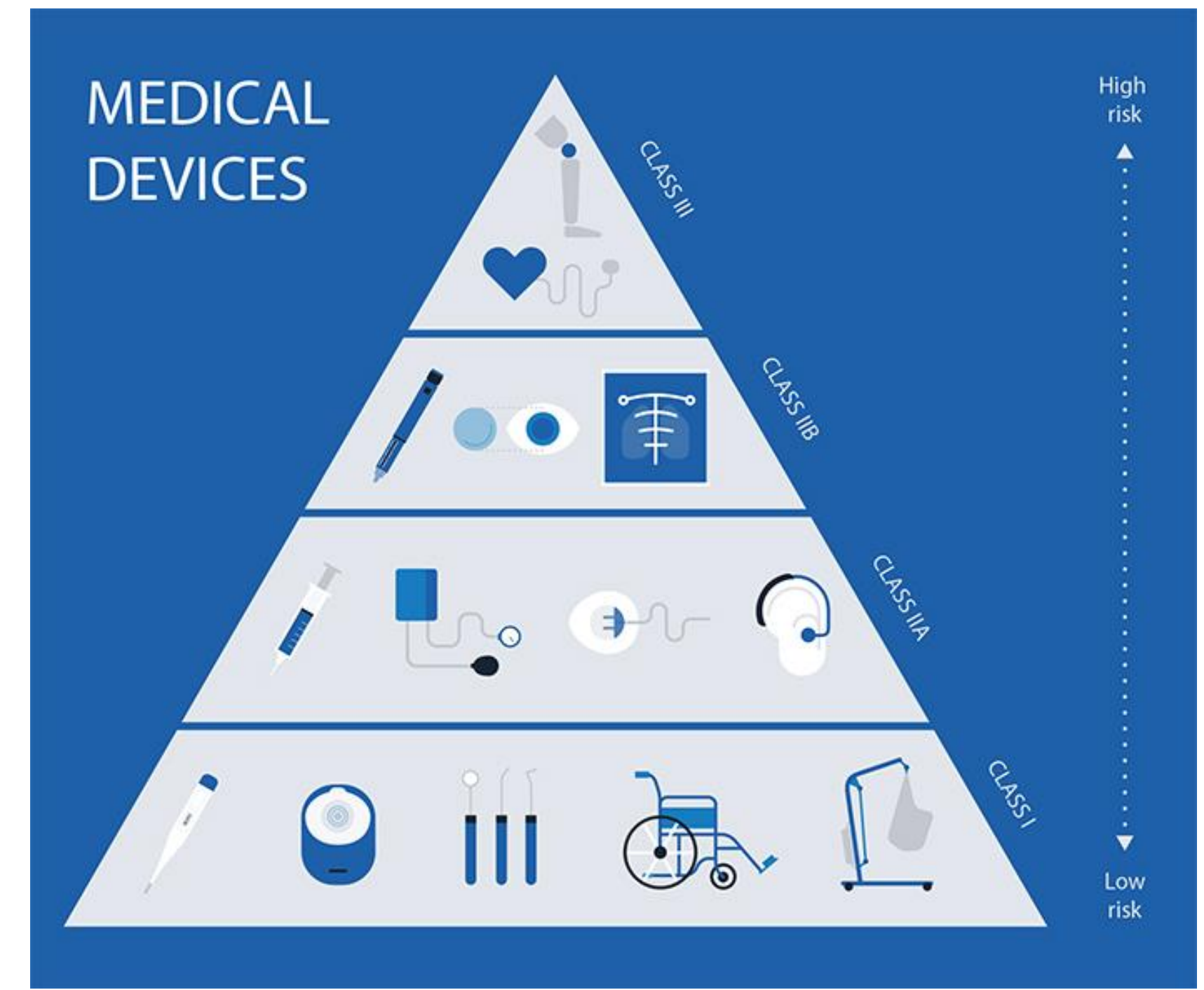
* All devices require will require clinical data. Most of these data should refer to the subject device. Clinical studies are required for Class IIb and III implants. Existing clinical data may be acceptable. Clinical trials in Europe must be pre-approved by a European Competent Authority. This is a simplified overview of the process. Your Notified Body may choose to audit your submission and request more documents, which will add time to your approval.

CE CERTIFICATION: CLASSES

- Software intended to provide information which is used to take decisions with diagnosis or therapeutic purposes is classified as **class IIa**, except if such decisions have an impact that may cause:
 - death or an irreversible deterioration of a person's state of health, in which case it is in **class III**; or
 - a serious deterioration of a person's state of health or a surgical intervention, in which case it is classified as **class IIb**.
- Software intended to monitor physiological processes is classified as **class IIa**,
 - except if it is intended for monitoring of vital physiological parameters, where the nature of variations of those parameters is such that it could result in immediate danger to the patient, in which case it is classified as **class IIb**.
- All other software is classified as **class I**.
- MDD vs **MDR**: European Medical Device Directives (soon to be replaced by the Medical Device Regulation)

CE CERTIFICATION: CLASSES

		Significance of Information provided by the MDSW to a healthcare situation related to diagnosis/therapy		
		High Treat or diagnose ~ <i>IMDRF 5.1.1</i>	Medium Drives clinical management ~ <i>IMDRF 5.1.2</i>	Low Informs clinical management (<i>everything else</i>)
State of Healthcare situation or patient condition	Critical situation or patient condition ~ <i>IMDRF 5.2.1</i>	Class III <i>Category IV.i</i>	Class IIb <i>Category III.i</i>	Class IIa <i>Category II.i</i>
	Serious situation or patient condition ~ <i>IMDRF 5.2.2</i>	Class IIb <i>Category III.ii</i>	Class IIa <i>Category II.ii</i>	Class IIa <i>Category I.ii</i>
	Non-serious situation or patient condition (<i>everything else</i>)	Class IIa <i>Category II.iii</i>	Class IIa <i>Category I.iii</i>	Class IIa <i>Category I.i</i>



<https://towardsdatascience.com/how-to-get-clinical-ai-tech-approved-by-regulators-fa16dfa1983b>

FDA APPROVAL SUBMISSION TYPES

510(K) SUBMISSION

Each person who wants to market in the U.S., a Class I, II, and III device intended for human use, for which a Premarket Approval (PMA) is not required, must submit a 510(k) submission to FDA to demonstrate that the device to be marketed is at least **as safe and effective (substantially equivalent) to a legally marketed device that is not subject to PMA**. Submitters must support their substantial equivalency claims.

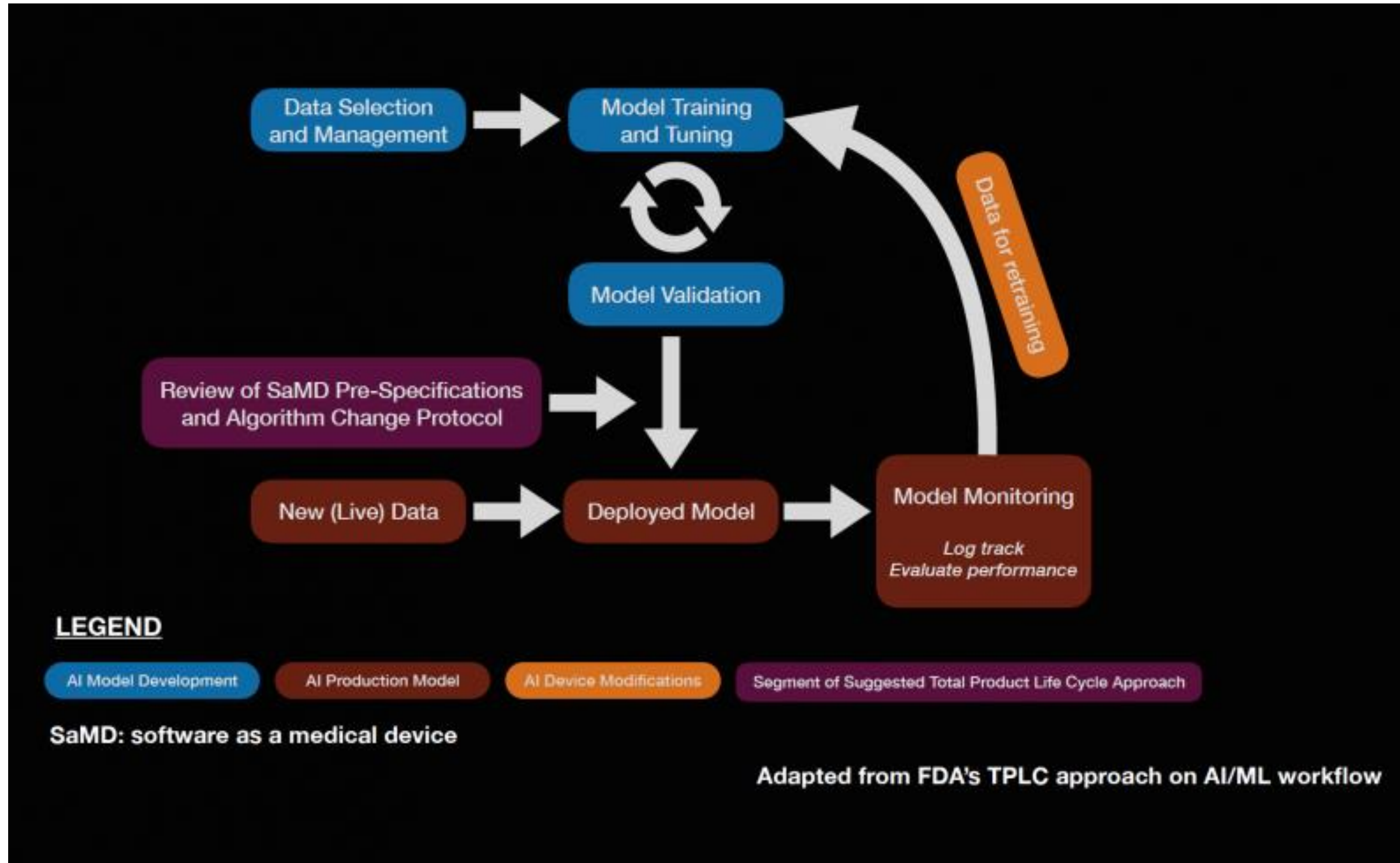
PMA

Premarket approval (PMA) is the FDA process of scientific and regulatory review to evaluate the safety and effectiveness of **Class III** medical devices, and the **most stringent** of the device marketing applications. Class III devices are those that **support or sustain human life**, are of substantial importance in preventing impairment of human health, or which **present a potential, unreasonable risk of illness or injury**. General and special controls alone are insufficient to assure the safety and effectiveness of Class III devices. PMA applications will include technical sections, usually divided into non-clinical laboratory studies and clinical investigations. PMA approval typically requires a facility inspection.

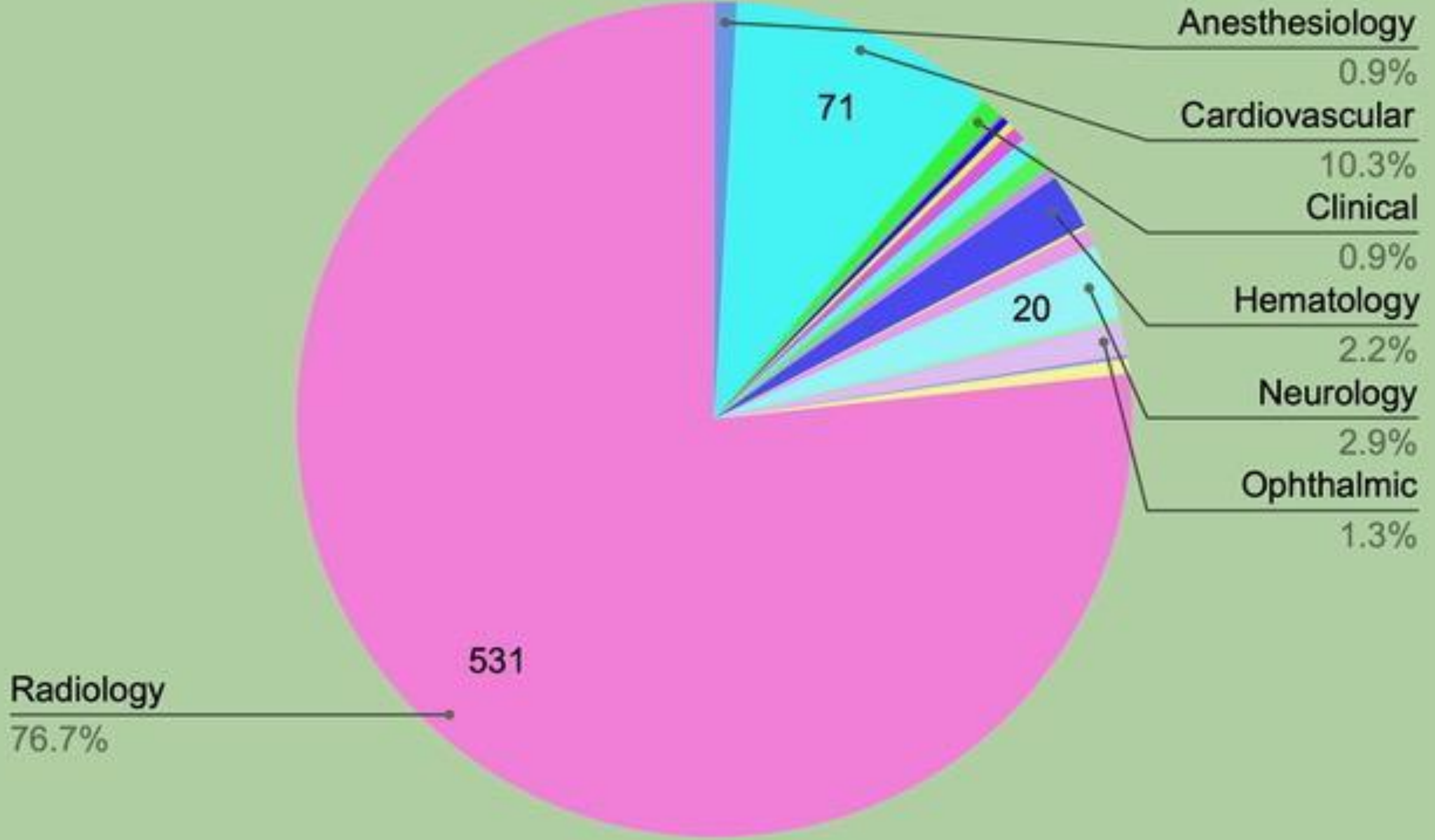
DE NOVO

The de novo pathway for device marketing rights was added to address **novel devices of low to moderate risk that do not have a valid predicate device**. Upon successful review of a de novo submission, FDA **creates a classification for the device**, a regulation if necessary, and identifies any special controls required for future premarket submissions of substantially equivalent devices. PRE-SUBMISSIONS (PRE-SUBS) Pre-submissions are made to the FDA in order to request FDA feedback. Pre-subs are used for various reasons including meeting requests, to study risk determination, for submission issues, and for FDA feedback to specific questions related to a pending submission or protocol. The main purpose of the Pre-Sub Program (previously known as the Pre-IDE Program) is to provide the opportunity for a sponsor to obtain FDA feedback prior to an intended submission of an IDE or marketing application. The Pre-Sub Program can also provide a mechanism for the Agency to provide advice to sponsors who are developing protocols for clinical studies for which an IDE would not be required, such as studies of non-significant risk (NSR) devices or for clinical studies conducted outside of the U.S. to support future U.S. marketing applications. Consequently, the Pre-Sub program can provide an efficient path from device concept to market while facilitating the agency's goal of fostering the development of new medical devices.

TPLC adaptive algorithms require a total product lifecycle (TPLC) regulatory approach vs. „locked algorithm”



FDA CLEARED ALGORITHMS



• S Benjamins, P Dhunoo & B. Meskó
npj Digital Medicine volume 3,
Article number: 118 (2020)

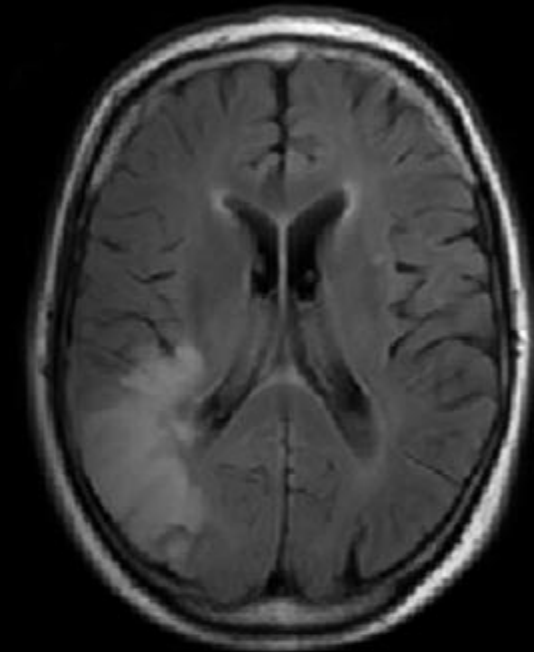
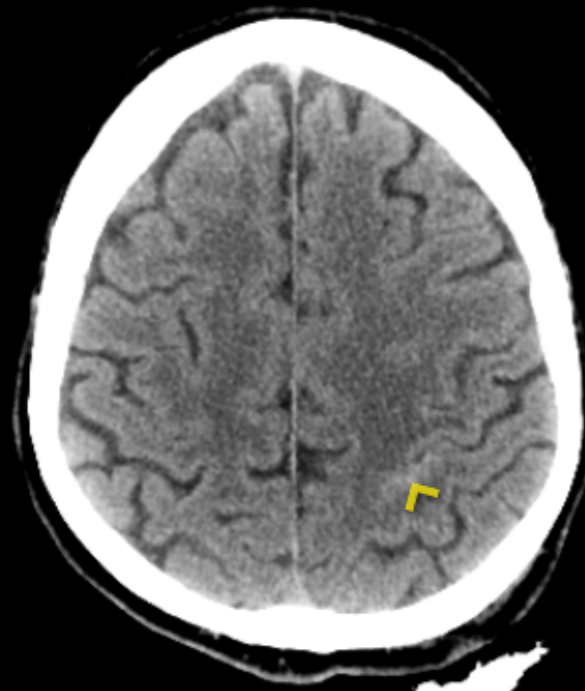


Cranial AI

STROKE

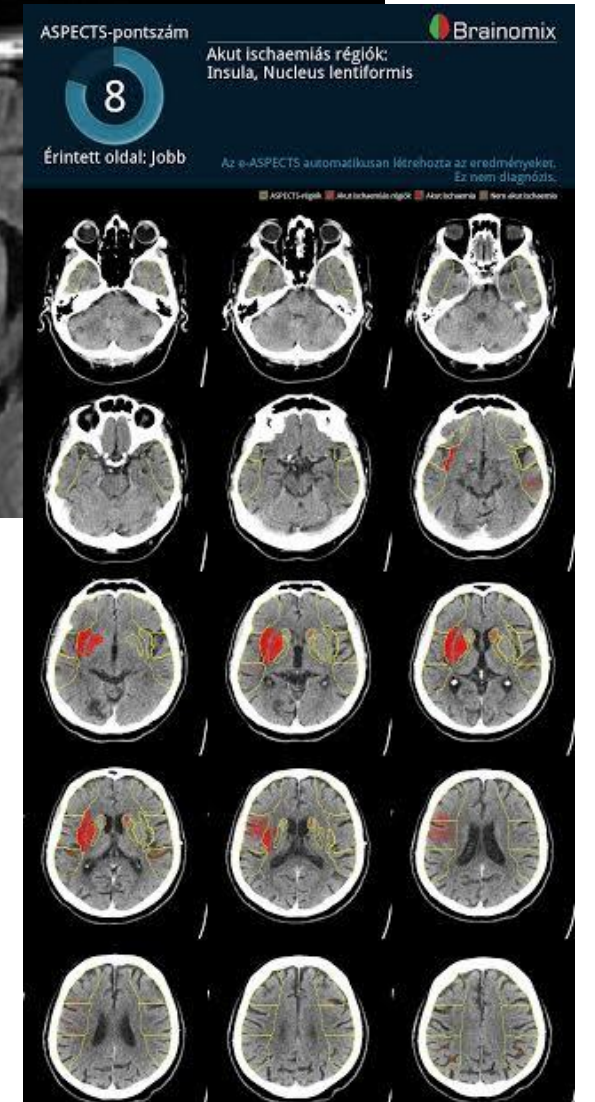


FDA clearance:
non-contrast CT ICH
detection
Sensitivity: 93.6% (95% CI:
86.6%-97.6%)
specificity: 92.3% (95%
CI: 85.4%-96.6%)."



Viz L
Using artificial intelligence to automatically

FDA CLEARANCE A



Köszönet: Martos János 2018

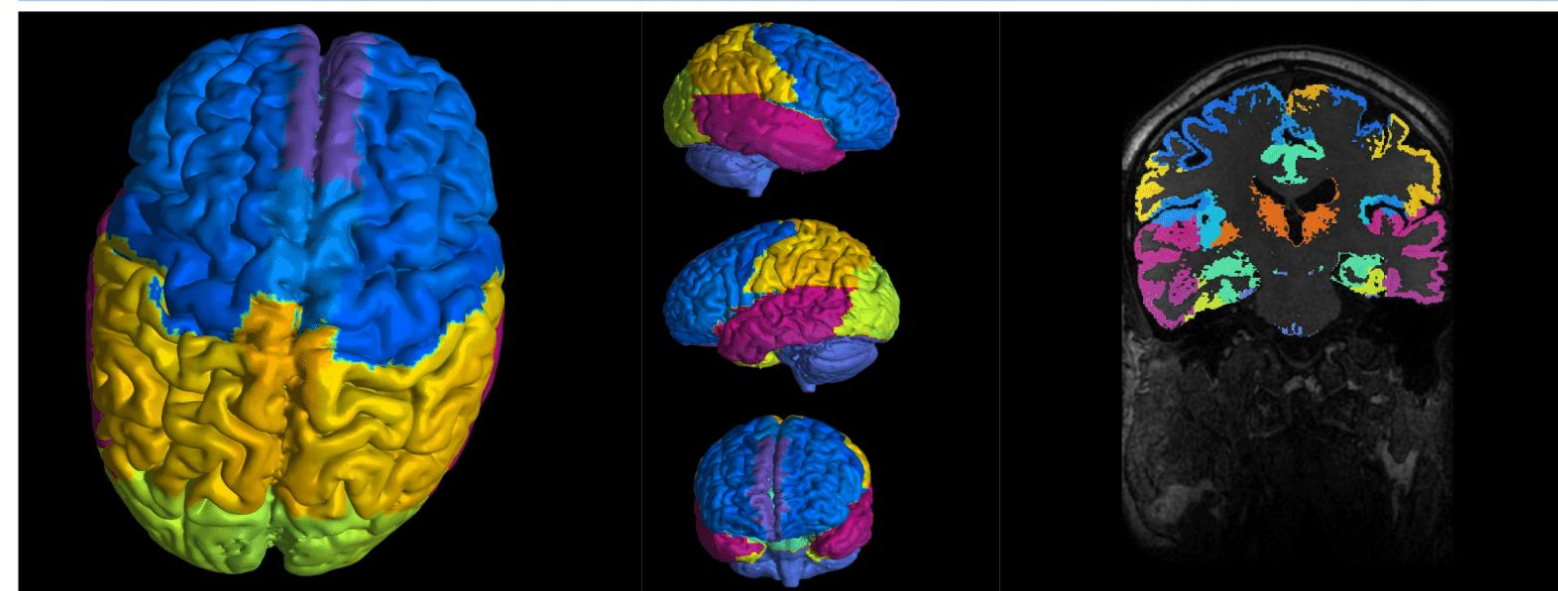
SEGMENTATION/VOLUMETRY



http://quibim.com
info@quibim.com

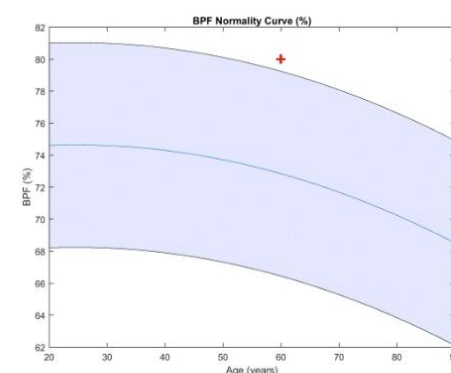
Brain Volumetry

Imaging Center	ERESCANER H U POLITECN...	Patient Name	Glioblas
Modality	MR	Patient ID	
Study Description	RM111 - RM CEREBRAL	Patient Sex	M
Study Date	16/03/2016	Birthdate	



Brain Parenchyma Fraction (%)	80.02	
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	Absolute Volume (mL)	Relative Volume (% of ICV)
Gray Matter	759.70	41.82
White Matter	693.73	38.19
CSF	362.87	19.97



	Absolute Volume (mL)		Relative Volume (% of ICV)	
	Right	Left	Right	Left
Hippocampus	5.58	4.63	0.30	0.25
Frontal	55.13	57.04	3.03	3.14
Amygdala	1.89	1.20	0.10	0.06
Temporal	61.67	43.08	3.39	2.37
Precentral	9.12	9.10	0.50	0.50
Cerebellum	51.11	50.45	2.81	2.77

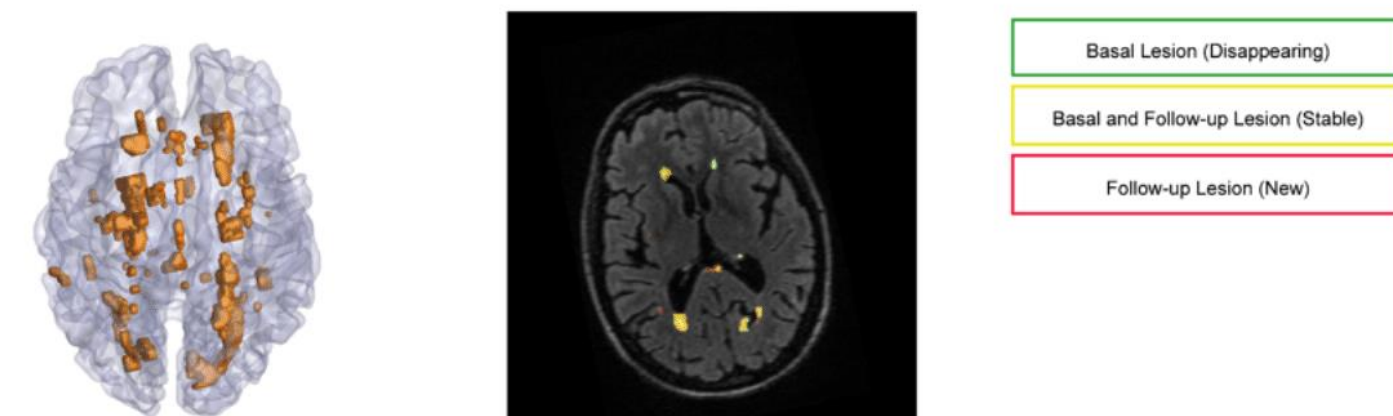
Data from this quantification report should be considered as the results of research with an evidence level 2 (Centre for Evidence-based Medicine) in phase of clinical approval. QUIBIM S.L. - Quantitative Imaging Biomarkers in Medicine. Avenida Fernando Abril Martorell 106, Torre A, Biopolo La Fe, Valencia (SPAIN)



http://www.quibim.com
info@quibim.com

Brain Longitudinal MS Lesions

Imaging Center		Patient Name	
Modality	MR	Patient ID	
Study Description		Patient Sex	
Study Date		Birthdate	



	Lesion Study			
	Basal		Follow-up	
Flair Lesions	Significant Lesion Count	44	Significant Lesion Count	51
	Total Lesion Volume [mL]	15.87	Total Lesion Volume [mL]	17.07
	Dominant Lesion Volume [mL]	6.37	Dominant Lesion Volume [mL]	8.50
	Dominant Relative Lesion Volume [%]	40.14	Dominant Relative Lesion Volume [%]	49.79
	Brain Parenchyma Fraction (BPF) [%]	0.77	Brain Parenchyma Fraction (BPF) [%]	0.77
Texture Biomarkers	Minimum Kurtosis	1.62	Minimum Kurtosis	1.27
	Mean Kurtosis	3.25	Mean Kurtosis	3.31
	Mean Entropy	1.55	Mean Entropy	1.39
	Maximum Entropy	3.13	Maximum Entropy	2.99

Longitudinal Analysis			
New Significant Lesion Count	7	Change of Brain Parenchyma Fraction (BPF) [%]	0.09
Percentage of Estimated Brain Volume Change per Year (PBVC/y) [%]	0.40	Change of Minimum Kurtosis [%]	-21.31
Percentage of Estimated Brain Lesion Volume Change (PBLV) [%]	7.54	Change of Mean Kurtosis [%]	1.82
Annualized Rate of Brain Volume Loss (AR-BVL) [%]	0.40	Change of Mean Entropy [%]	-10.66
New Flair Lesions [mL]	0.94	Change of Maximum Entropy [%]	-4.62
Enlarging Flair Lesions [mL]	3.95	Volume of Lesion Decrease [mL]	2.91

Data from this quantification report should be considered as the results of research with an evidence level 2 (Centre for Evidence-based Medicine) in phase of clinical approval. QUIBIM S.L. - Quantitative Imaging Biomarkers in Medicine. Avenida Fernando Abril Martorell 106, Torre A, Biopolo La Fe, Valencia (SPAIN)

SEGMENTATION / VOLUMETRY



ID referring to MR session id of the processed scans. atrophy is

PATIENT	NAME	ID	DATE OF BIRTH	MRI DATE
	icometrix	ICO-ID_141110_141110 ICO-ID_150330_150330	1966-02-01	2014-11-10 01:01:01 2015-03-30 01:01:01

1. QC

QC Status	Remarks
INTERNAL	This report is for internal reviewing only.

2. VISUAL RESULTS

A qt grey and axia corc

Brain structure	Volume (current MRI)	Normal range (5th and 95th percentile)	Normative percentile	Annual Atrophy
Whole brain volume	1387.1 ml*	1492-1585 ml*	<1	0.56 %
Grey matter volume	830.9 ml*	899-985 ml*	<1	0.58 %

3. BRAINVOLUMES

Vol (bla (whi pop

* Displayed brain volumes are normalised for head size. The normalisation factor for this patient equals 0.71.

4. LESION LOAD

Type	Lesion volume (current MRI)	Lesion volume change (compared to previous MRI)
FLAIR lesions	5.87 ml	1.02 ml
New FLAIR lesions		0.78 ml
Enlarging FLAIR lesions		0.36 ml
Gd enhanced lesions		

CorticoMetrics

TECHNOLOGY FUNDING PORTFOLIO ABOUT TEAM CONTACT BLOG

ABOUT

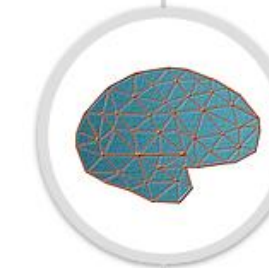
1997-now FreeSurfer Development

FreeSurfer is born out of The Athinoula A. Martinos Center for Biomedical Imaging at Massachusetts General Hospital and is the first surface-based neuroimaging analysis tool, revolutionizing the way that researchers can study the human brain in both healthy and disease states.



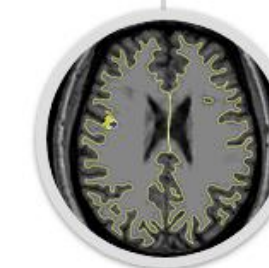
June 2012 CorticoMetrics Formation

Dr. Bruce Fischl and Mr. Nick Schmansky formed CorticoMetrics LLC with their sights set on bringing quantitative neuroimaging to clinical settings.



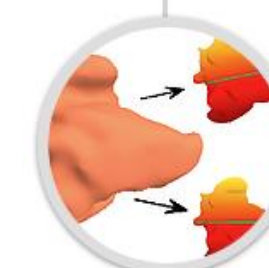
AUG 2013 1st Grant Awarded

Awarded Phase I STTR from NIH-NINDS to create and evaluate a software tool to detect focal cortical dysplasias in MRI images allowing easier visual detection by a neuro-radiologist. Award amount \$359,391



APR 2017 8th Grant Awarded

Awarded Phase II STTR from NIH-NCI to create a software-based system for an MRI scanner to reduce the error in tumor measurement introduced by varying patient head positioning across multiple scan imaging sessions. Award amount \$750,000

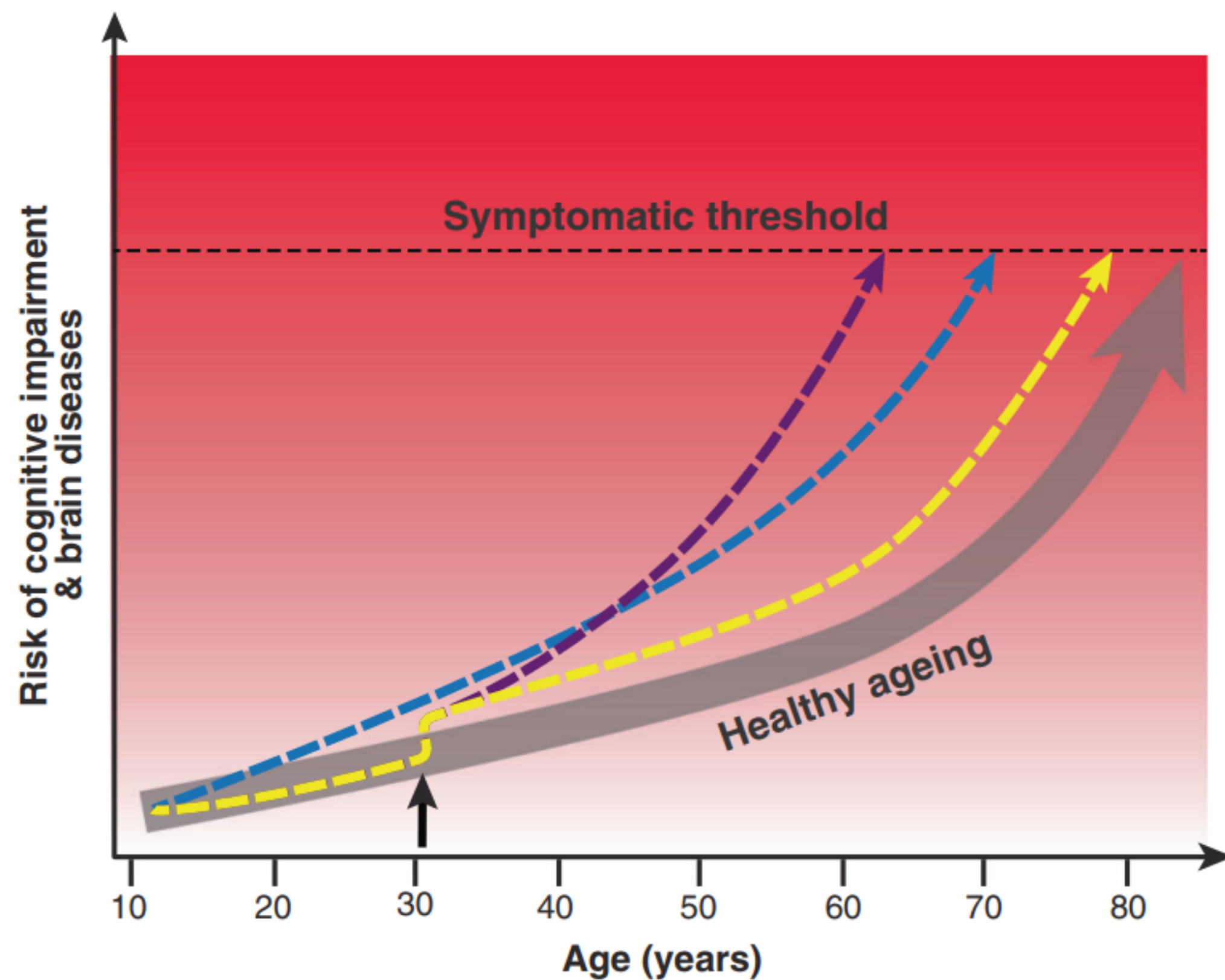


CONFIDENTIAL

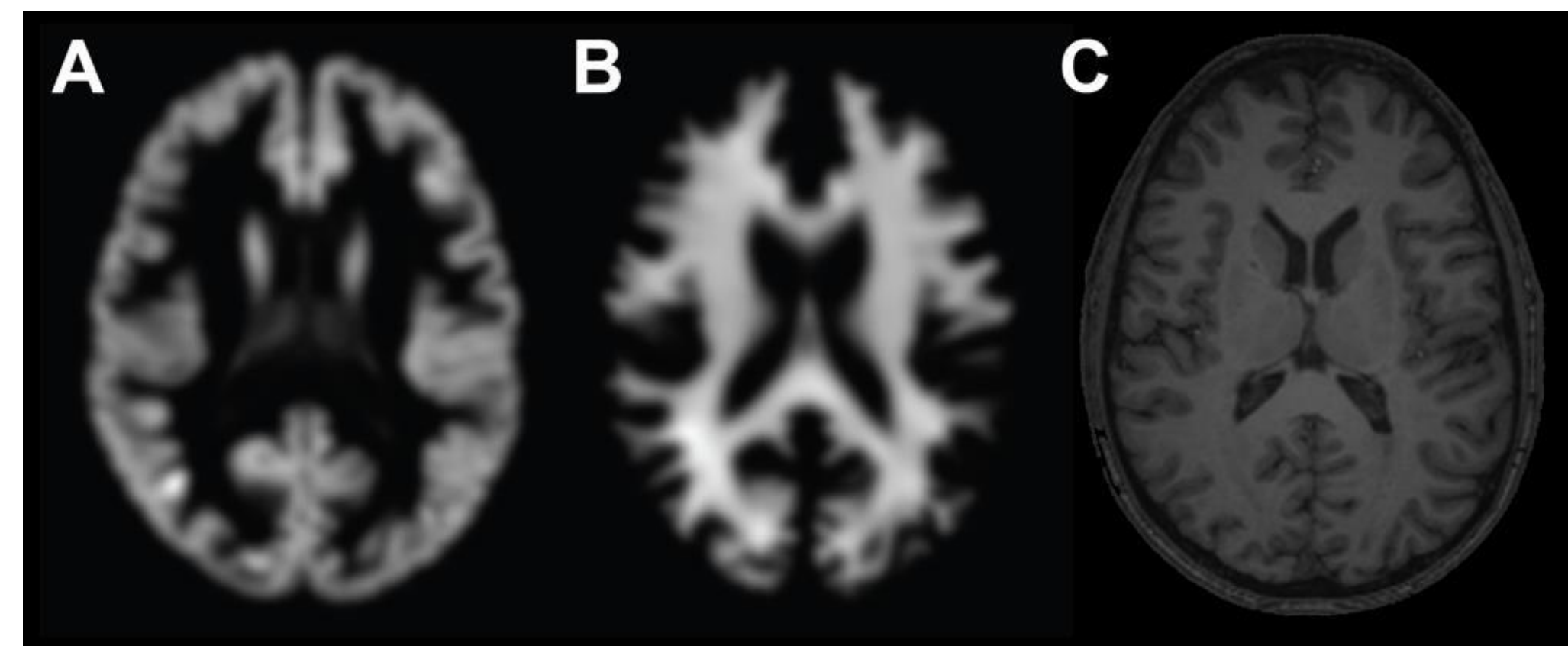
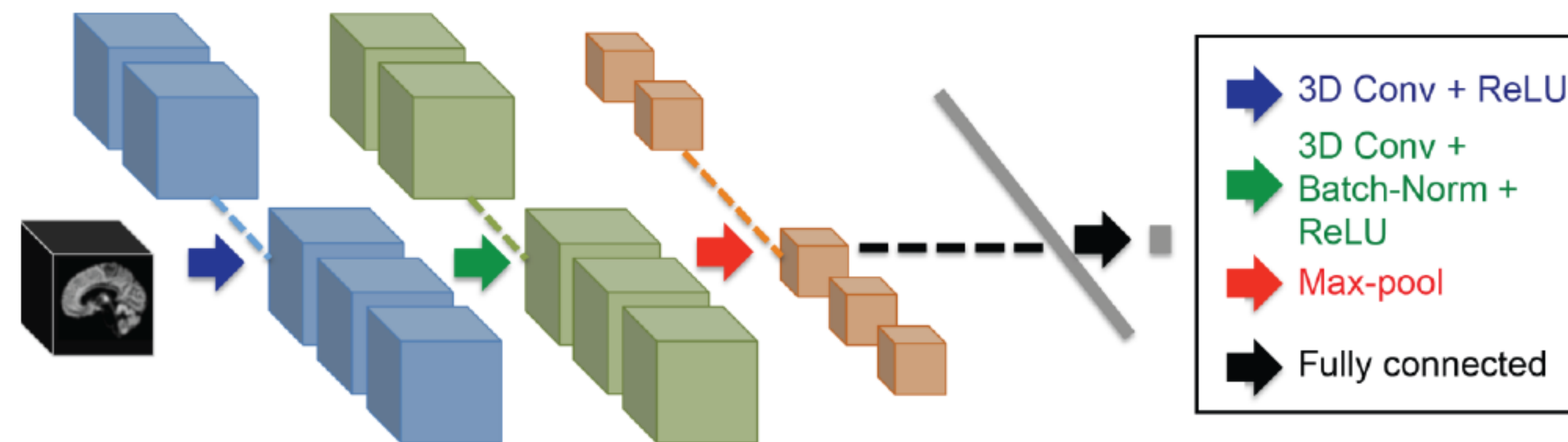
© 2015 icometrix NV. www.icometrix.com MSmetrix 1.3.2 344-g816c071 msmetrix_E01890
This report is approved in the EU, CA and IN. Please consult the HCP manual for additional guidance.

Indication of the CE-labeled software version

PREDICTING BRAIN AGE: EARLY DIAGNOSTICS OF ALZHEIMER?



Cole et al., 2018, *Mol Psych*



Cole et al. Neuroimage. 2017

**Chest/Mammo
XRAY/ CT AI**

CHEST XRAY ANALYSIS



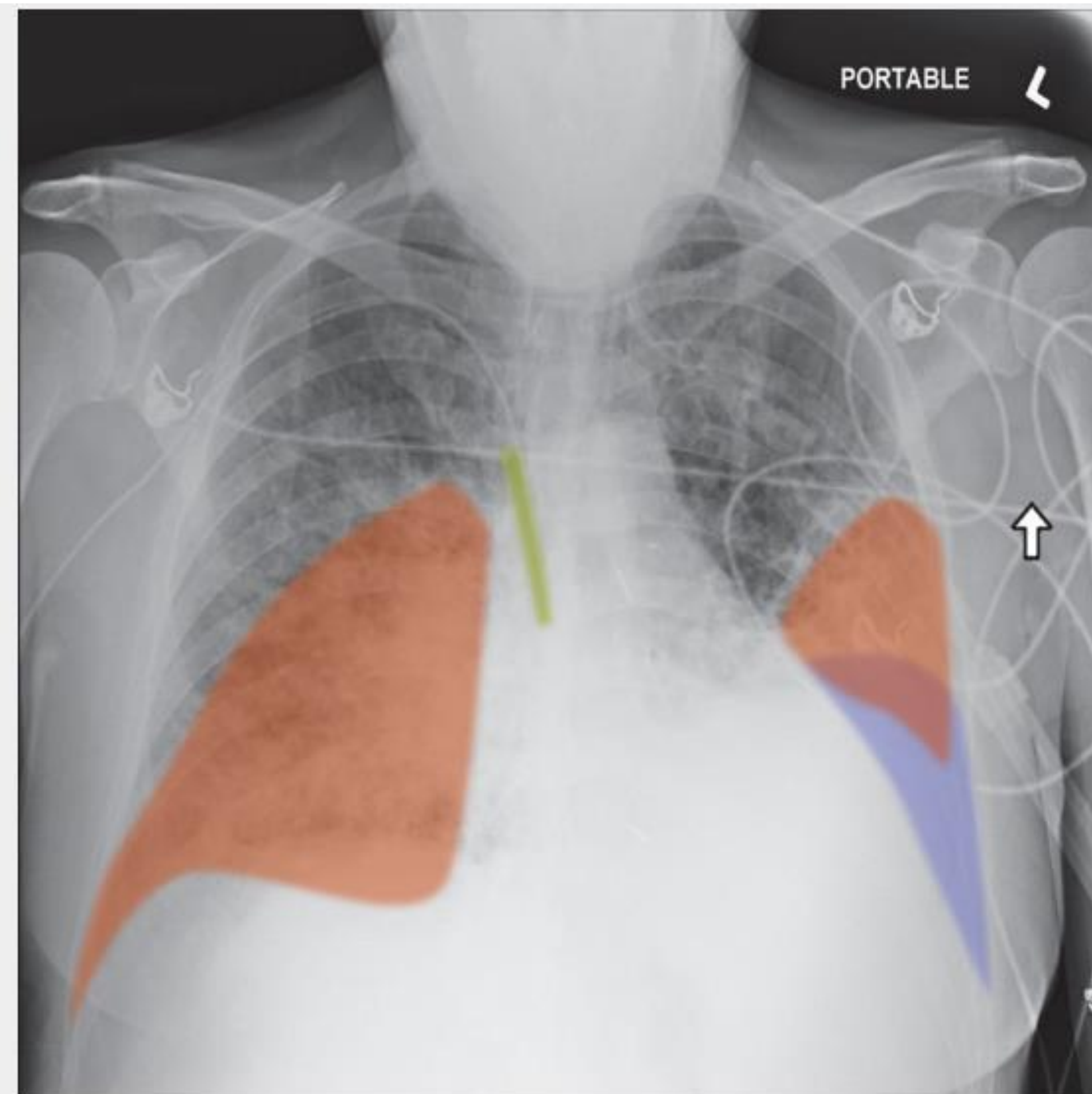
Example results

Findings

- There is volume loss in both lungs. Ill defined opacities are present bilaterally.
- A left sided pleural effusion is seen filling the costophrenic sulcus.
- The hilar area is enlarged.
- The mediastinum is within normal limits.
- Central venous catheter is observed with tip at the superior vena cava.

Impression

Bilateral consolidation. Left pleural effusion.



FRACTURE DETECTION- EXTREMITIES

XRAY

BoneView
Your AI companion for bone trauma X-Rays

GLEAME

Discover BoneView



RAYVOLVE

The first French CE-marked medical device in its category.

Rayvolve is a computer-aided diagnosis tool designed by radiologists for radiologists to optimize their workflow without changing their habits.

Our software is capable of detecting fractures in standard X-rays.

It has been clinically tested and has shown outstanding performance.








Fracture



Dislocation



Fracture

External validation of a commercially available deep learning algorithm for fracture detection in children

Diagnostic and Interventional Imaging
Volume 103, Issue 3, March 2022, Pages 151-159

LUNG NODULE CLASSIFICATION

Nodule: 1
 Slice: 141
 Composition: Solid

Growth: 138%
 VDT: 292 days
 VDT CI: (264, 325)

Current study: 02-01-2001

	Diameter (mm)	Volume (mm ³)	Volume CI
Current	9x6 (8)	233	(223, 244)
Prior	6x4 (5)	98	(90, 106)

Prior study: 02-01-2000 - Slice 146

Veye Chest

aidence

Nodule Analysis

Patient ID
 Accession Number
 Study Date

Prior Accession Number
 Prior Study Date
 Time between 02-01-2000
 366 days

	Diameter (mm)	Volume (mm ³)	Volume CI
Current	7x5 (6)	98	(91, 106)
Prior	6x3 (5)	70	(64, 77)

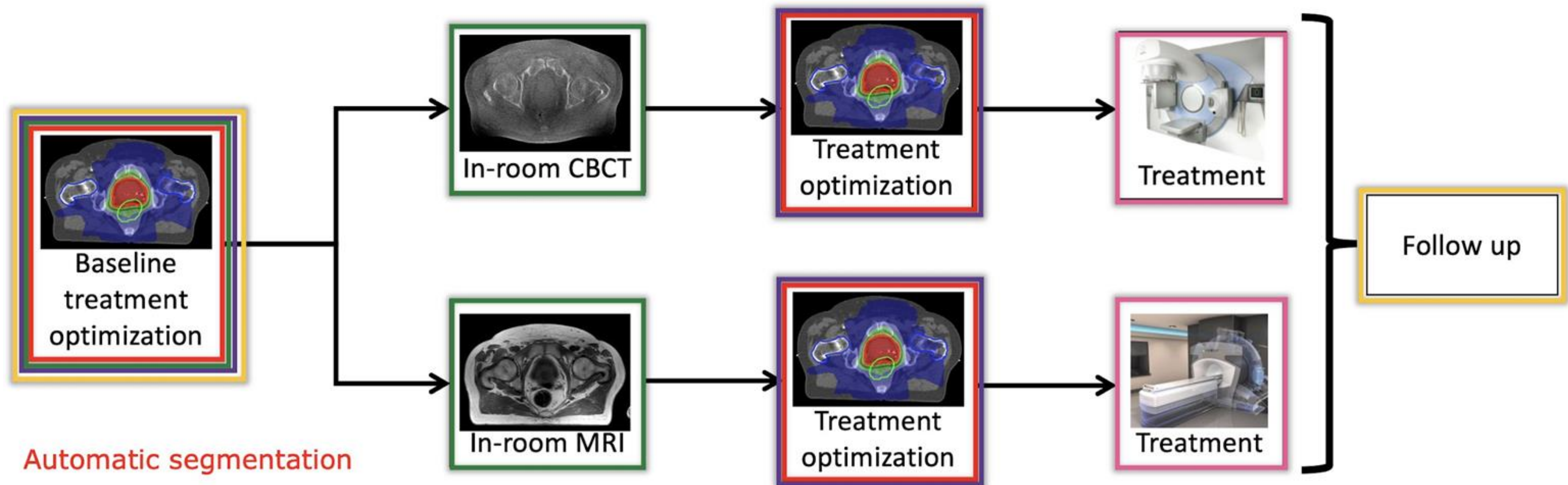
Growth: 40%
 VDT: 761 days
 VDT CI: (563, 1164)

Current study: 02-01-2001

Prior study: 02-01-2000 - Slice 105

- page 2 of 5 -

RADIOTHERAPY: ORGAN/LESION SEGMENTATION/DELINEATION



Automatic segmentation

Pseudo CT generation

Dose prediction and automatic planning

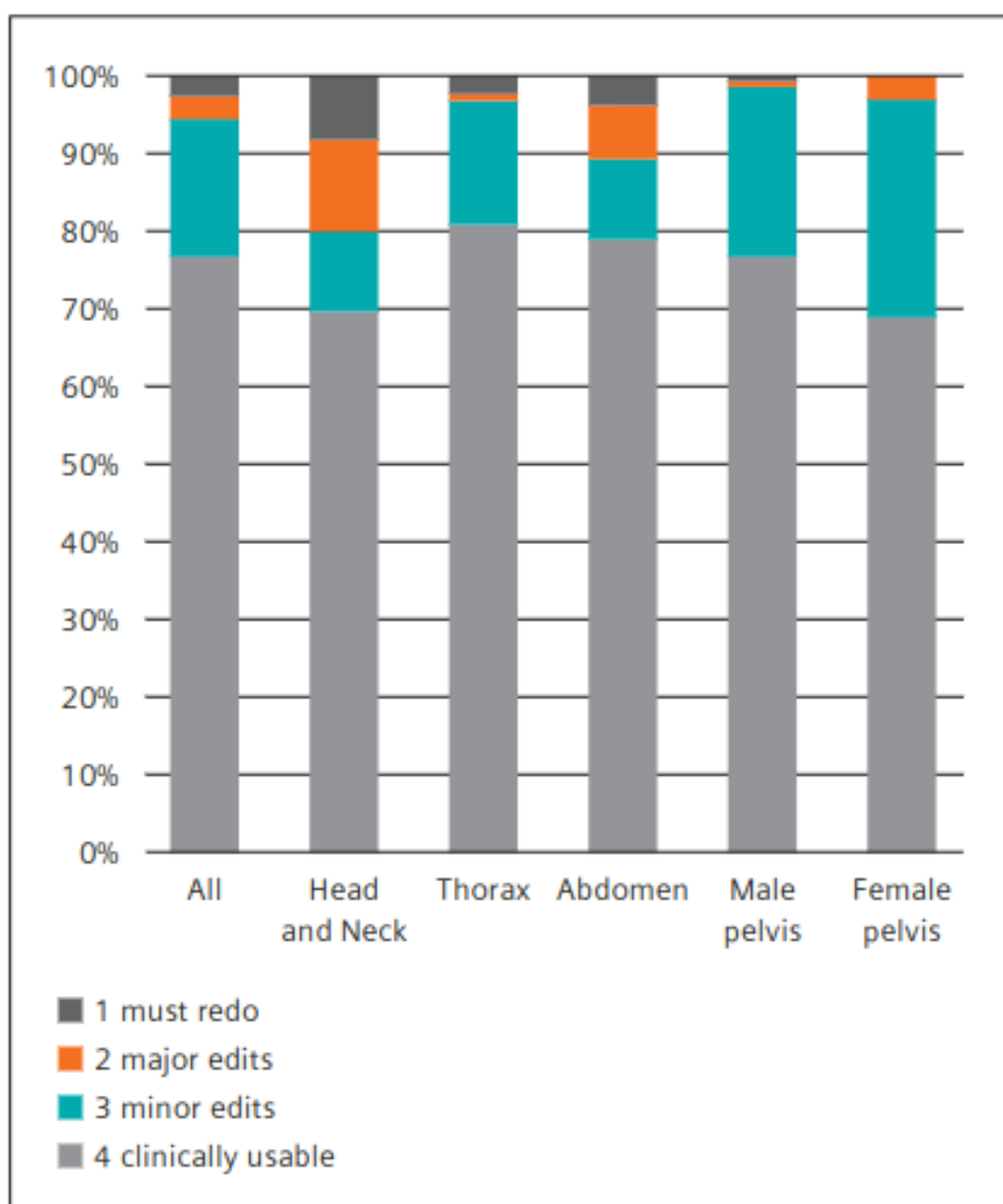
Motion tracking

Outcome prediction

RADIOTHERAPY: ORGAN/LESION SEGMENTATION/DELINEATION

Organs at risk Segmentation for radiotherapy

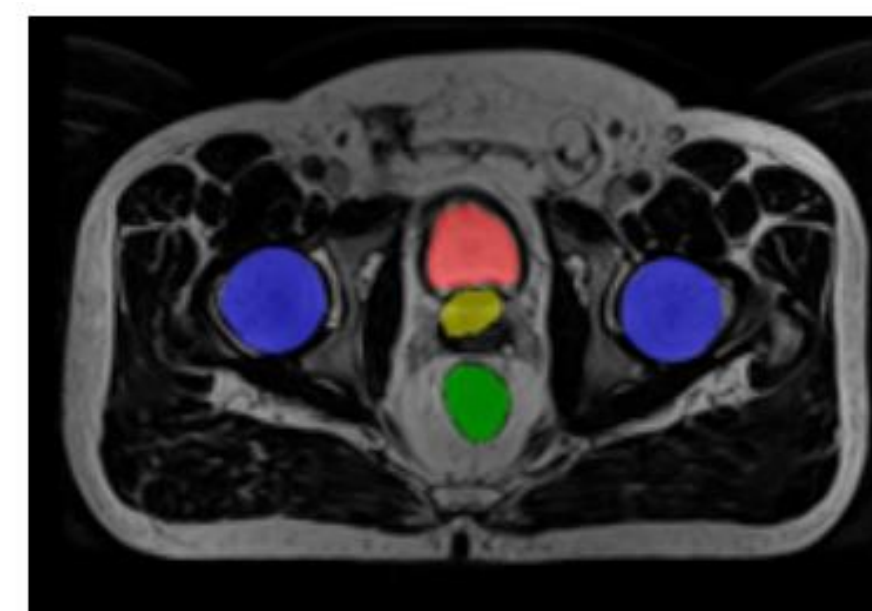
- Main benefit: faster contouring time
- Target Tumor volumes are not automatically segmented yet



Siemens AI-Rad
> 95%
 Clinically usable or minor edits

MRI only radiotherapy planning - pseudo CT

- **Spectronic Medical** - CT simulation based on MR
- Metrics: dose difference, position error based on "bone-alignment"



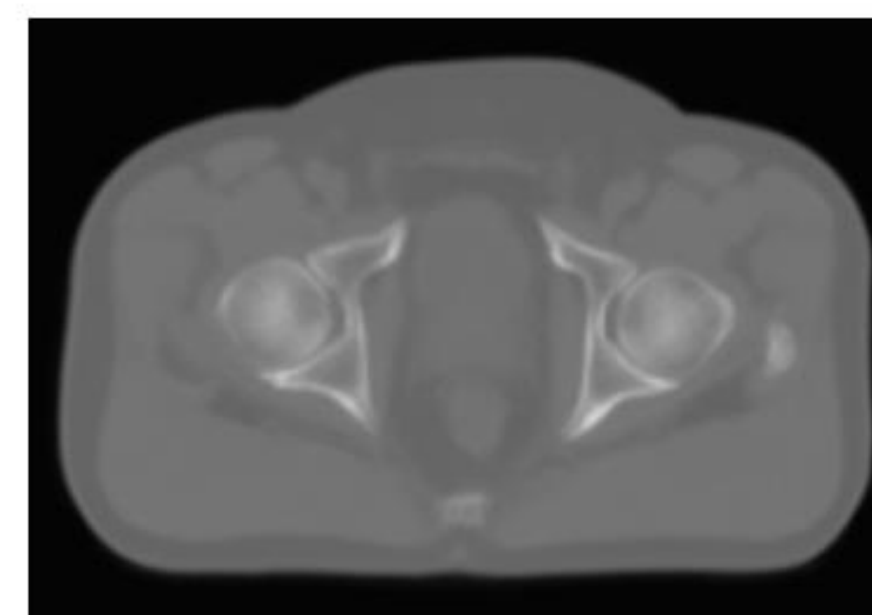
Prostate MRI with delineations generated by MRI Planner

	PTV	Bladder	Rectum	Hips
Dose diff (%)	0.29 (0.38)	0.06 (0.24)	0.10 (0.50)	-0.01 (0.11)
Dose diff (Gy)	0.23 (0.29)	0.05 (0.19)	0.06 (0.39)	-0.01 (0.08)

Difference in calculated mean doses (± 1 S.D.) between conventional and synthetic CT for 62 patients. Relative dose difference is described as fraction of nominal target dose (range 64-78 Gy).

	x	y	z
Pos diff (mm)	0.01 (0.37)	0.04 (0.41)	0.49 (0.60)

Difference in registered patient position (± 1 S.D.) between conventional and synthetic CT when bone-matched to 24 separate CBCT images for 8 patients.



Prostate synthetic CT generated by MRI Planner



BONE AGE ASSESSMENT

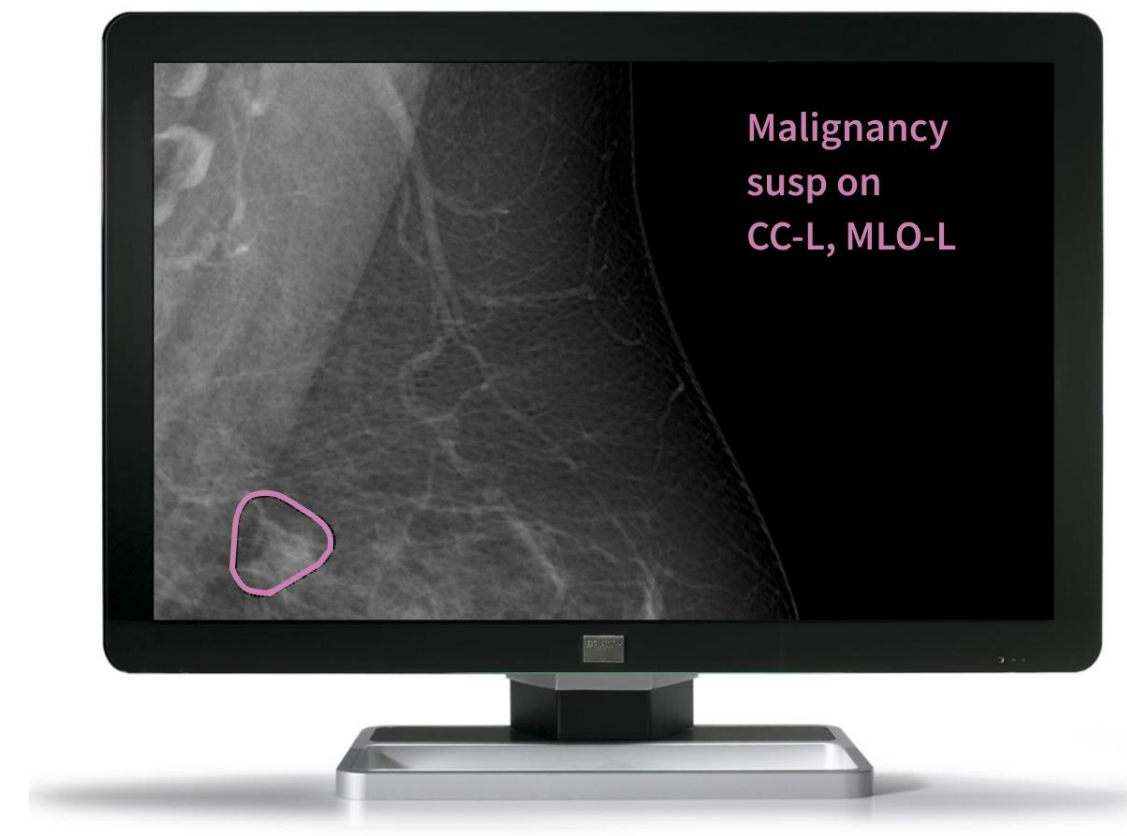
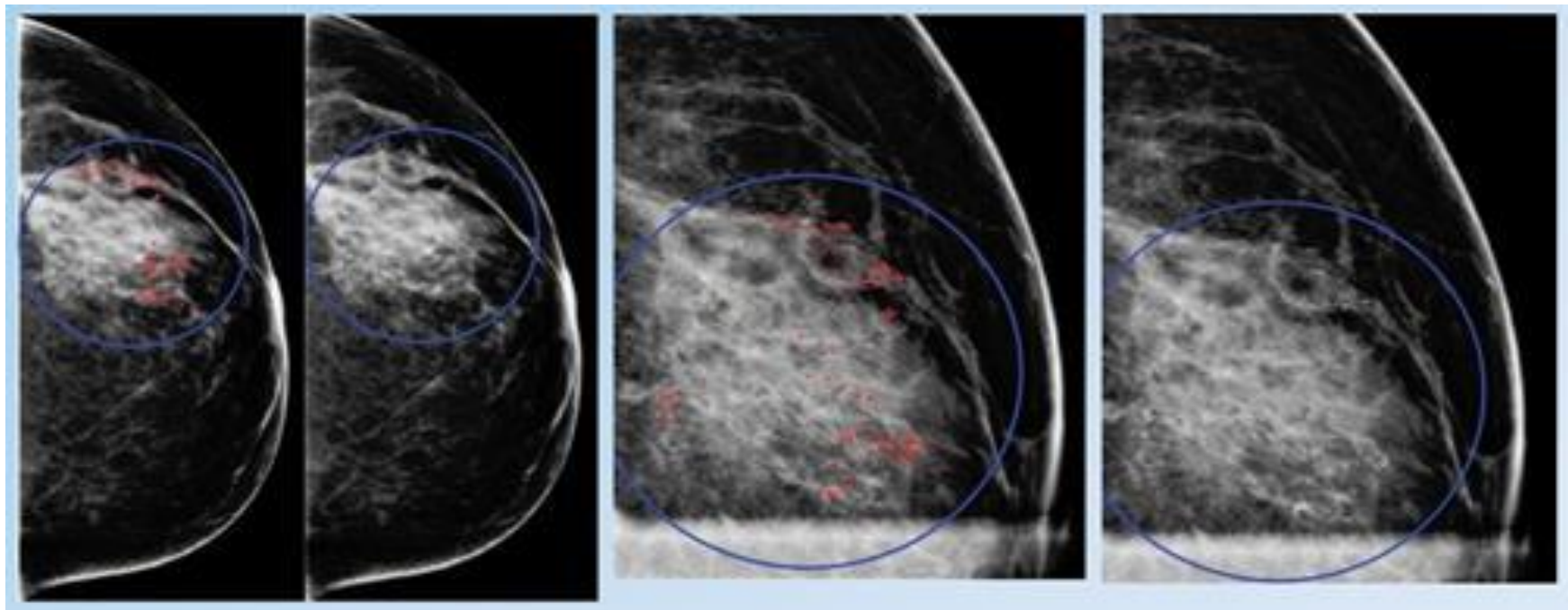
BoneXpert

Bone age estimation [1]	Height estimation [3]
Chronological Age (CA)/Patient Sex	Growth potential
8 Years 0 Months (96 Months) / M	78.6 % achieved
Bone Age	Height calculation
11 Years 6 Months (138 Months)	128 cm * 100 / 78.6 = 162.8 cm ± 2.5 cm
Natural standard Deviation (SD) [2]	To calculate the growth potential including standard deviation, use the formula above.
9.1 Months	
Status (based on 2SD) [2]	
ADVANCED	

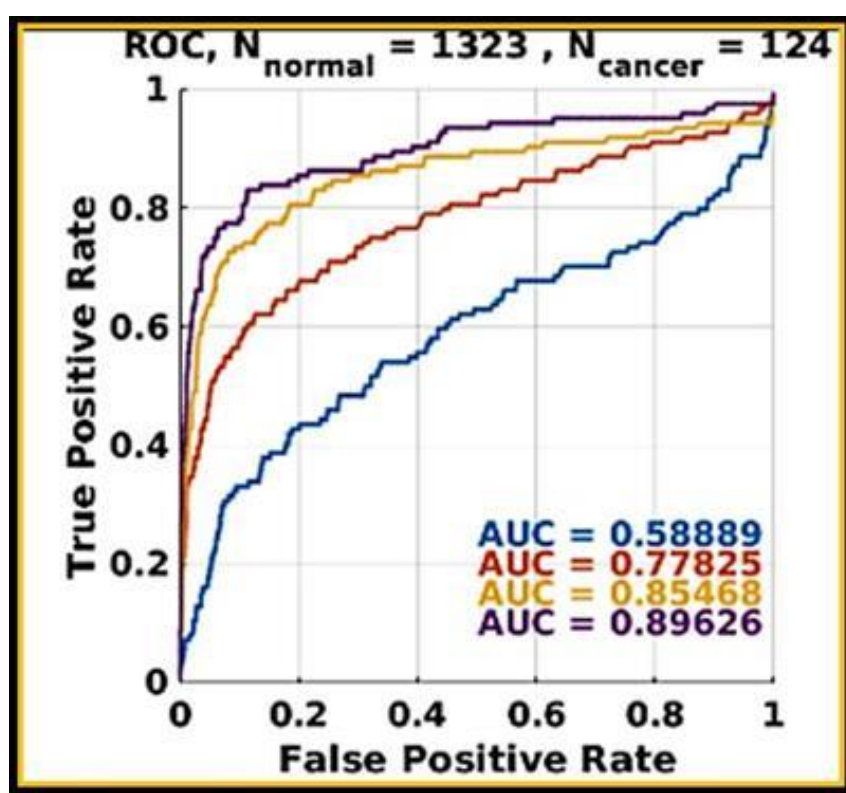
Greulich & Pyle ±4.3 months mean absolute deviation
adult height estimation according to Bailey and Pineau ±2.5 cm



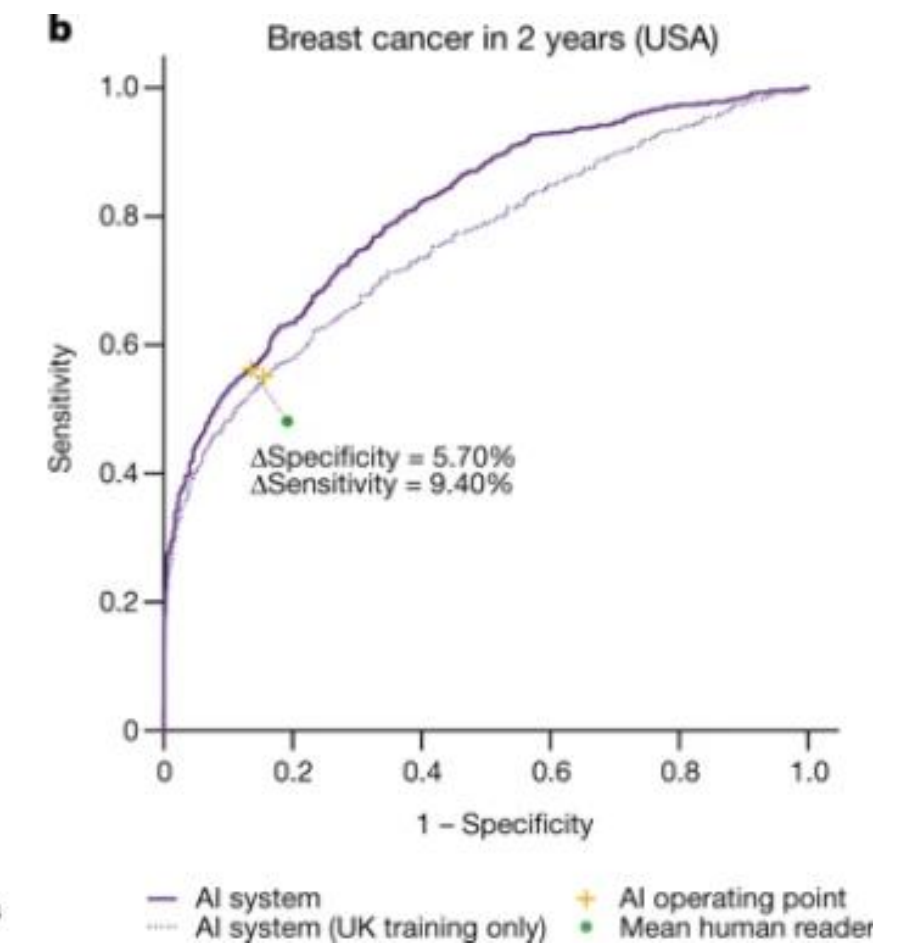
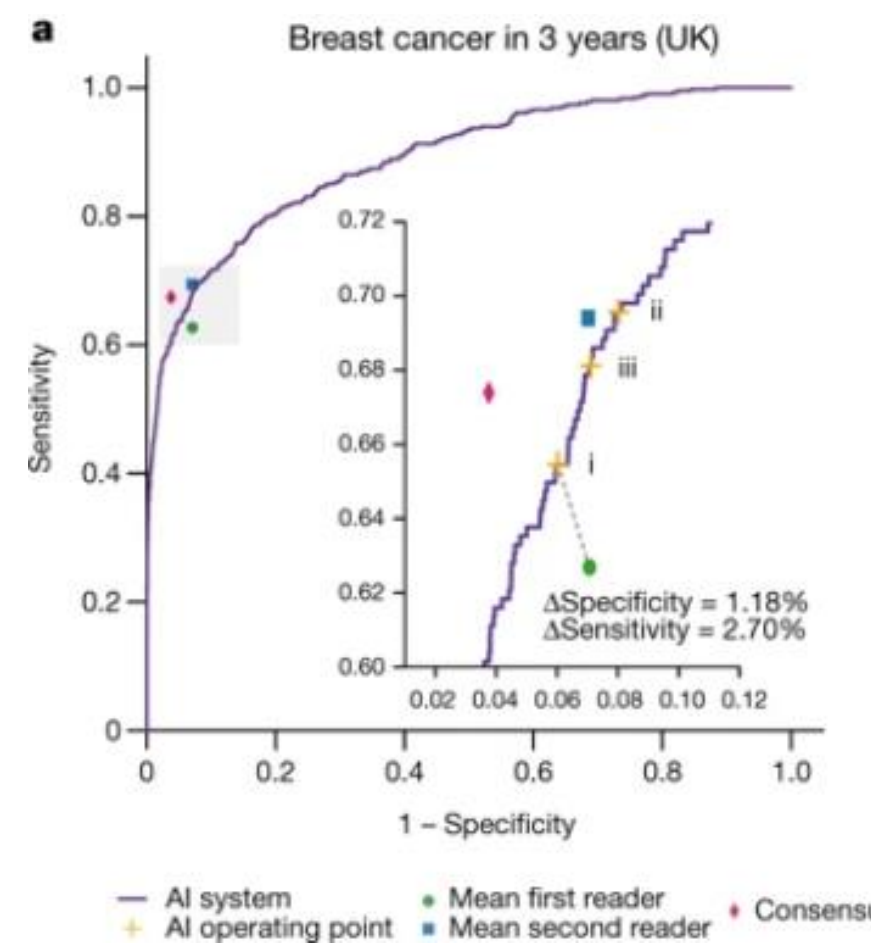
CAD OUTPERFORMING MAMMOGRAPHERS?



Mia™
CE



Performance of CAD software vs. radiologists by false-positive rates			
	Radiologists	CAD software (at 100% sensitivity threshold for cancer detection)	Potential reduction in breast biopsies from use of CAD software
Academic radiology department	80%	35%	57%



Alyssa Watanabe of the University of Southern California (USC) Keck School of Medicine, ECR 2017

McKinney, S. M. *et al. Nature* 577, 89–94 (2020).

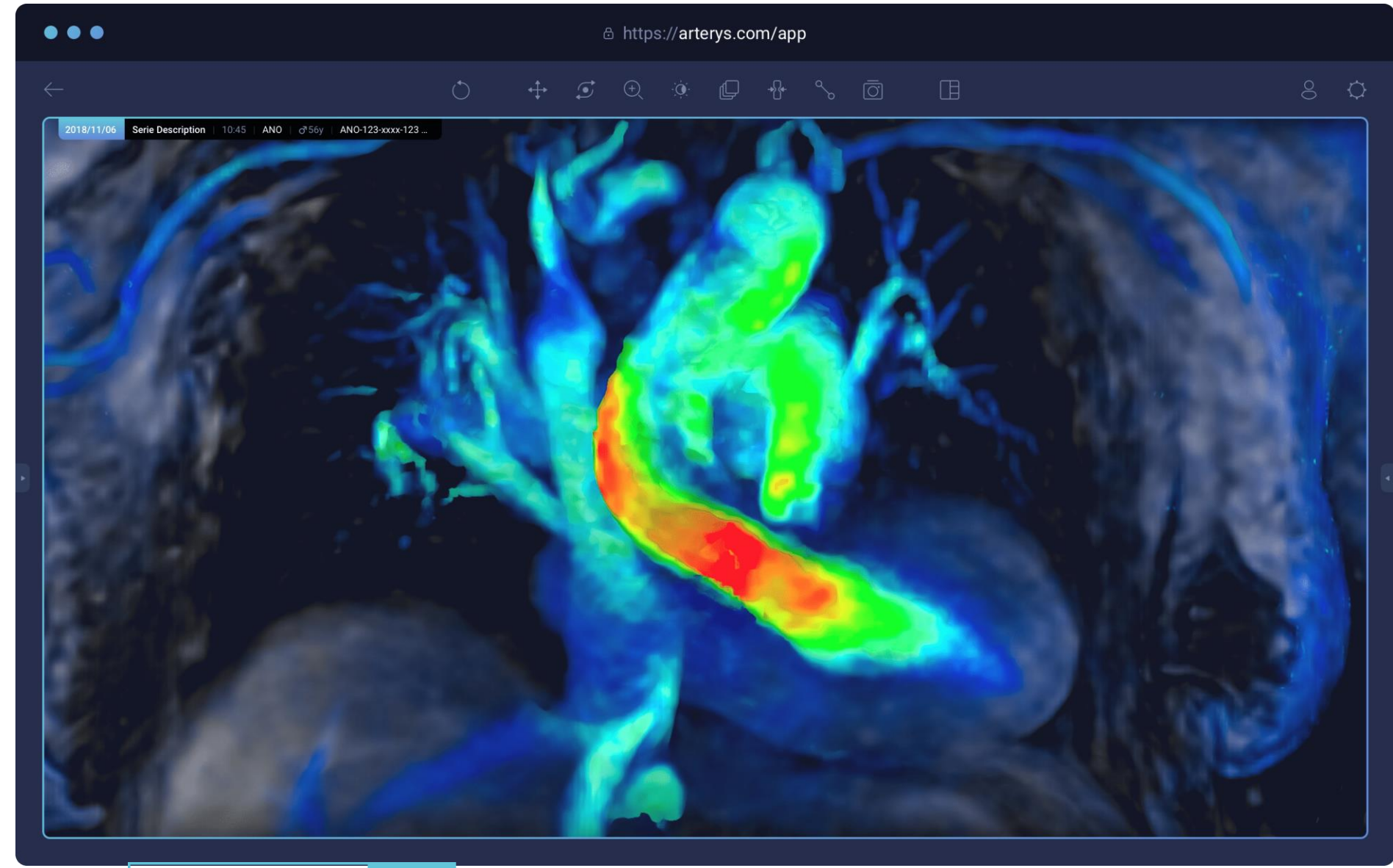
AI in CARDIOLOGY

CARDIAC IMAGING: PLANNING, 4D FLOW, CA+++ SCORE

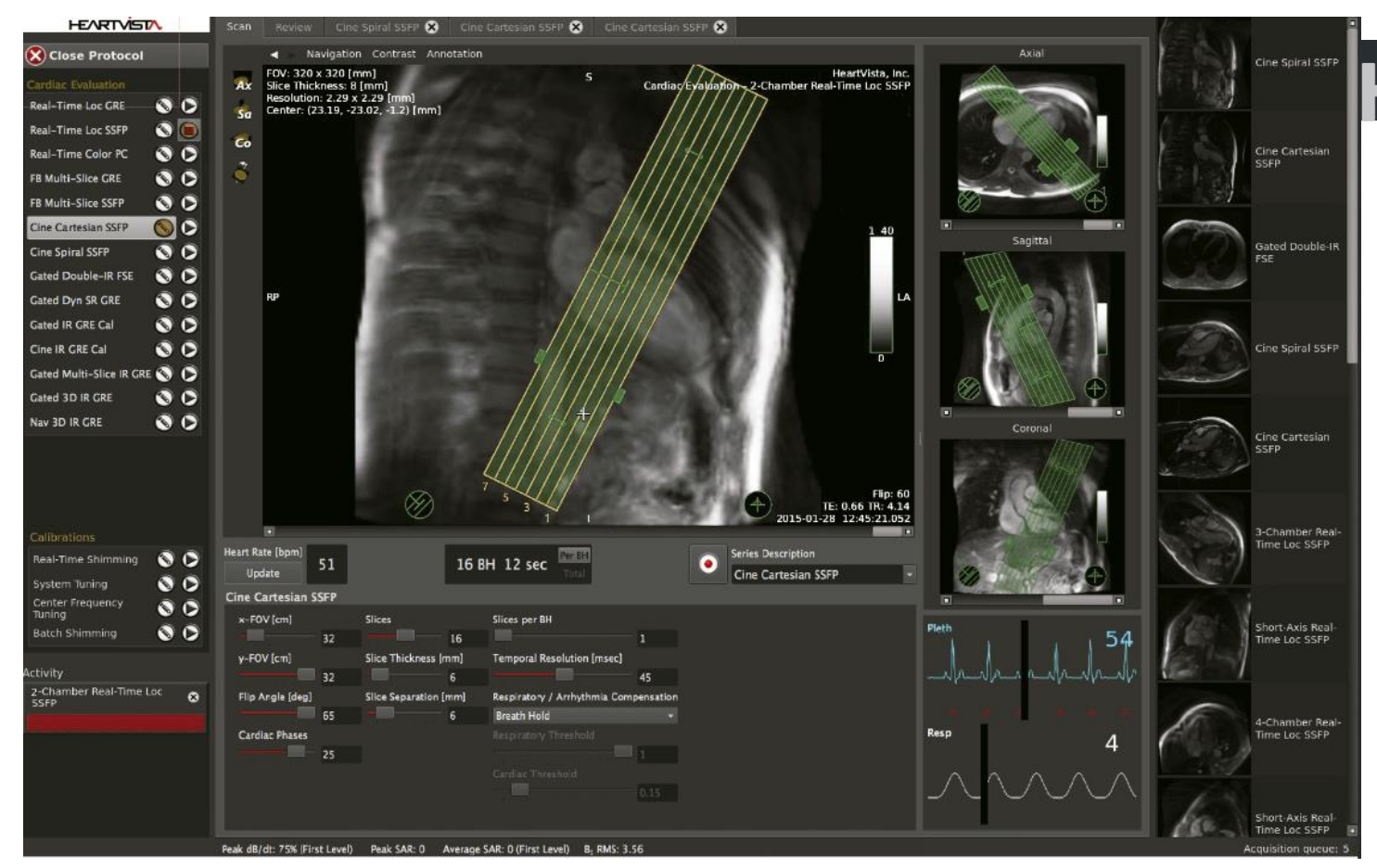
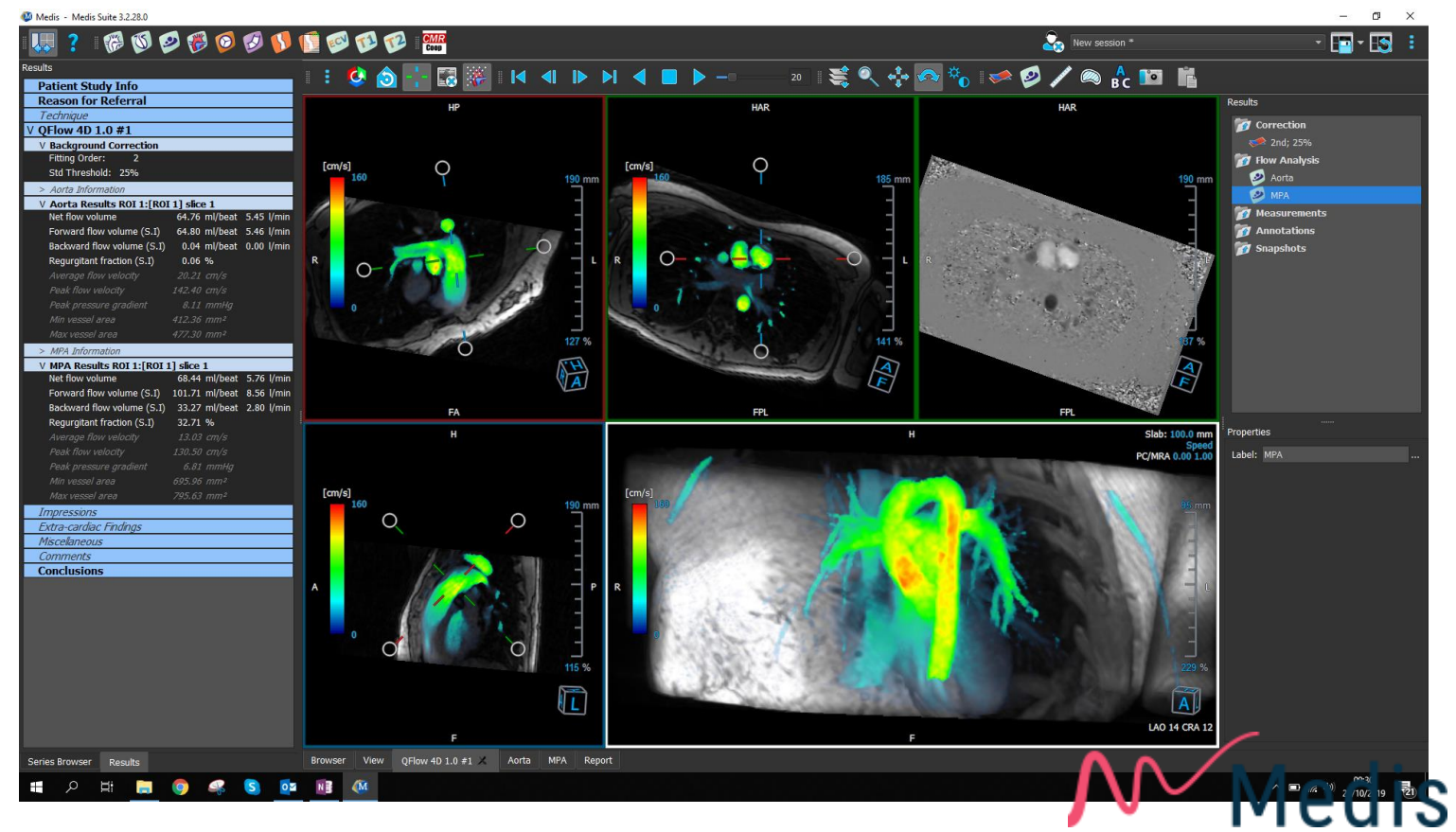


Calcium score, FDA cleared

4D Flow (FDA approved)



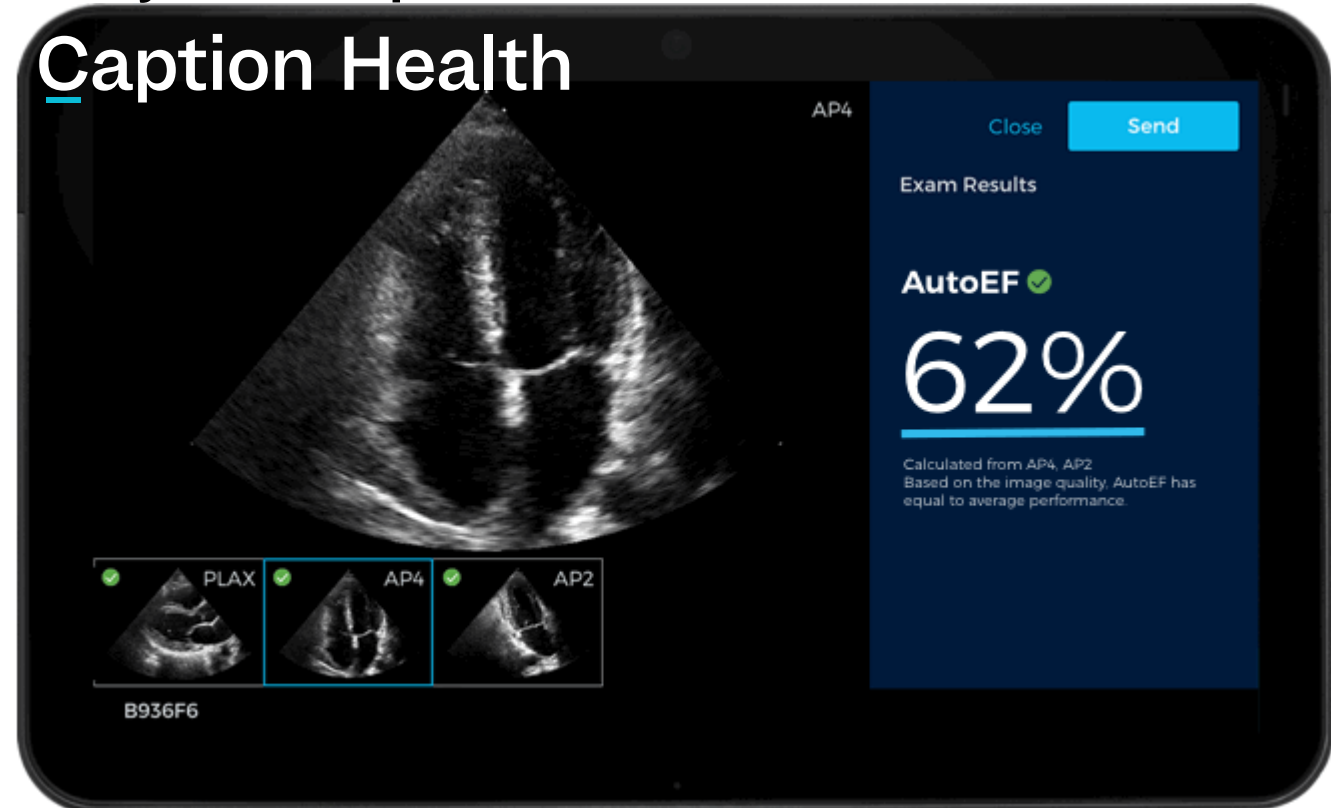
4D Flow (FDA approved)



OneClick (FDA approved)

CARDIAC SEGMENTATION/ CONTOUR DETECTION: EF /STRAIN/VOLUMETRY ASSESSMENT

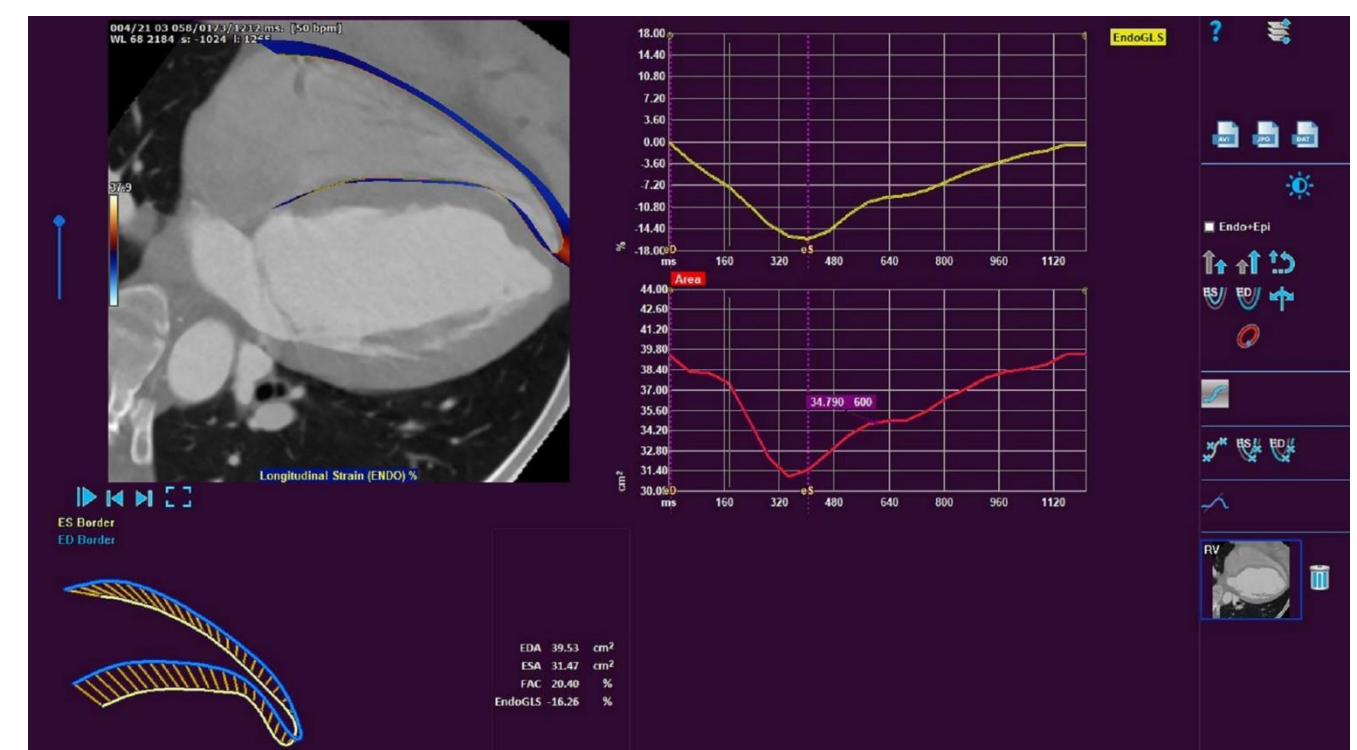
Bay Labs/Caption Health



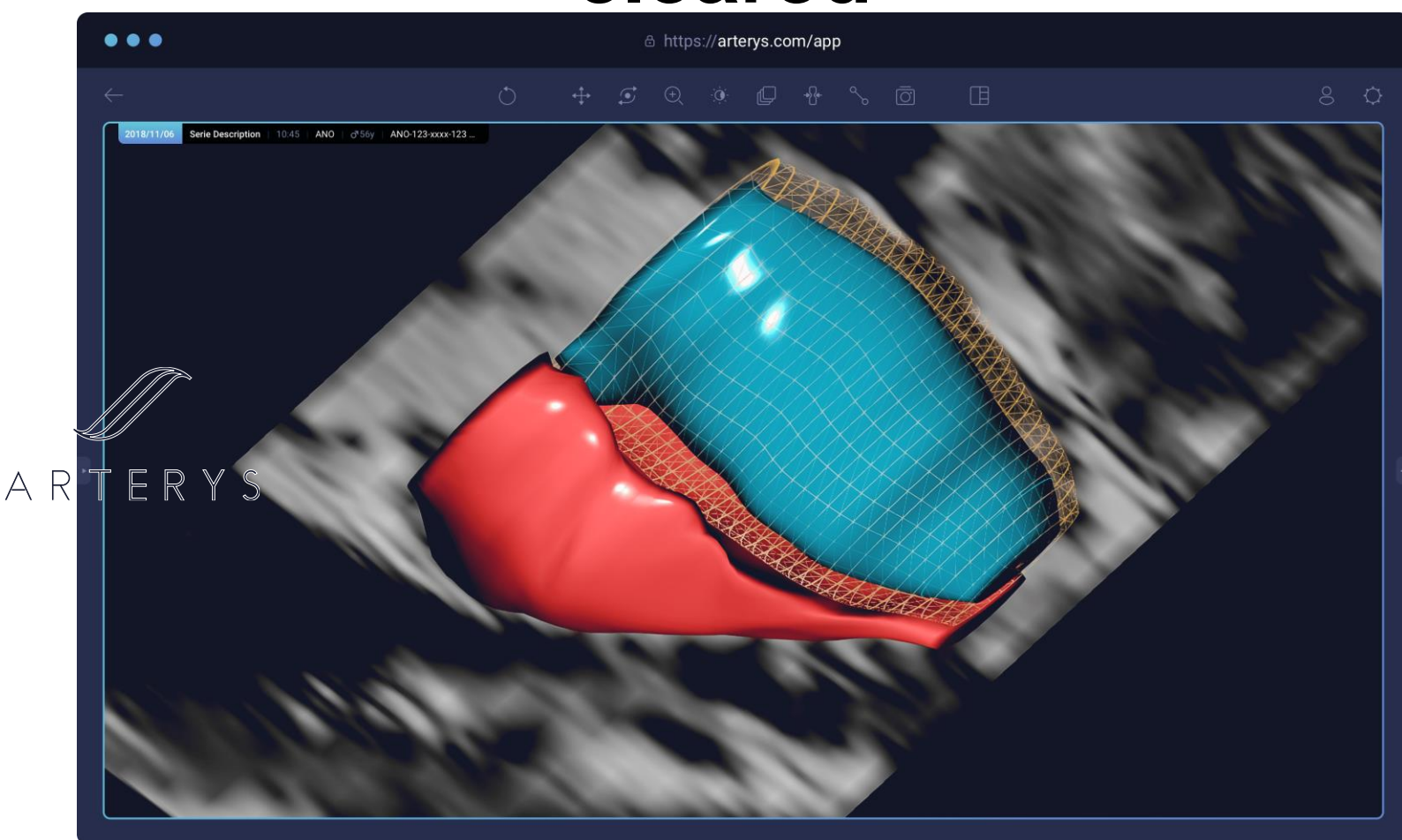
Echo: AI contour, EF, FDA cleared



Echo: AI contour, EF, strain FDA cleared

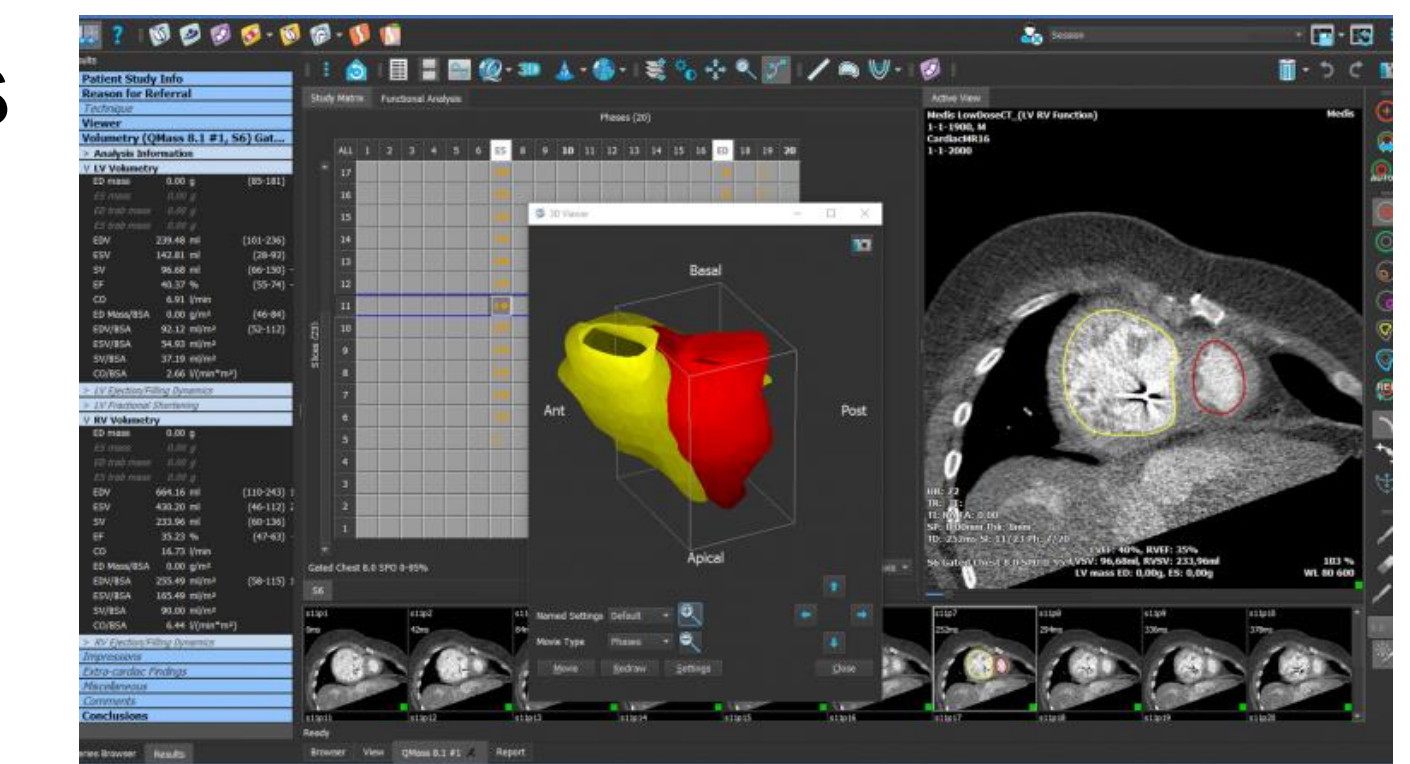
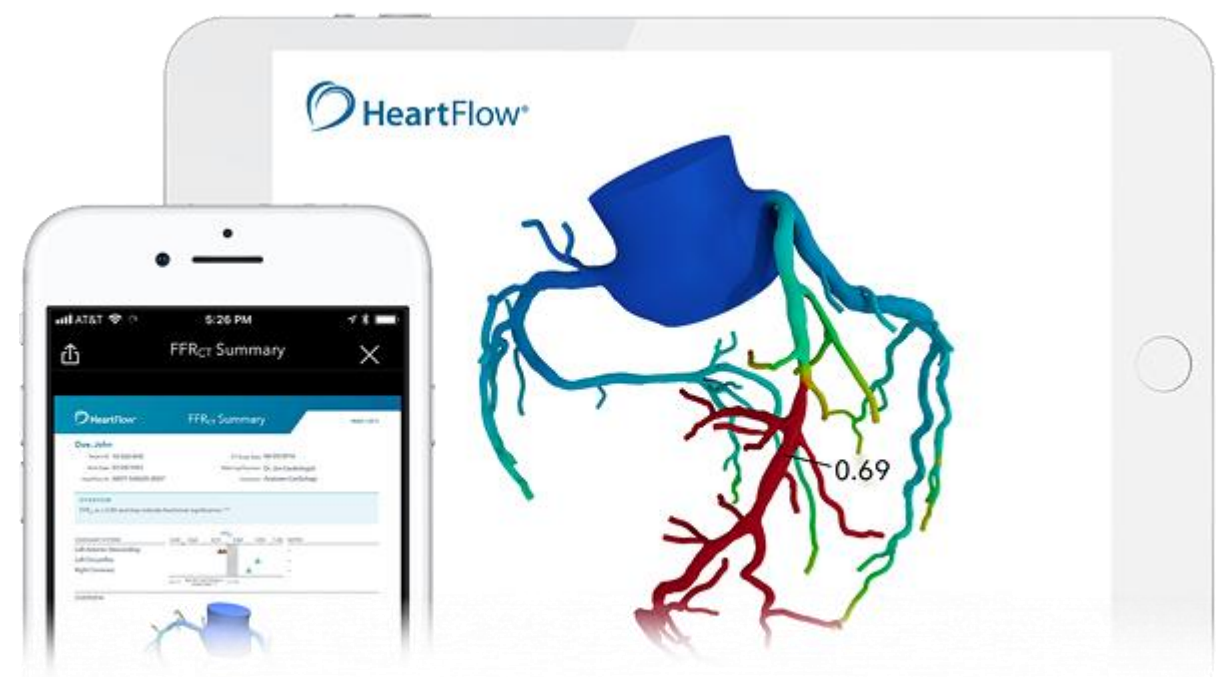


CT strain (Research)



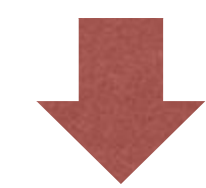
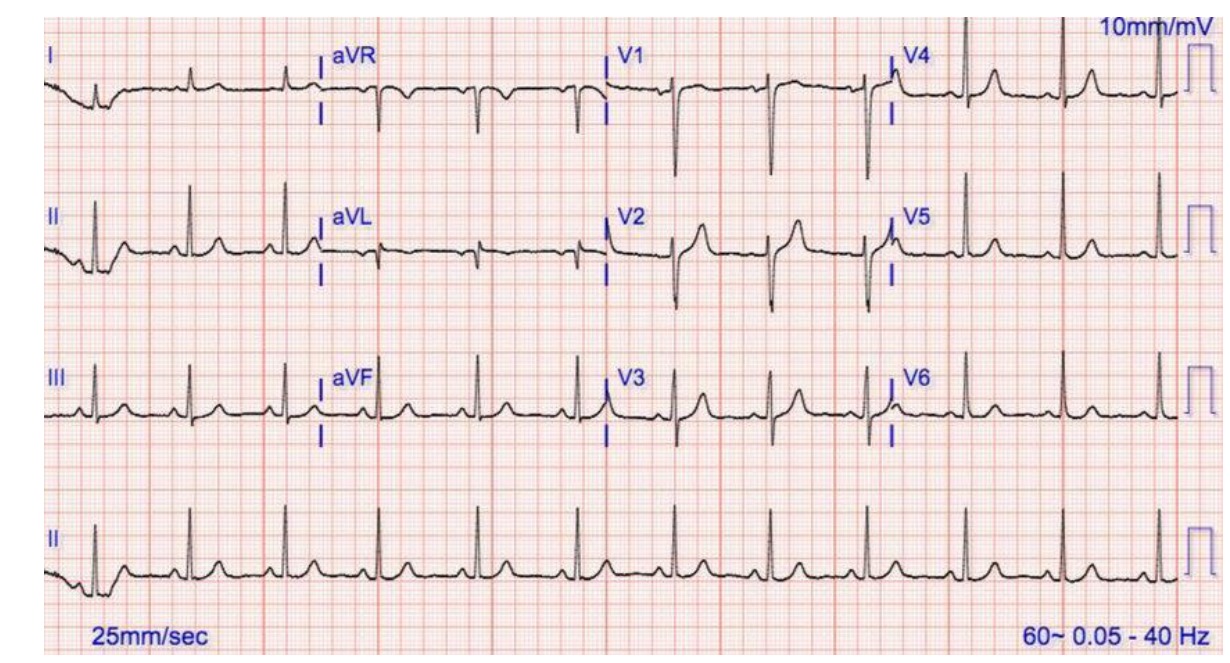
2D SSFP and 3D Cine LV/RV segmentation(FDA)

HeartFlow FFR_{CT} Analysis

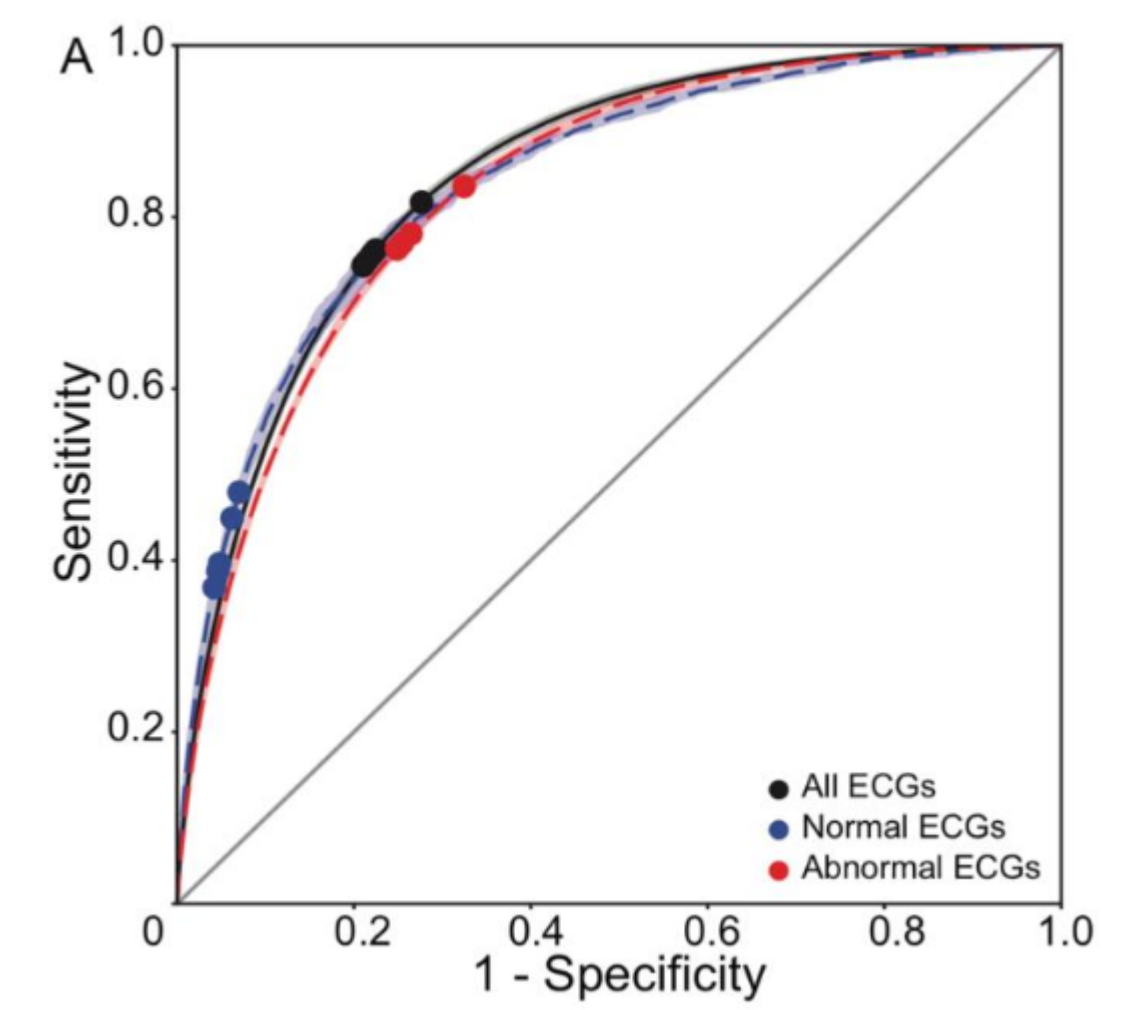
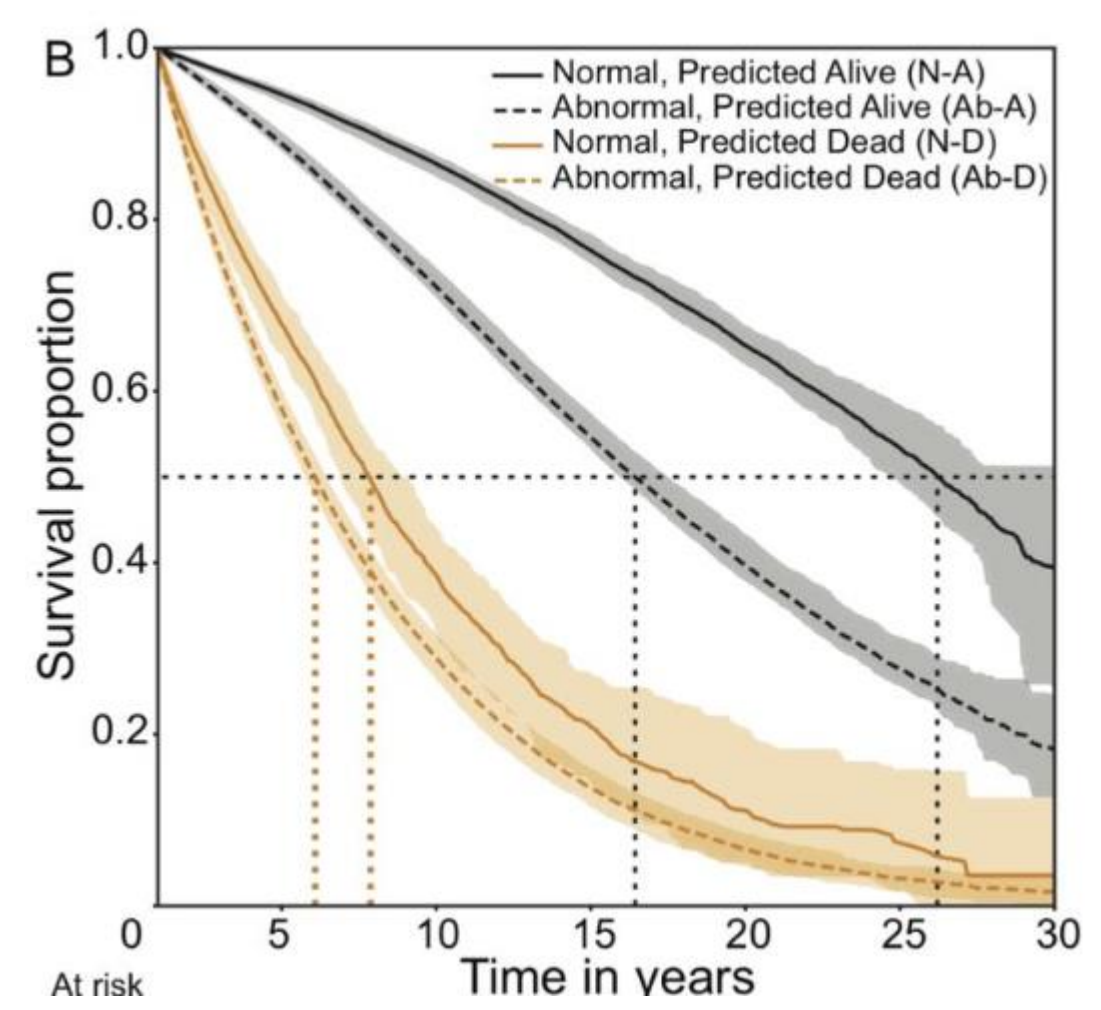
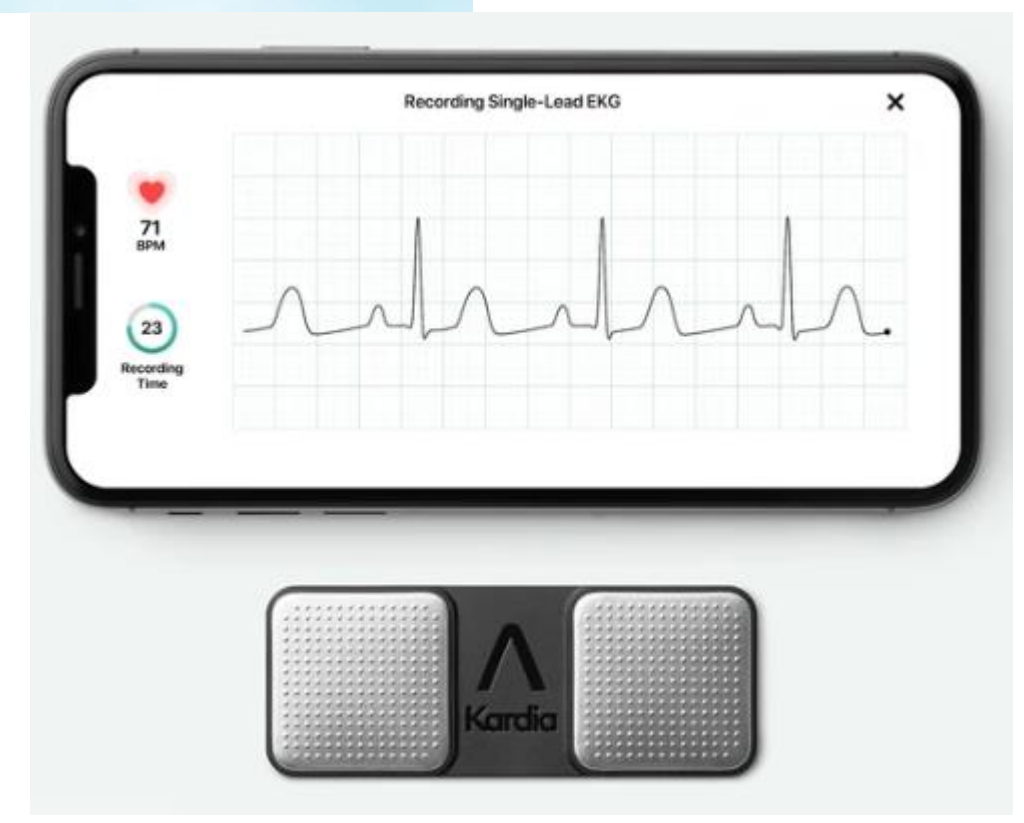


Medis CT contour (FDA cleared)

ECG: AFIB DETECTION, ARRHYTHMIA/ DEATH PREDICITON



Predicting 1-year all-cause mortality: 0.830 (AUC)

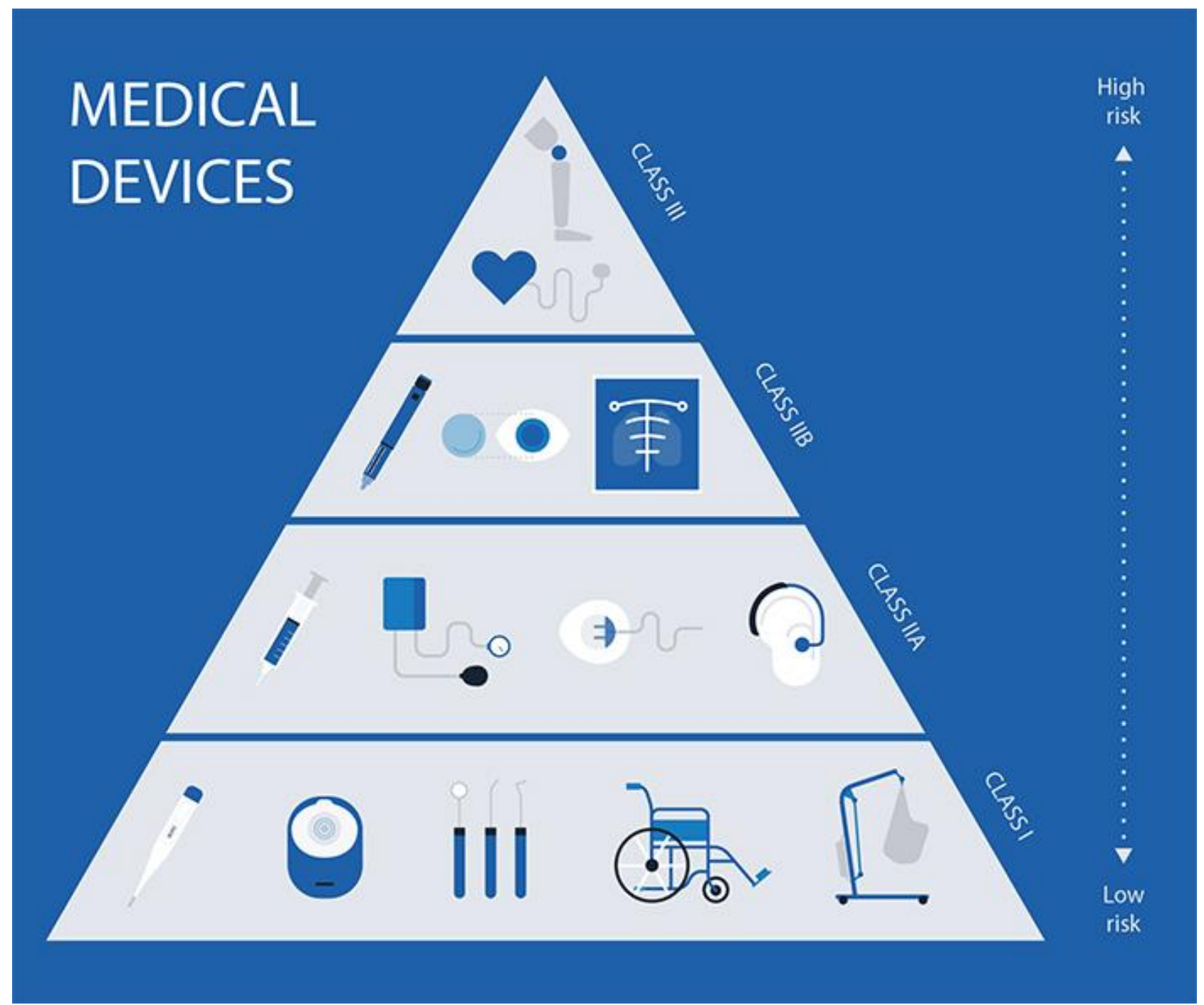


Raghunath et al, Nature Medicine 2020

**Ethical issues,
regulations**

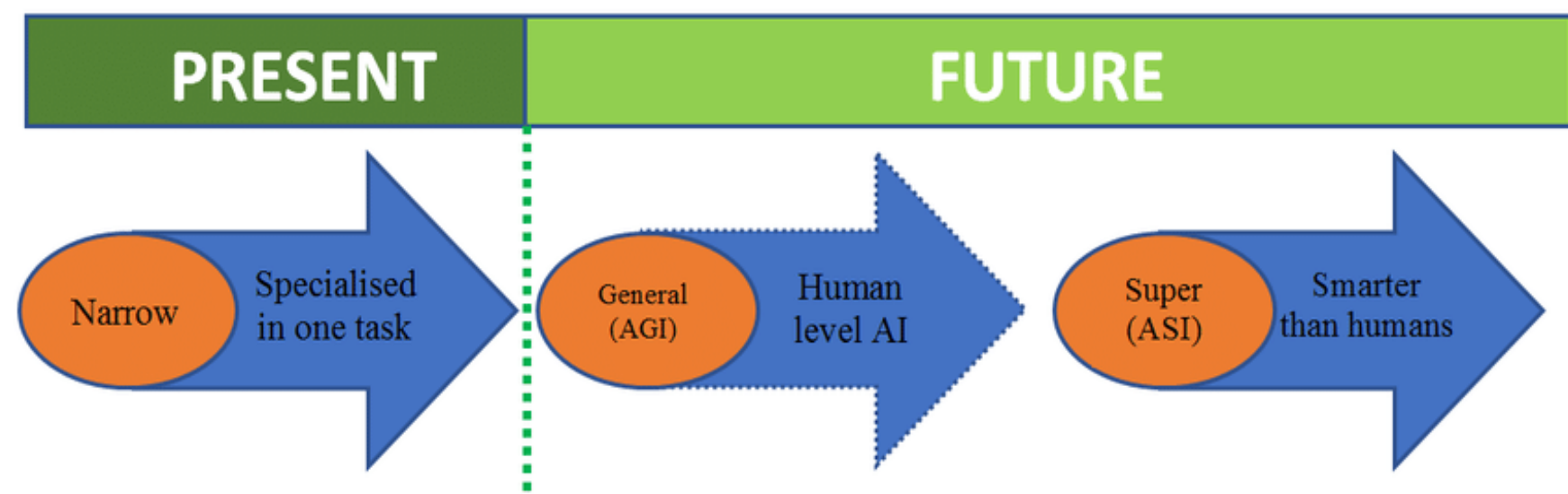
CE CERTIFICATION: CLASSES

		Significance of Information provided by the MDSW to a healthcare situation related to diagnosis/therapy		
		High Treat or diagnose ~ <i>IMDRF 5.1.1</i>	Medium Drives clinical management ~ <i>IMDRF 5.1.2</i>	Low Informs clinical management (<i>everything else</i>)
State of Healthcare situation or patient condition	Critical situation or patient condition ~ <i>IMDRF 5.2.1</i>	Class III <i>Category IV.i</i>	Class IIb <i>Category III.i</i>	Class IIa <i>Category II.i</i>
	Serious situation or patient condition ~ <i>IMDRF 5.2.2</i>	Class IIb <i>Category III.ii</i>	Class IIa <i>Category II.ii</i>	Class IIa <i>Category I.ii</i>
	Non-serious situation or patient condition (<i>everything else</i>)	Class IIa <i>Category II.iii</i>	Class IIa <i>Category I.iii</i>	Class IIa <i>Category I.i</i>

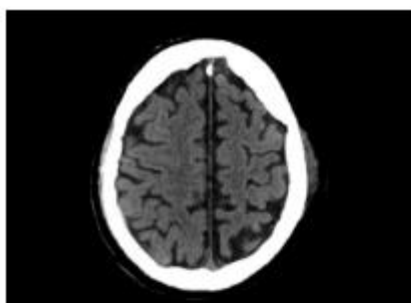

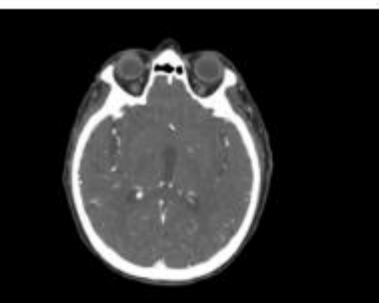
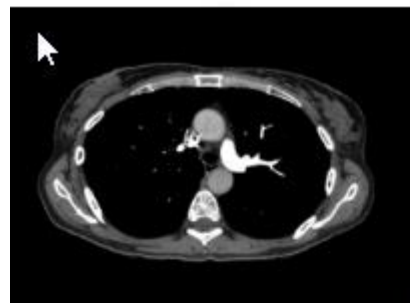
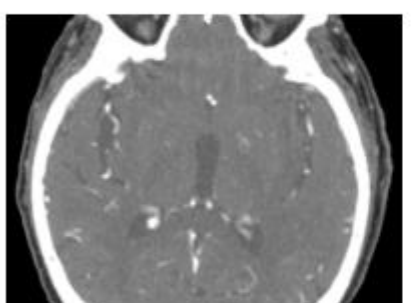
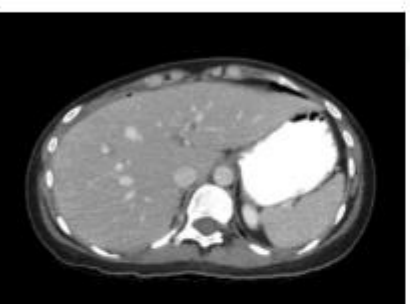



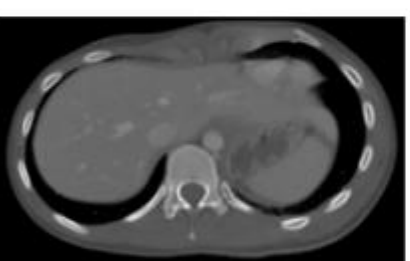



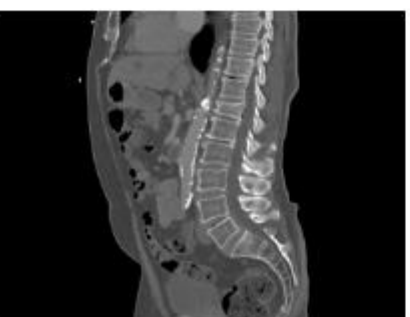
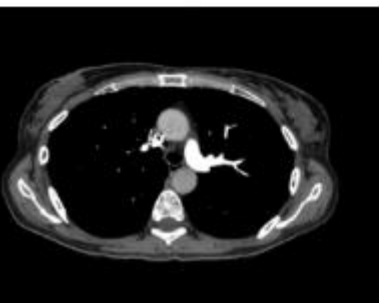
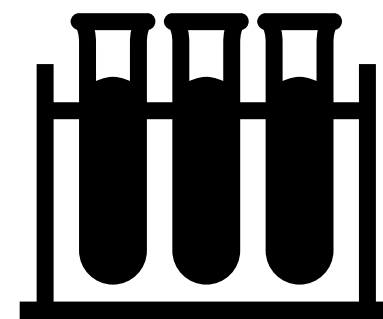



<https://towardsdatascience.com/how-to-get-clinical-ai-tech-approved-by-regulators-fa16dfa1983b>

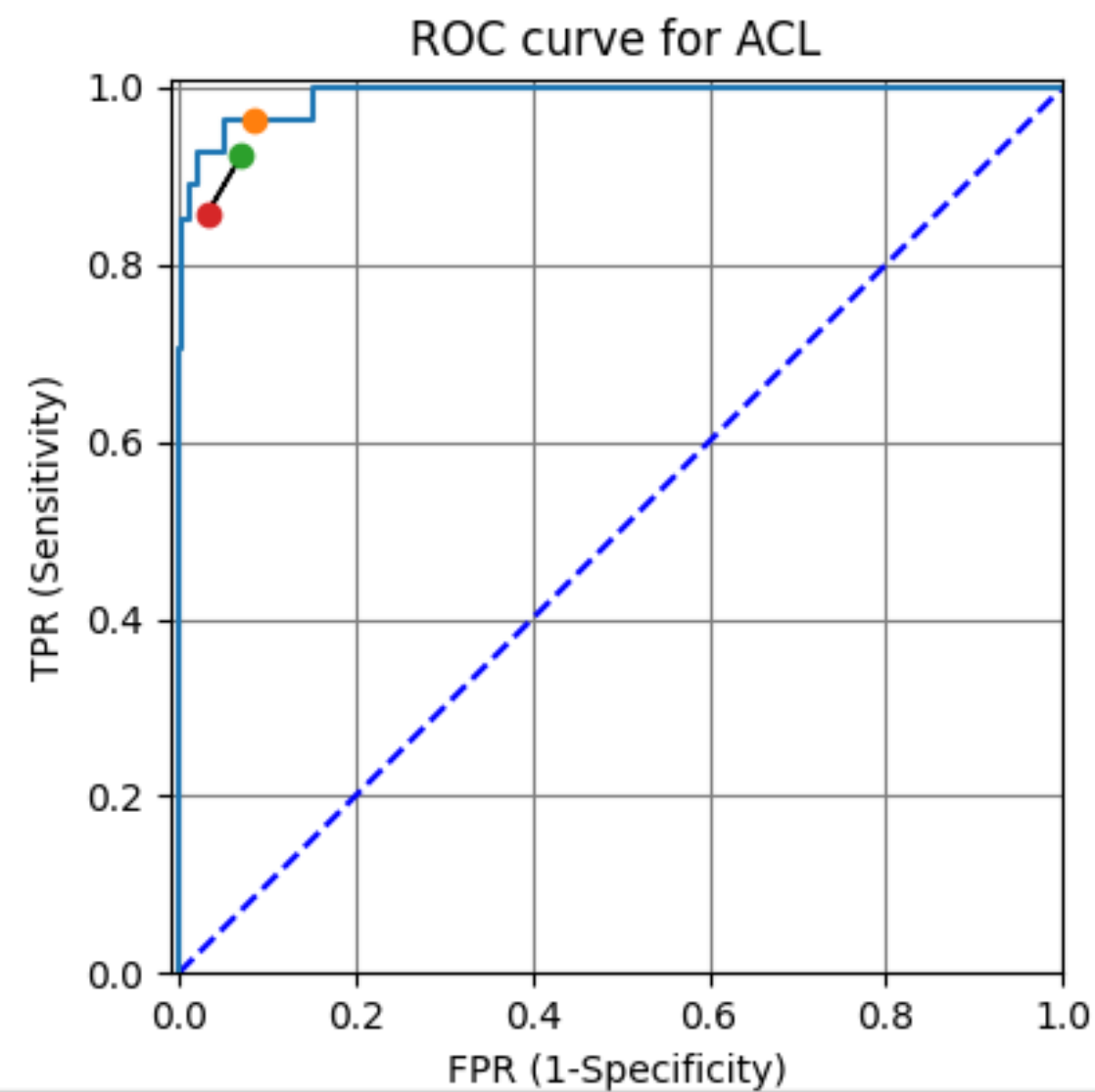
NARROW AI: LIMITATIONS



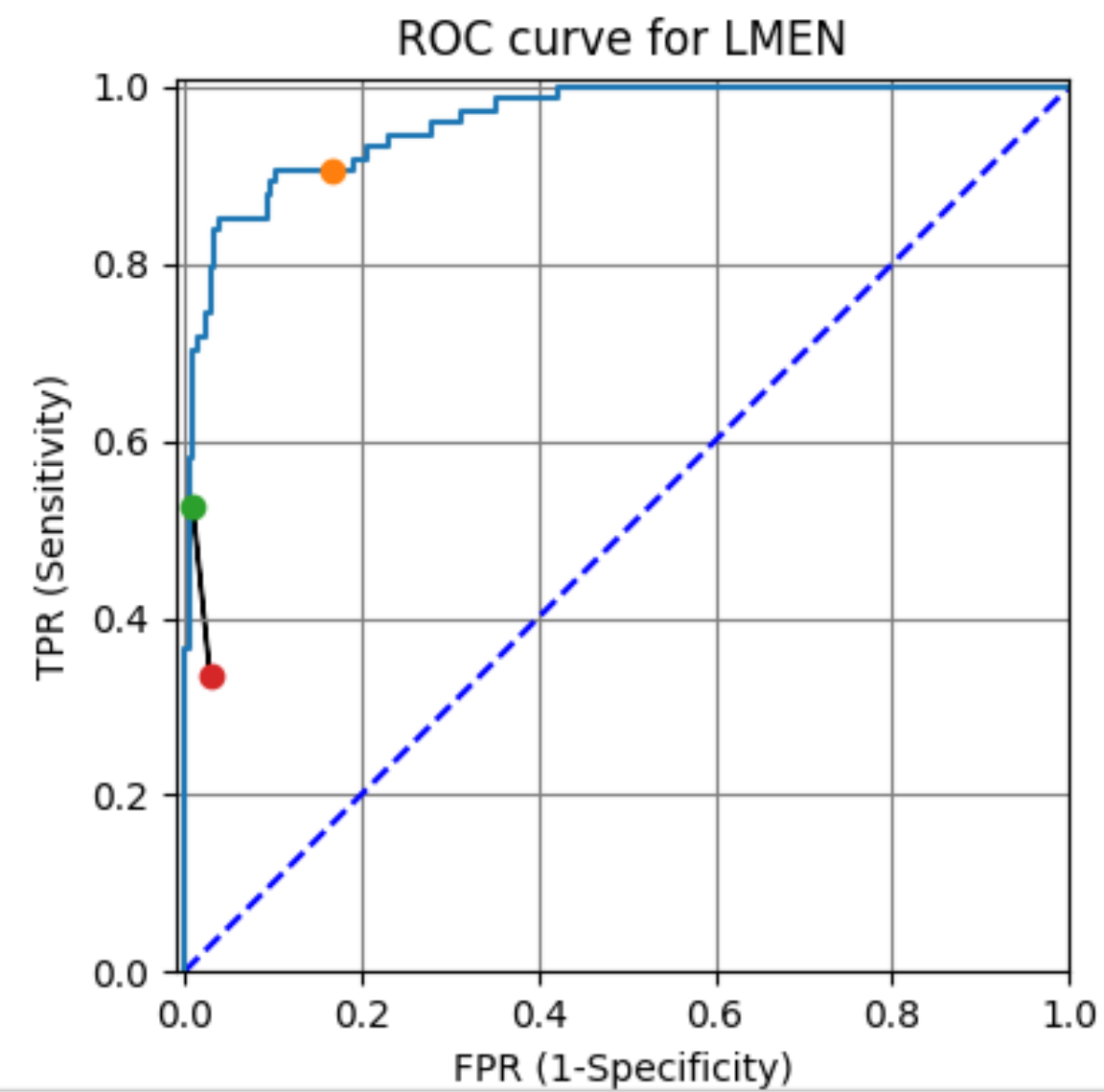
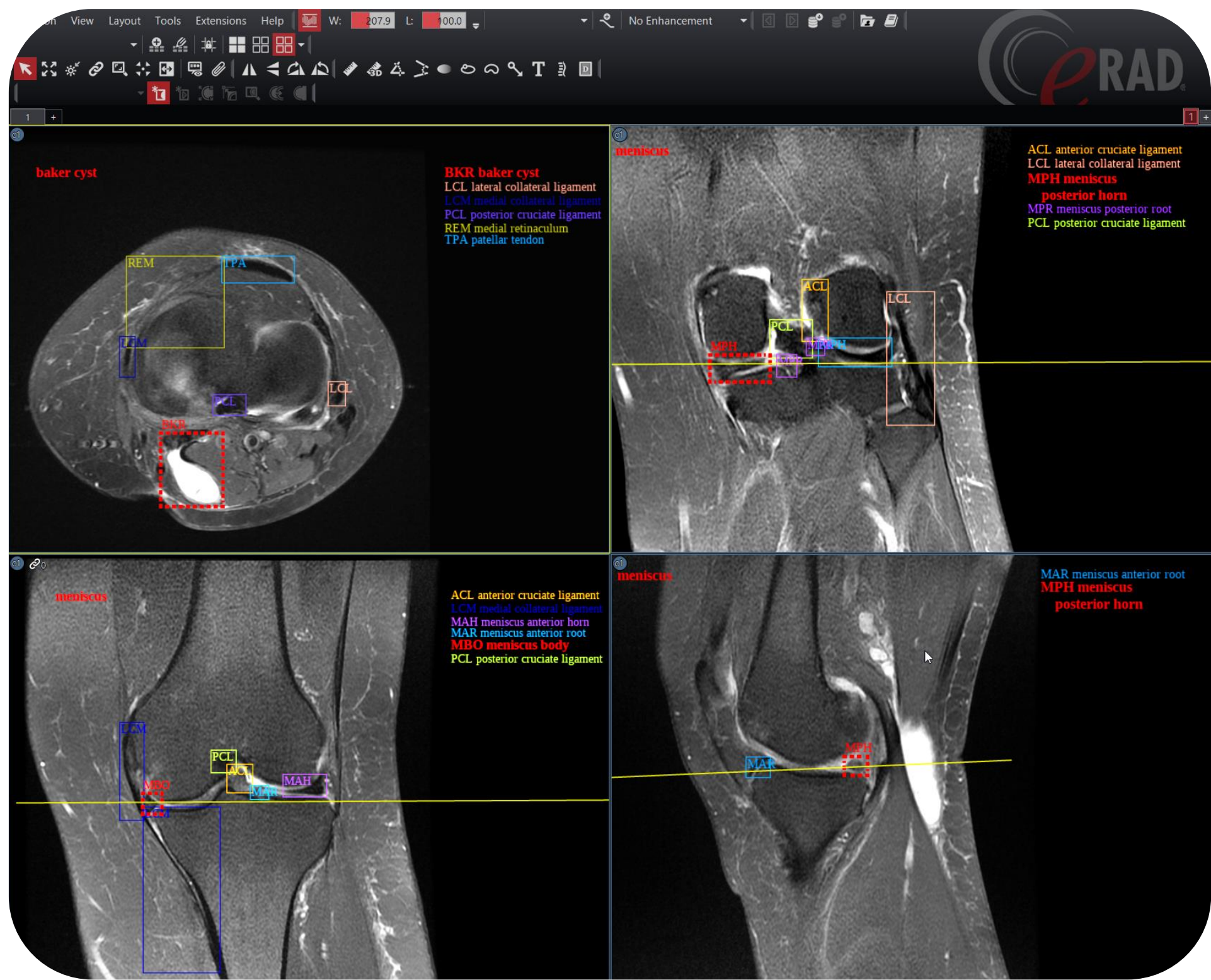
Class IIa: diagnostic support

 <p>Intracranial Hemorrhage 510(k) Triage and notification software indicated for use in the analysis of non-enhanced head CT images; flags and communicates suspected positive findings of pathologies in head CT images, namely Intracranial Hemorrhage (ICH).</p>	 <p>Acute C-Spine Fractures 510(k) Triage and notification software indicated for use in the analysis of cervical spine CT images; flags and communicates suspected positive findings of linear lucencies in the cervical spine bone in patterns compatible with fractures.</p>	 <p>M1 Large Vessel Occlusions 510(k) Triage and notification software indicated for use in the analysis of head CTA images; flags and communicates suspected positive findings of M1 Large Vessel Occlusion (M1 LVO).</p>	 <p>Pulmonary Embolism 510(k) Triage and notification software indicated for use in the analysis of CTPA images; flags and communicates Pulmonary Embolism (PE).</p>
 <p>Vessel Occlusion 510(k) Triage and notification software indicated for use in the analysis of head CTA images; flags and communicates suspected positive findings of Vessel Occlusion.</p>	 <p>Intra-Abdominal Free Gas 510(k) Triage and notification software indicated for use in the analysis of abdomen CT images; flags and communicates suspected positive cases of Intra- Abdominal Free Gas (IFG).</p>	 <p>Aortic Dissection 510(k) Triage and notification software indicated for use in the analysis of CT exams with contrast that includes the chest; flags and communicates suspected positive findings of Aortic Dissection.</p>	 <p>Incidental PE 510(k) Triage and notification software indicated for use in the analysis of CT images (not dedicated CTPA protocol) ; flags and communicates incidental Pulmonary Embolism (PE).</p>
 <p>Pneumothorax 510(k) Triage and notification software indicated for use in the analysis of Chest X-Ray images; flags and communicates Pneumothorax (Ptx).</p>	 <p>Rib Fractures 510(k) Triage and notification software indicated for use in the analysis of chest CTs (with or without contrast); flags and communicates suspected cases of three or more acute Rib fracture (RibFx) pathologies.</p>	 <p>Brain Aneurysm 510(k) Triage and notification software indicated for use in the analysis of Head CTA images; flags and communicates Brain Aneurysm (BA).</p>	 <p>Malpositioned Endotracheal Tubes (ETT) 510(k) Triage and notification software indicated for use in the analysis of Frontal Chest X-Ray images; flags and communicates suspected positive cases of vertically malpositioned endotracheal tube (ETT) in relation to the carina.</p>
 <p>Acute C-Spine Fractures 510(k) Triage and notification software indicated for use in the analysis of cervical spine CT images; flags and communicates suspected positive findings of linear lucencies in the cervical spine bone in patterns compatible with fractures.</p>	 <p>Vertebral Fractures Compression 510(k) Triage and notification software indicated for use in the analysis of chest and abdominal CT images; flags and communicates suspected positive cases of Vertebral Compression Fractures (VCFx) findings.</p>	 <p>Pulmonary Embolism 510(k) Triage and notification software indicated for use in the analysis of CTPA images; flags and communicates Pulmonary Embolism (PE).</p>	  <p>Na, K, Cr</p>

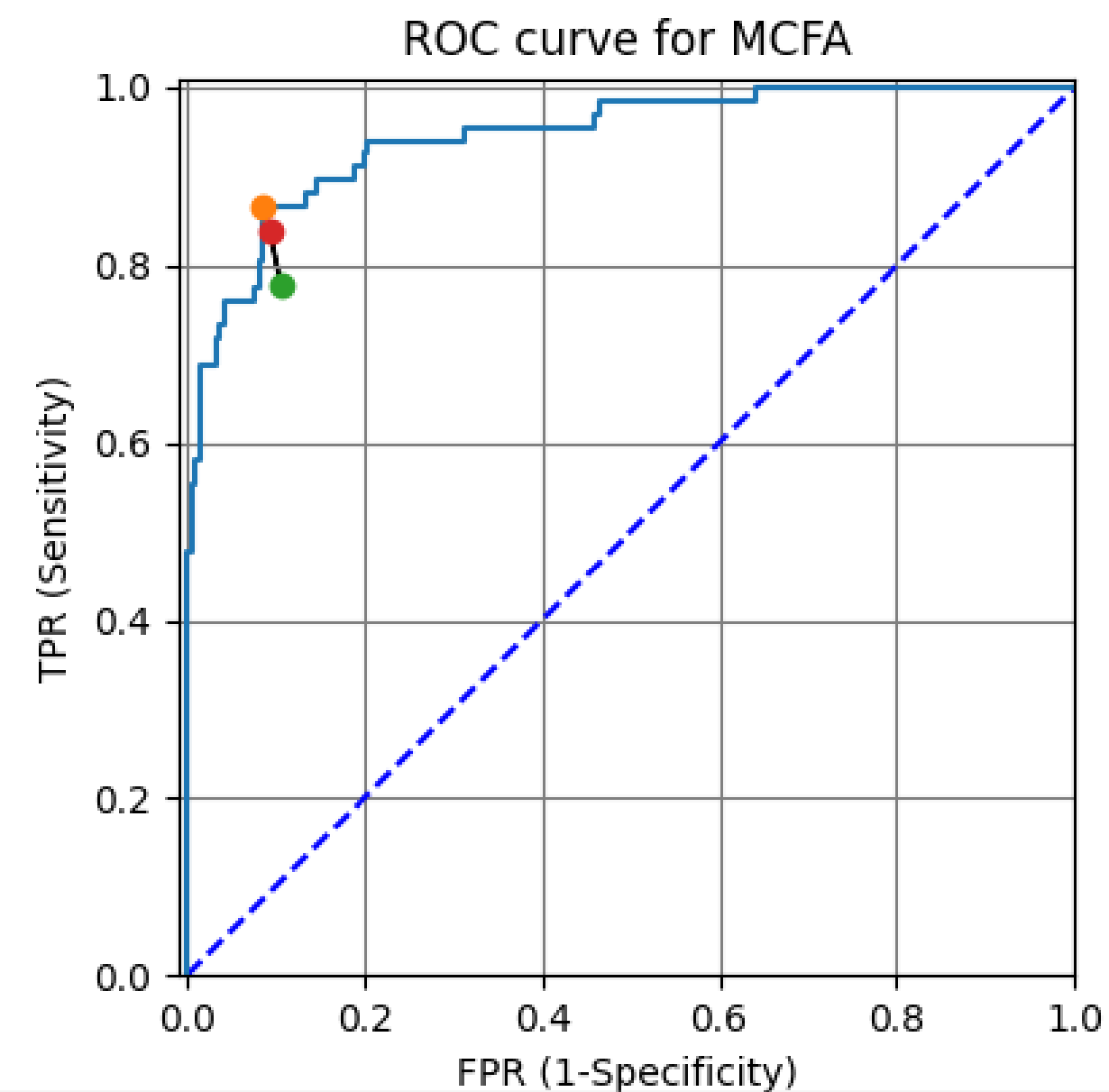
AI-HUMAN INTERACTION



Legend: random classifier (dashed blue line), ROC (AUC=0.991) (solid blue line), Without AI (red dot), With AI (green dot), AI thresholded (orange dot).

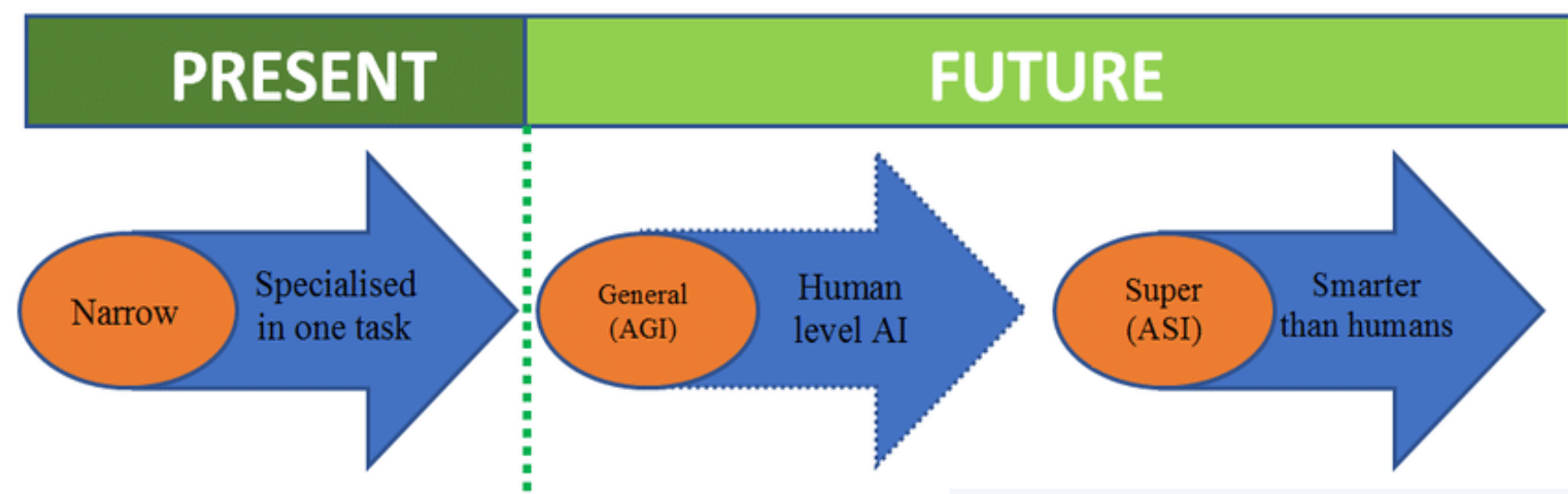


Legend: random classifier (dashed blue line), ROC (AUC=0.961) (solid blue line), Without AI (red dot), With AI (green dot), AI thresholded (orange dot).



Legend: random classifier (dashed blue line), ROC (AUC=0.945) (solid blue line), Without AI (red dot), With AI (green dot), AI thresholded (orange dot).

GENERAL AI: RESPONSIBILITY?



Class IIb: autonomous

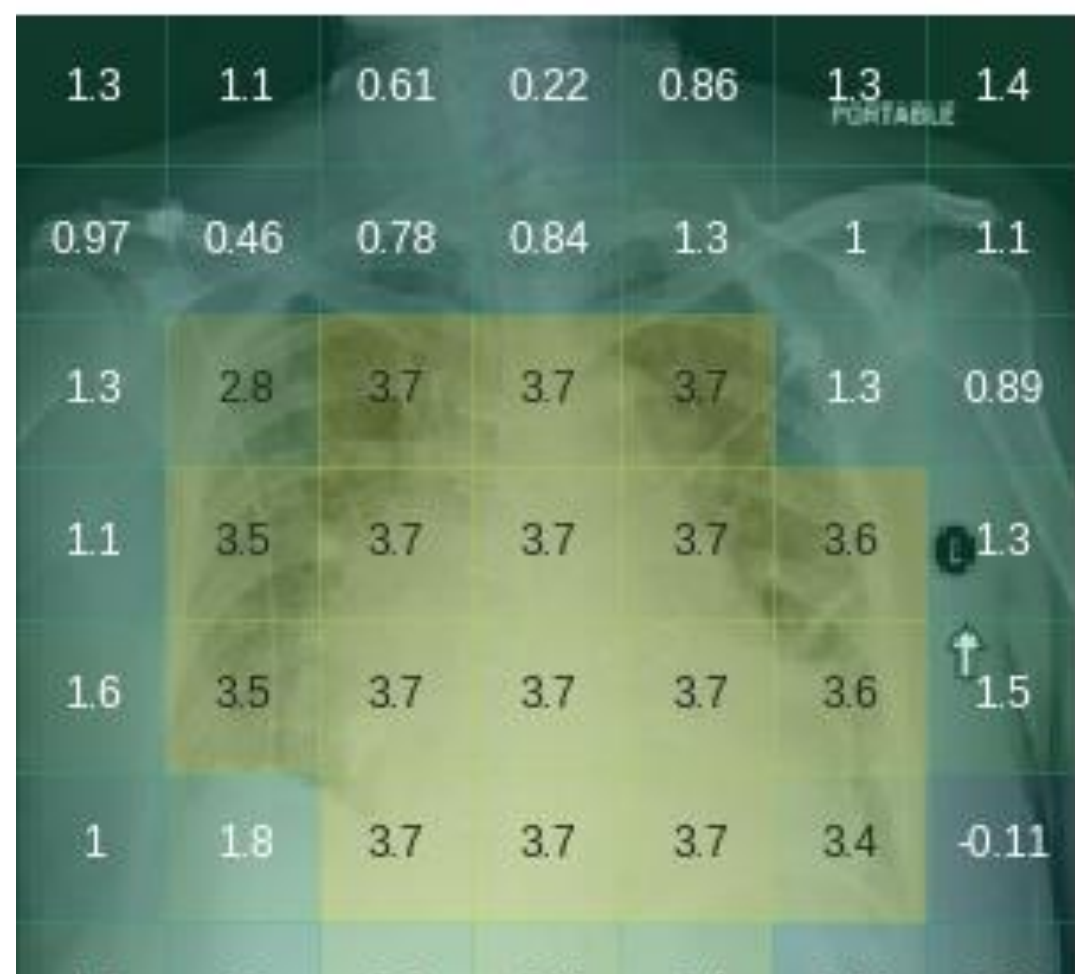


Who signs the report?

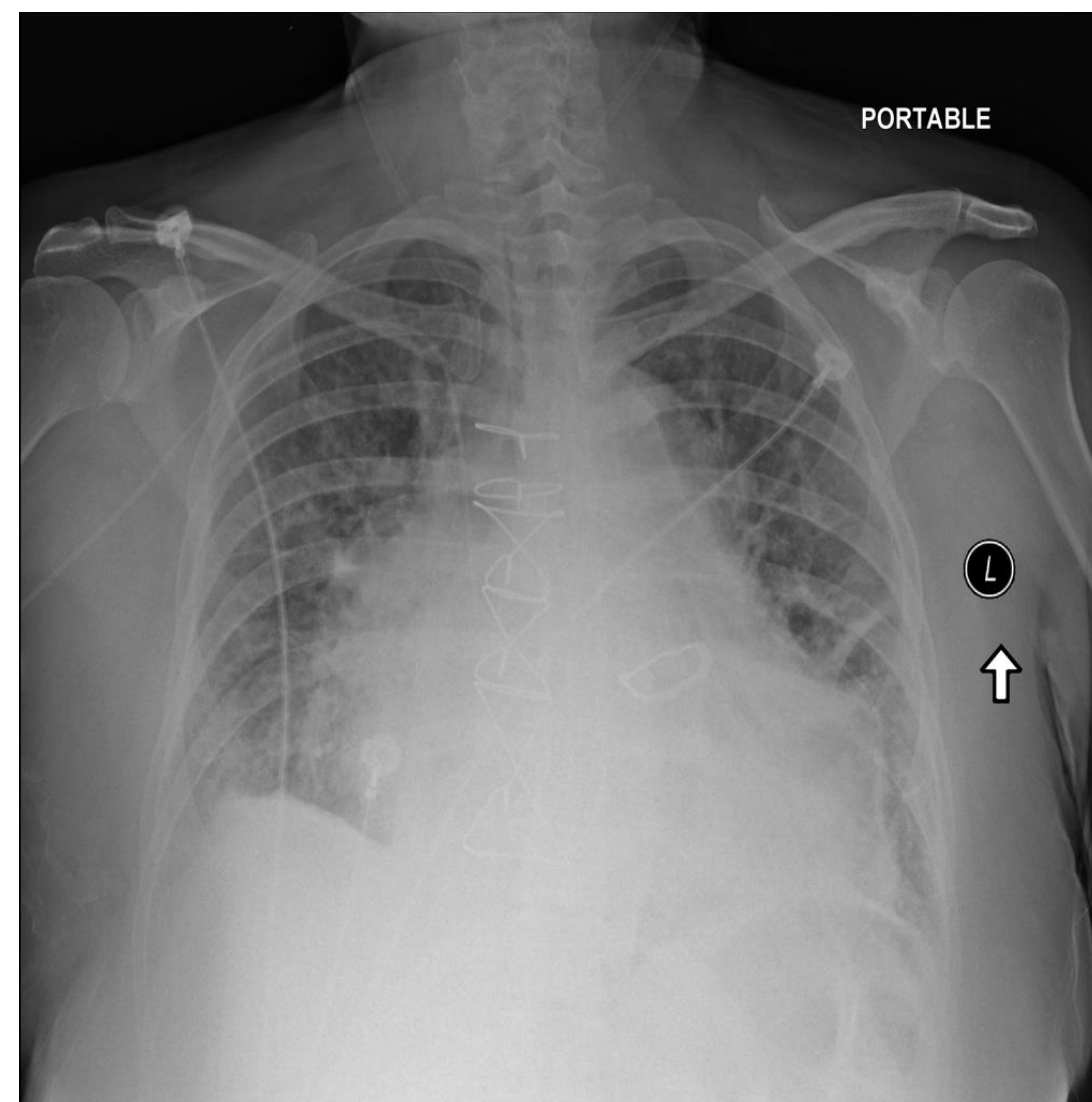
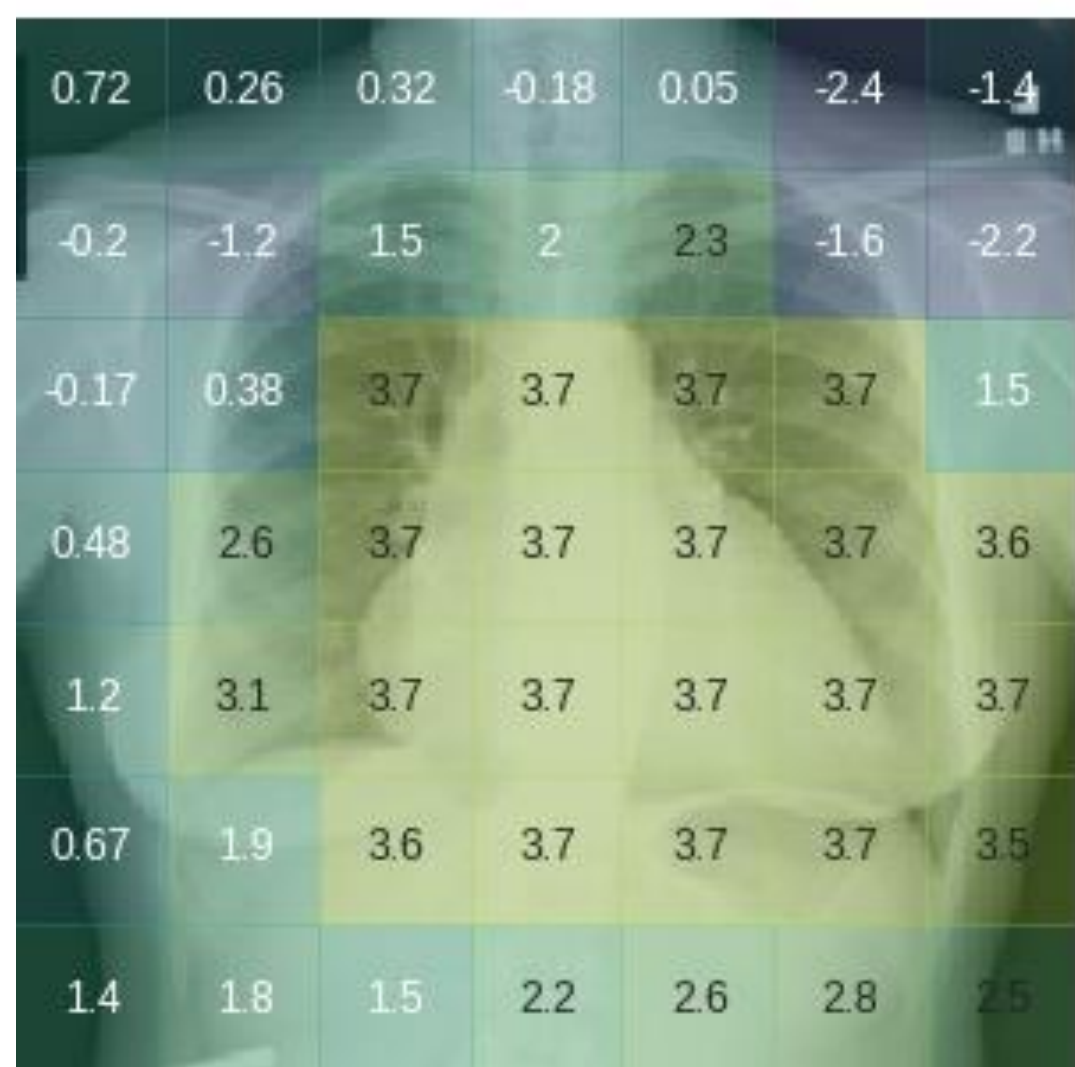
AI mistakes

TYPICAL AI MISTAKES

P(Cardiomegaly)=0.752



P(Cardiomegaly)=0.937



Zech et al. PLOS Medicine 2018

CONCLUSION

ML can surpass human level performance in certain, **narrowly defined** areas (narrow AI)

- Diagnostic/imaging reports **accelerated, precision** improved (segmentation, ECG)
- Correlation/regression information exploration (prediction, age assessment)
- Requirements: big database, standardized protocols, unified annotation

Ethical issues:

- low cost vs. narrow solution
- Responsibility?





Thank you for your attention!