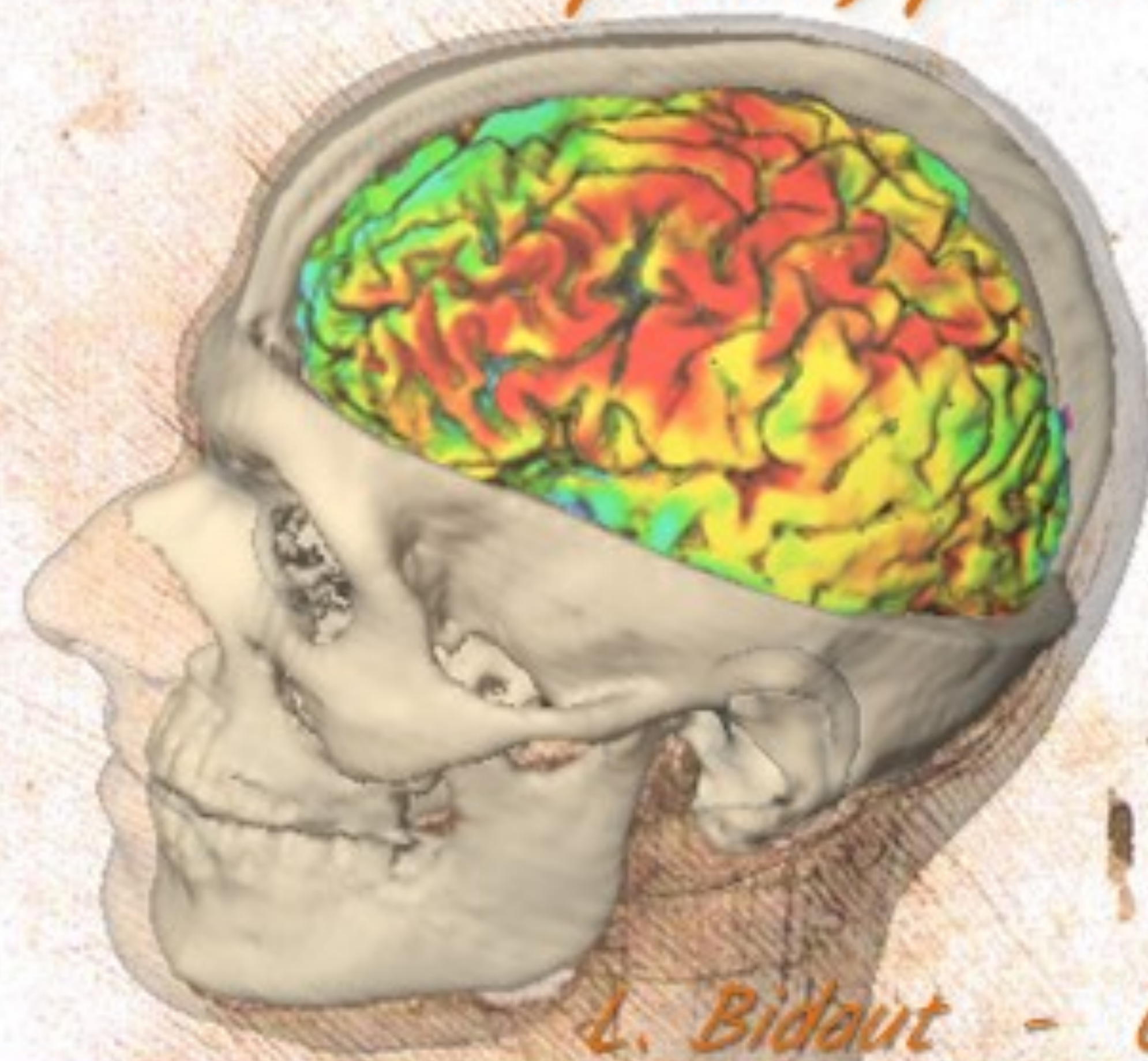


Software-based image registration:

past, present, future...



L. Bidaut - University of Lincoln

Once upon a time: Wilhelm Roentgen discovers X-rays in 1895 while experimenting on his wife Anna-Berthe ...

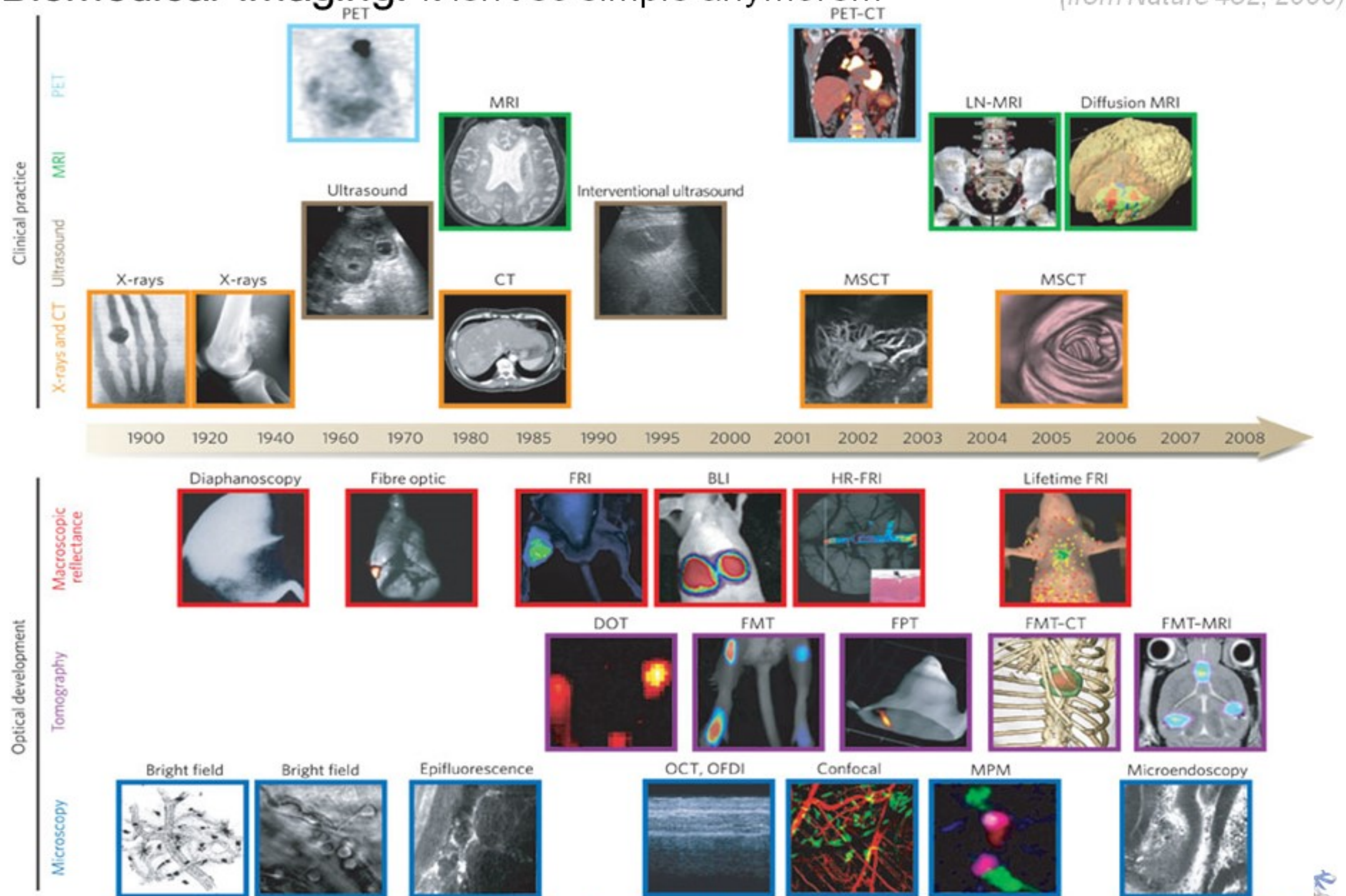


and the rest is History...




Biomedical imaging: it isn't so simple anymore...

(from Nature 452, 2008)



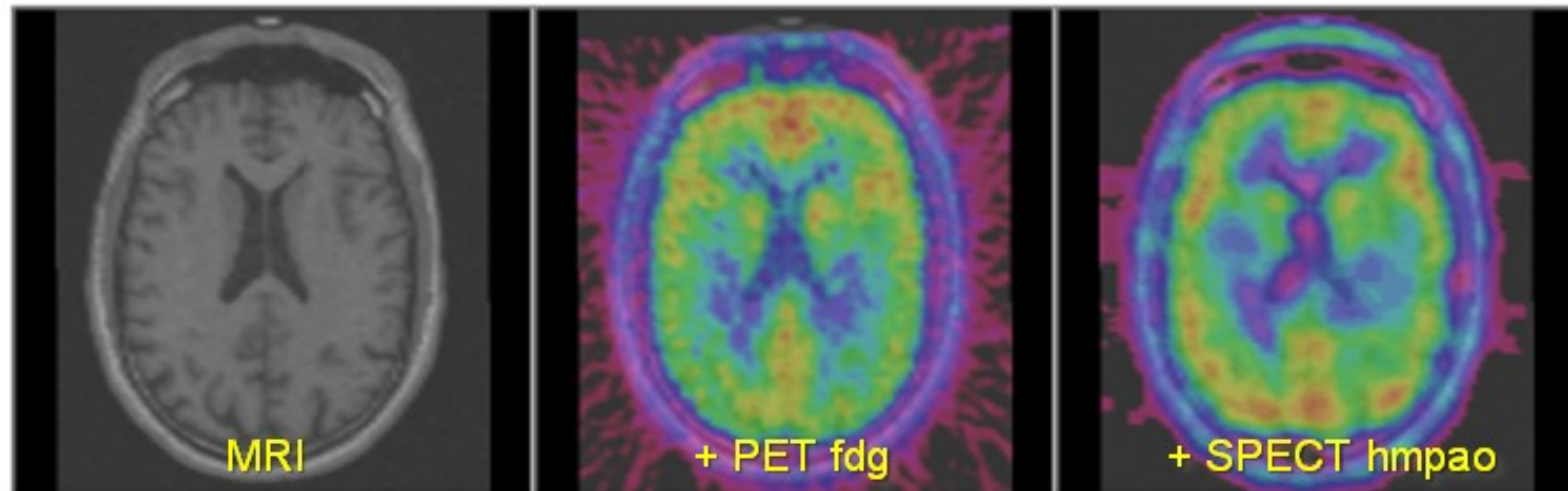
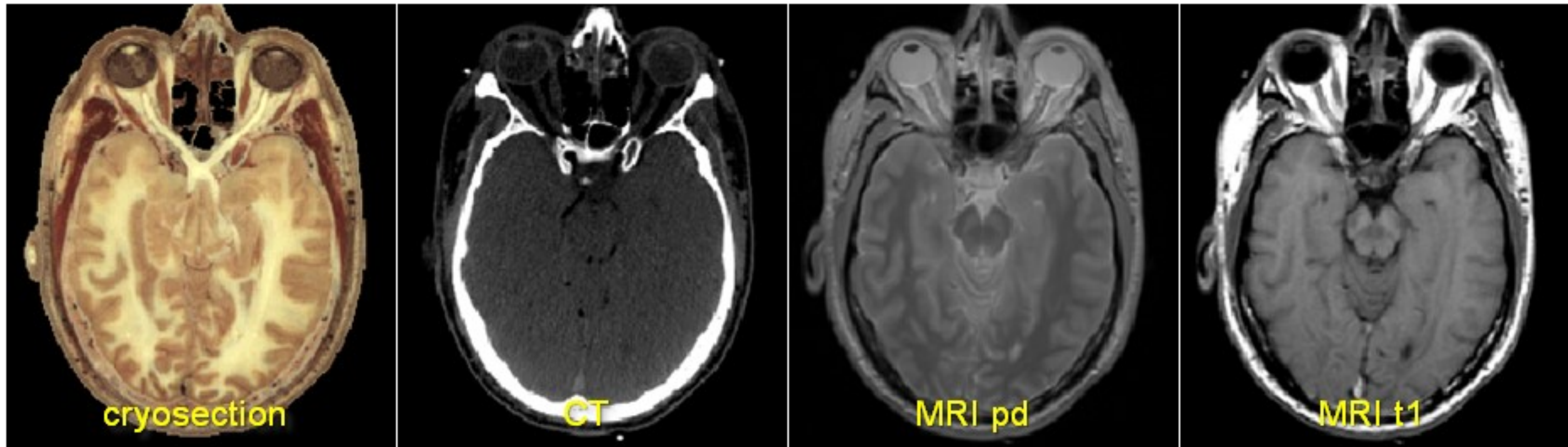
... differing principles and scales (both spatial and time)
=> show different things...

L.Bidaut – Plovdiv 20190917



Multimodality/-sensor \leq modalities

(collab. VH - NLM, HUG)



... differing principles and scales (both spatial and time)
=> show different things...

assumption: "the sum is greater than the parts"

Multi-modality imaging

- Many different types of images
- They have different properties, such as resolution, contrast, property being imaged such as anatomy vs. physiology
- Combining such information can provide additional information (e.g., sum greater than the parts)



need for **fusion**



Various clinical imaging data sets

- Each modality (data set) has its own spatial reference system and, to make things slightly worse, time and physiology also matter...



need for **registration**



Multimodality Imaging: 2 complementary concepts

- **Registration**



$\langle = \rangle$ establishes the spatial relationship between two (or more) data sets

- **Fusion**

$\langle = \rangle$ combines data from different sources, times, procedures for extracting and exploiting complementary information



Registration/Fusion contexts:

- Intra-patient: same patient
- Inter-patient: different patients
- Intra-modality: same modality (e.g., protocol)
- Inter-modality: different modalities (or protocol)

=>> research or clinical registration can be of interest for any combination(s) of the above



Registration techniques

Goal:

find the optimal geometric transformation so that
a location (point, pixel, voxel) in one data set
can be linked in a unique and reversible manner
to the corresponding one in the other data set

Means:

images
points, landmarks, features (surfaces, etc.)
contents



Multimodality registration:

chain of process; most relevant items between parentheses

acquisition, transfer -> reference data set and source one

conversion (e.g., Dicom to other)

segmentation (various techniques; manual to semi-auto.)

estimate/initialize current geometric transformation (# of DOF: e.g., rigid vs non-rigid)

transform source and compare to ref. (geometric transformation: DOFs, interpolation, etc.)

compare source and ref. (cost definition, e.g., analog to a distance)

optimization (iterative techniques; multiscale, constraints, multicost, etc.
to avoid local extremas)

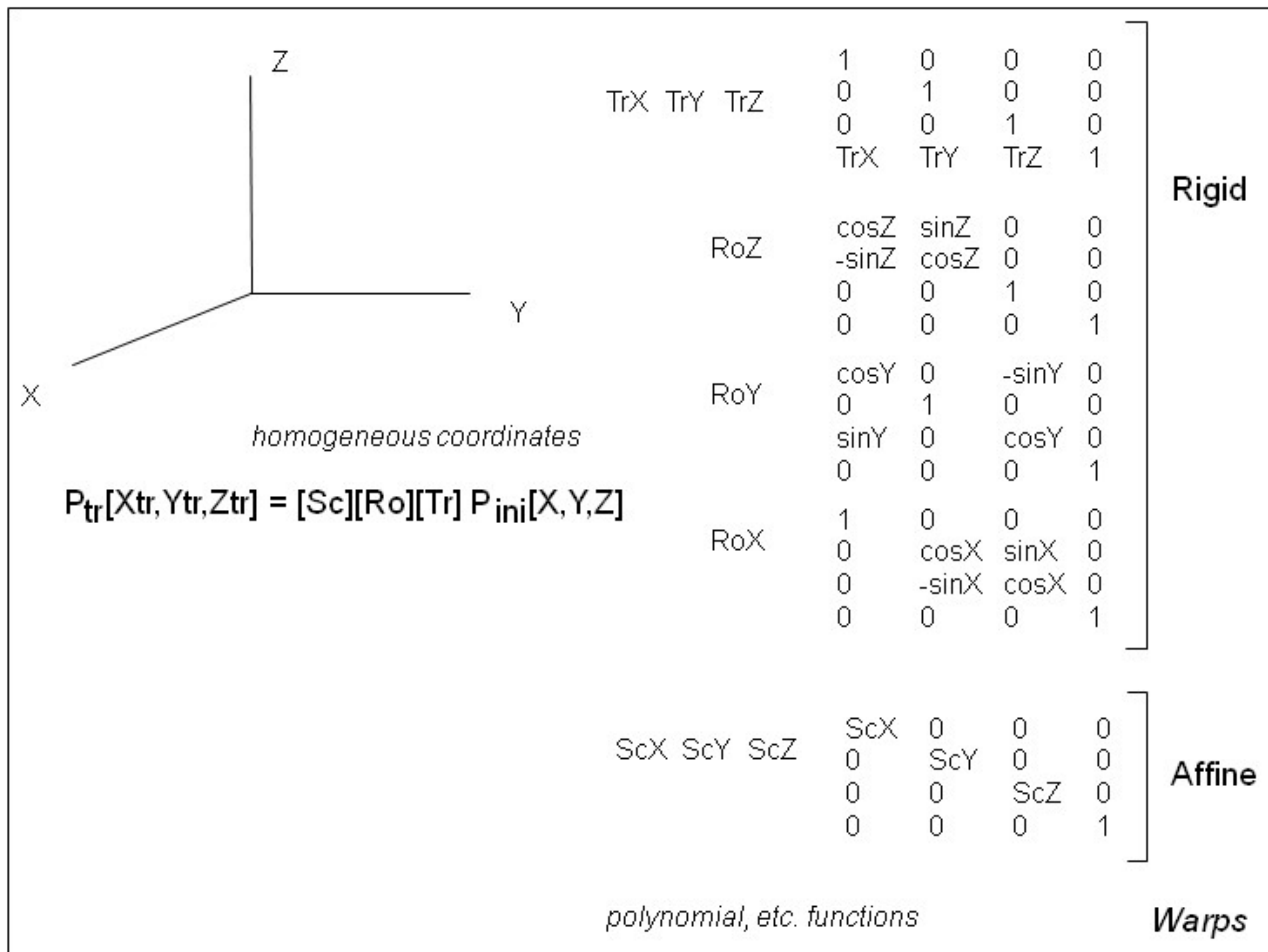
assess results (visual, stats, etc.)

visualize (various paradigms)

exploit ...



Registration techniques: 3D transformation:



Registration techniques: costs based on:

- images: interactive techniques
 - => minimize visual differences between paired images
- points, landmarks, features (lines, planes, surfaces): direct or iterative techniques
 - => minimize distance analog between paired features
- contents: iterative techniques

values: differences, ratios

=> minimize sum or variance of squared ds. or rs.

probability distribution: statistics

entropy: measure of randomness

$$H(X) = -E_x[\log(P_x)] = -\sum[P_x \log(P_x)]$$

joint entropy: measure of Y's randomness given X => minimize JE

$$H(X,Y) = -E_x[E_y[\log(P(X,Y))]] = H(X)+H(Y) \text{ if } X \text{ and } Y \text{ are independent}$$

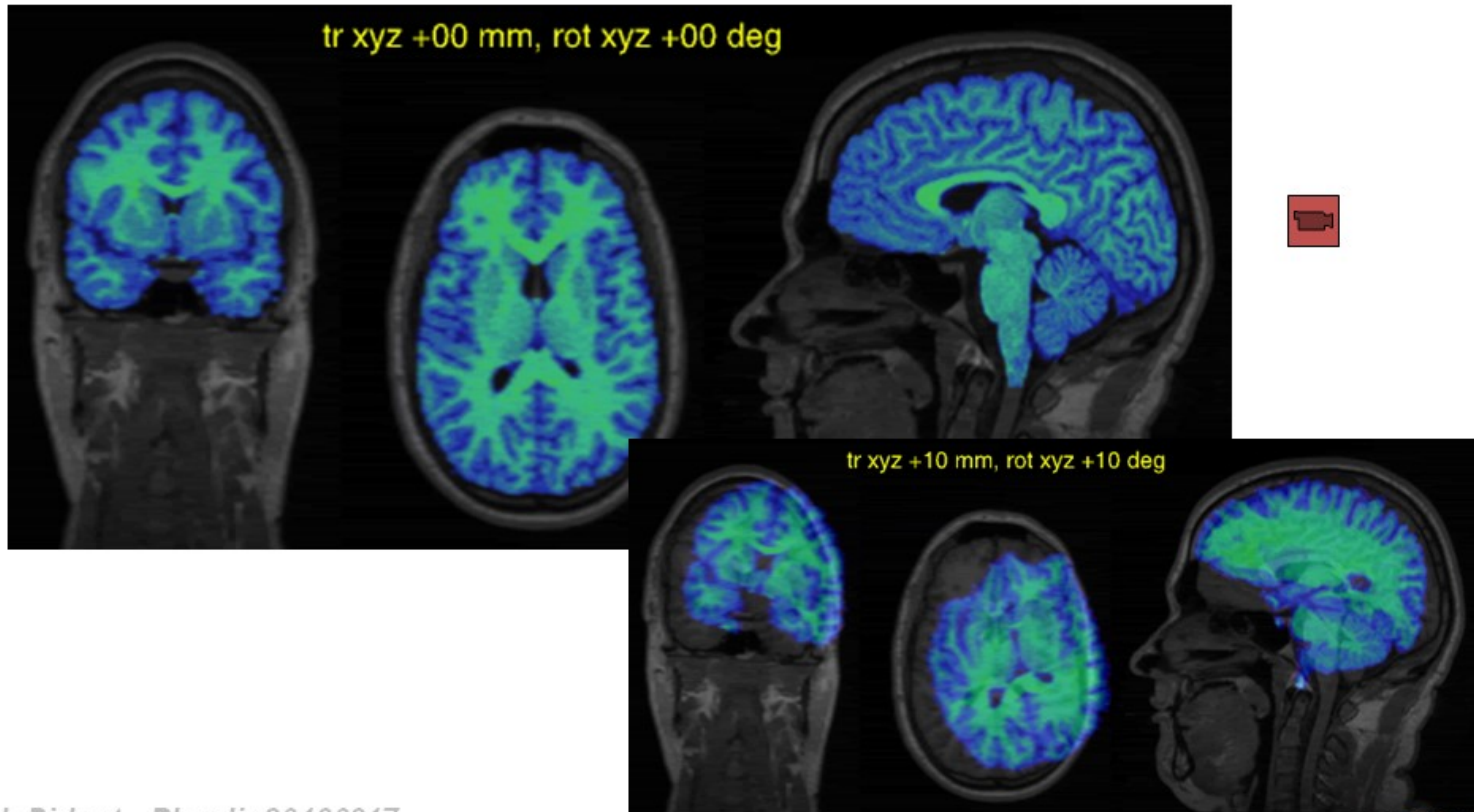
mutual information: measure of the reduction of H_y given X => maximize MI

$$MI(X,Y) = H(X) + H(Y) - H(X,Y) \quad NMI:(X,Y) = H(X)H(Y)/H(X,Y)$$



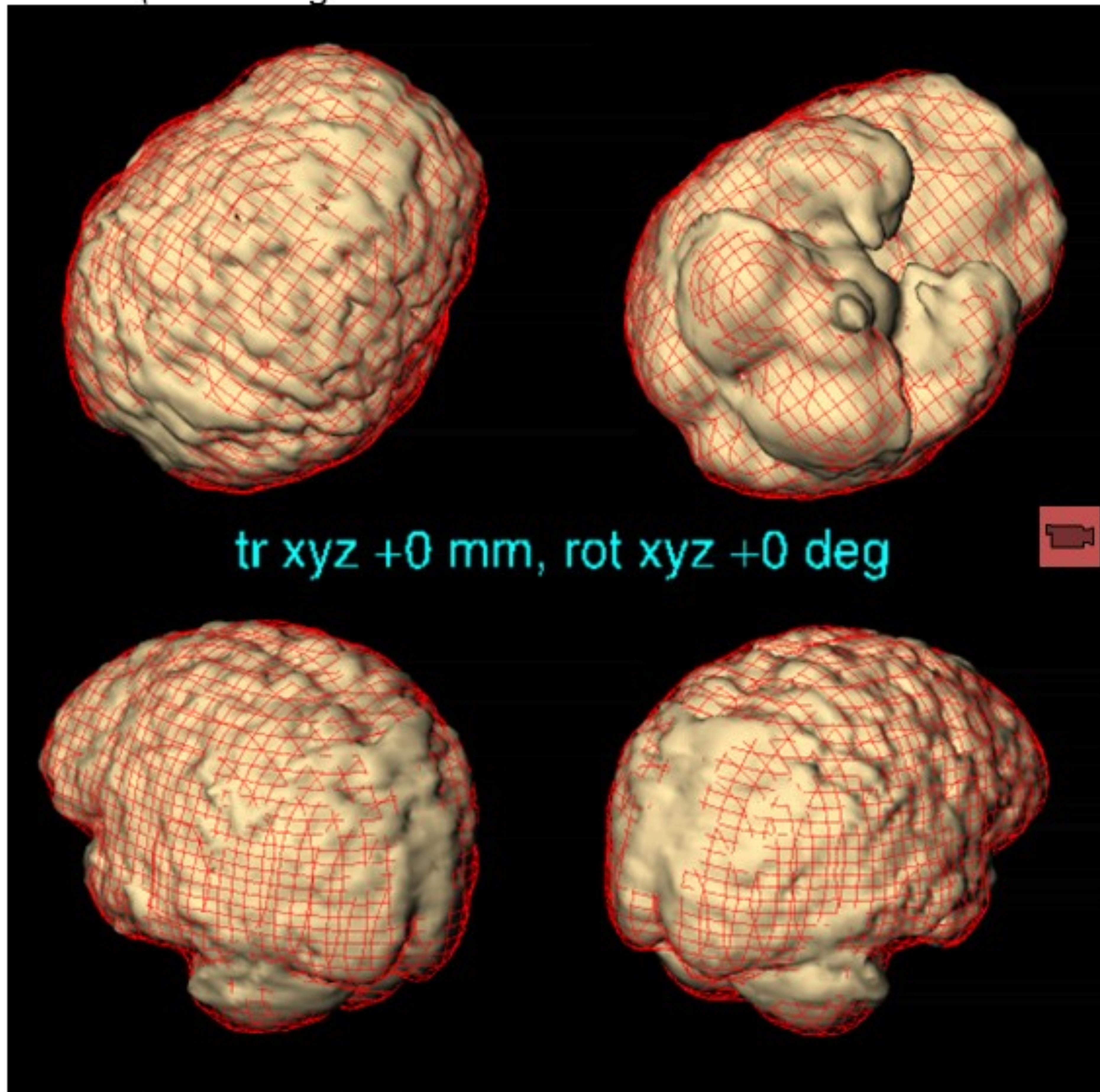
Registration techniques: cost based on:

- images: interactive techniques
=> minimize visual differences between paired images
(assuming that similar visual clues are visible on both IM_{ref} and IM_{ali})

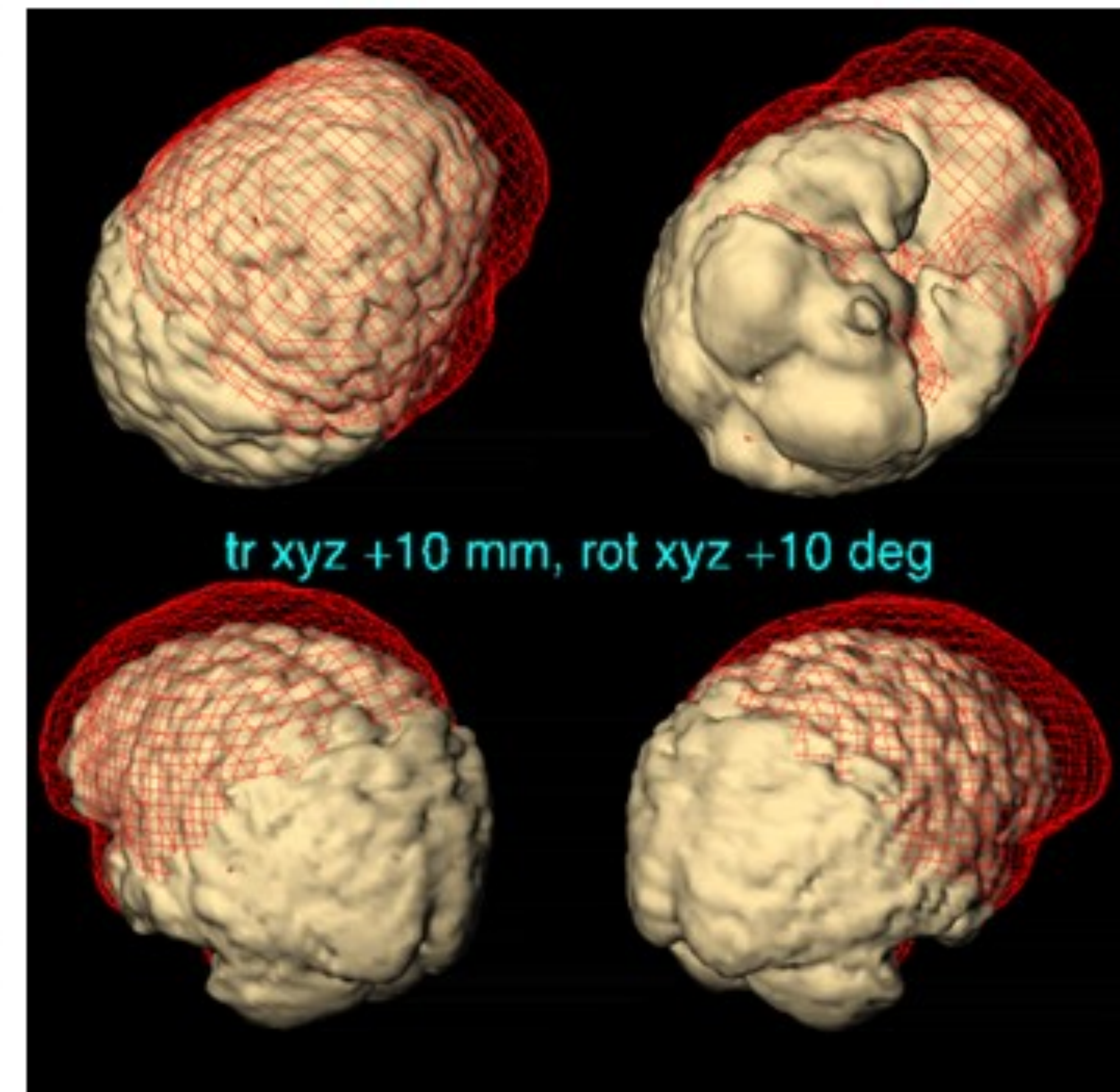


Registration techniques: cost based on:




- points, landmarks, features (lines, planes, surfaces): direct or iterative techniques
=> minimize distance (euclidean, chamfer, etc.) analog between paired features
(assuming that the same relevant features* can be extracted in both IM_{ref} and IM_{ali})



*intrinsic: belong to the data set
extrinsic: e.g., artificial markers, apparatus (stereotaxy,...), etc.

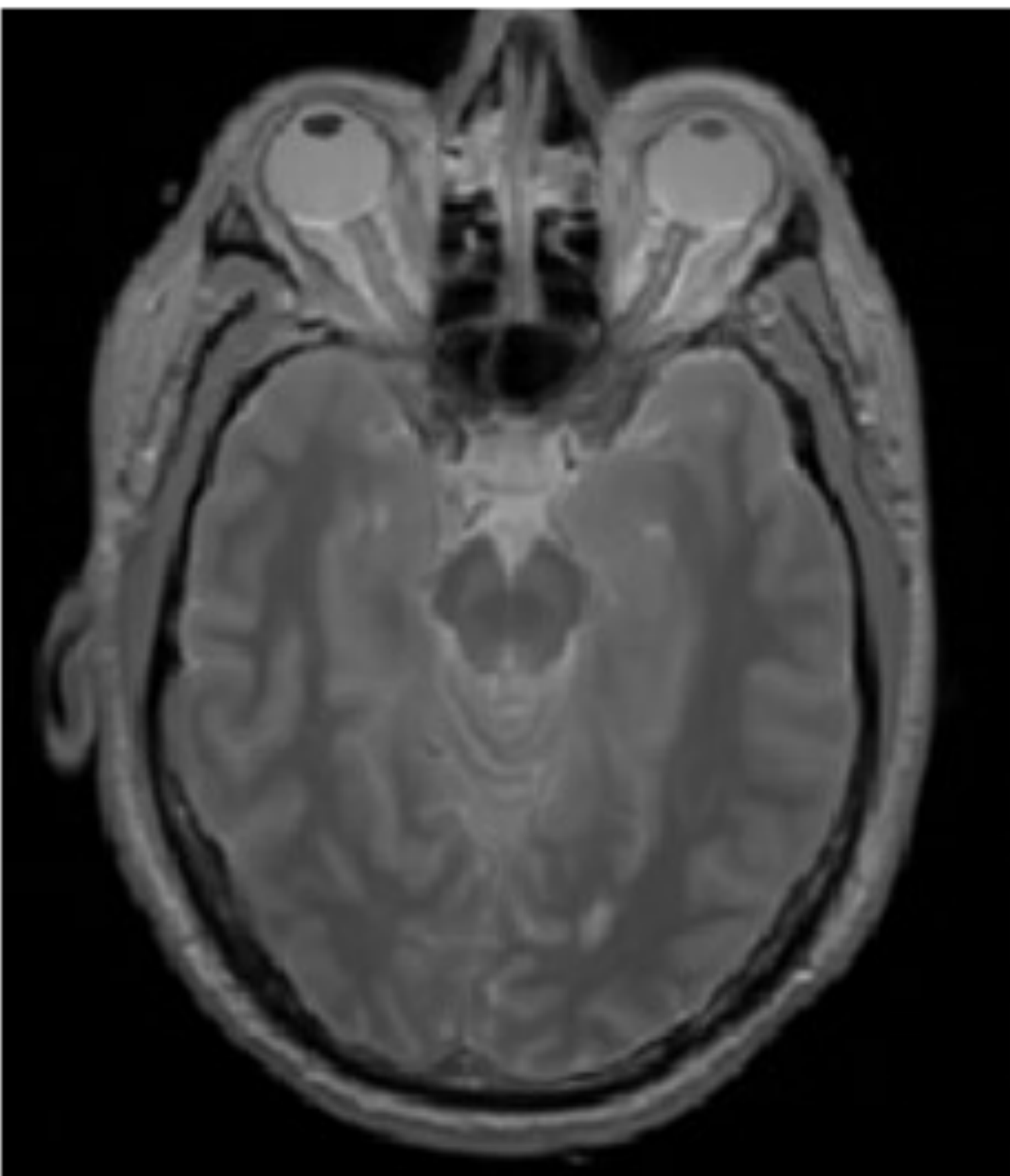


Registration techniques: cost based on:

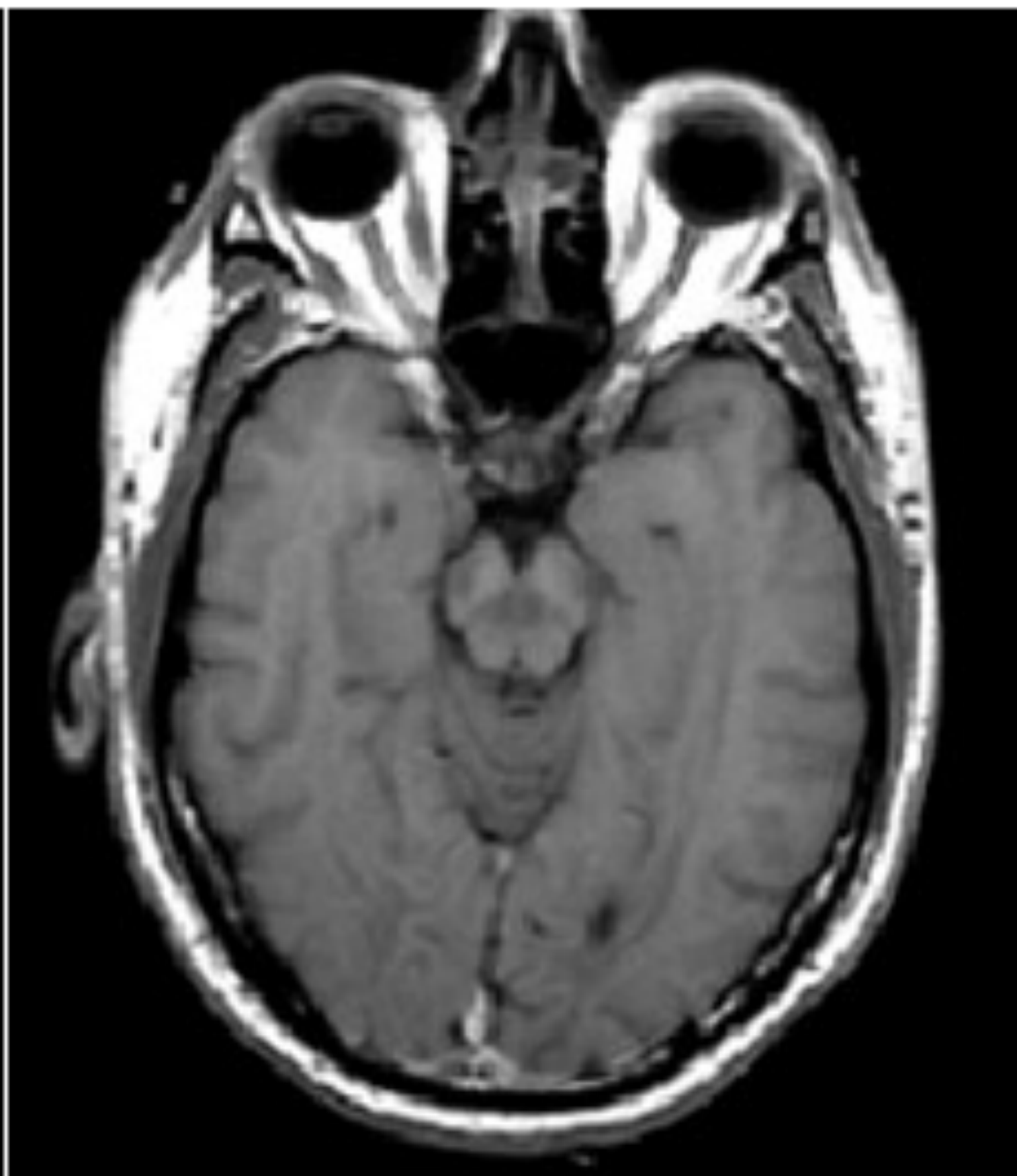
- contents: iterative techniques  *MR/CT iter*   *MR/PT iter*
direct values' comparison: differences (*assuming that $IMali \approx IMref$*)
or ratios (*assuming that $IMali \approx a IMref + b$*)
in relevant tissue classes/structures
=> minimize sum or variance of squared ds. or rs.



CT






MRI pd



MRI t1



Registration techniques: cost based on:

- contents: iterative techniques  *MR/CT iter*   *MR/PT iter*

probability distribution: statistics (*assuming that $IM_{ali} \approx IM_{ref}$*)

entropy: measure of randomness

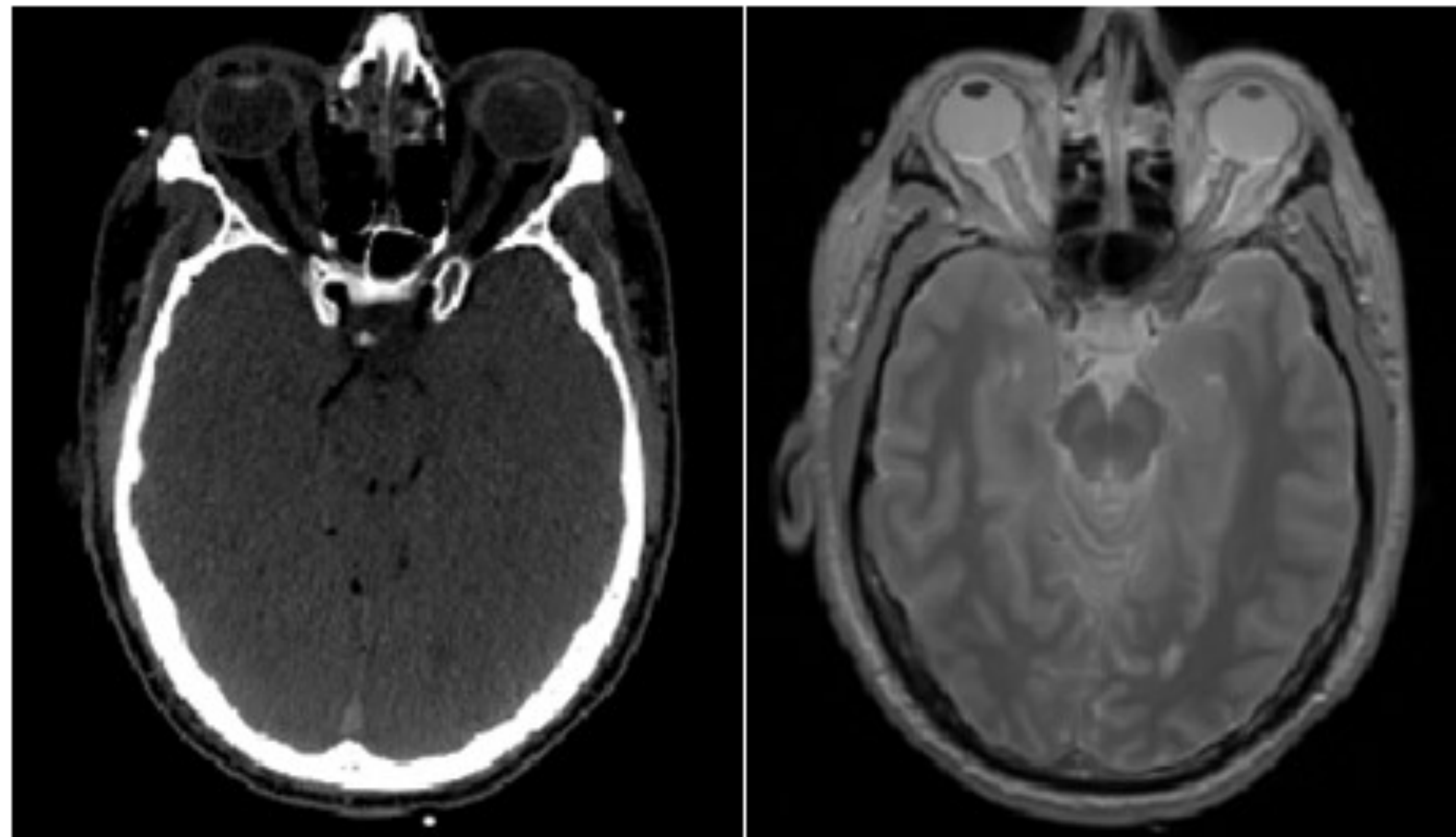
$$H(X) = -E_x[\log(P_x)] = -\sum[P_x \log(P_x)]$$

joint entropy: measure of Y's randomness given X \Rightarrow minimize JE

$$H(X,Y) = -E_x[E_y[\log(P(X,Y))]] = H(X)+H(Y) \text{ if } X \text{ and } Y \text{ are independent}$$

mutual information: measure of the reduction of H_y given X \Rightarrow maximize MI

$$MI(X,Y) = H(X) + H(Y) - H(X,Y) \quad NMI:(X,Y) = H(X)H(Y)/H(X,Y)$$

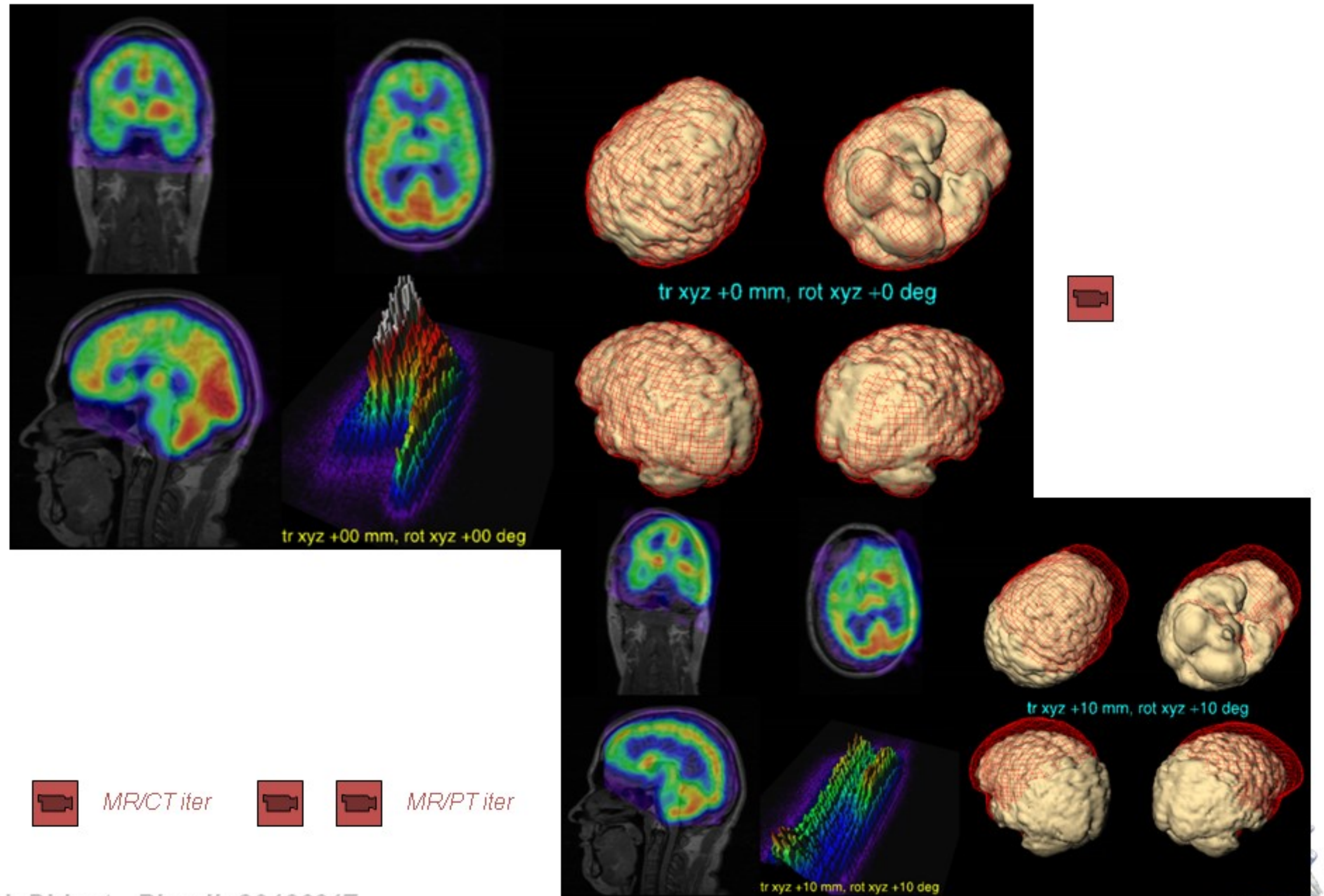


CT

MRI pd



Registration techniques: summary



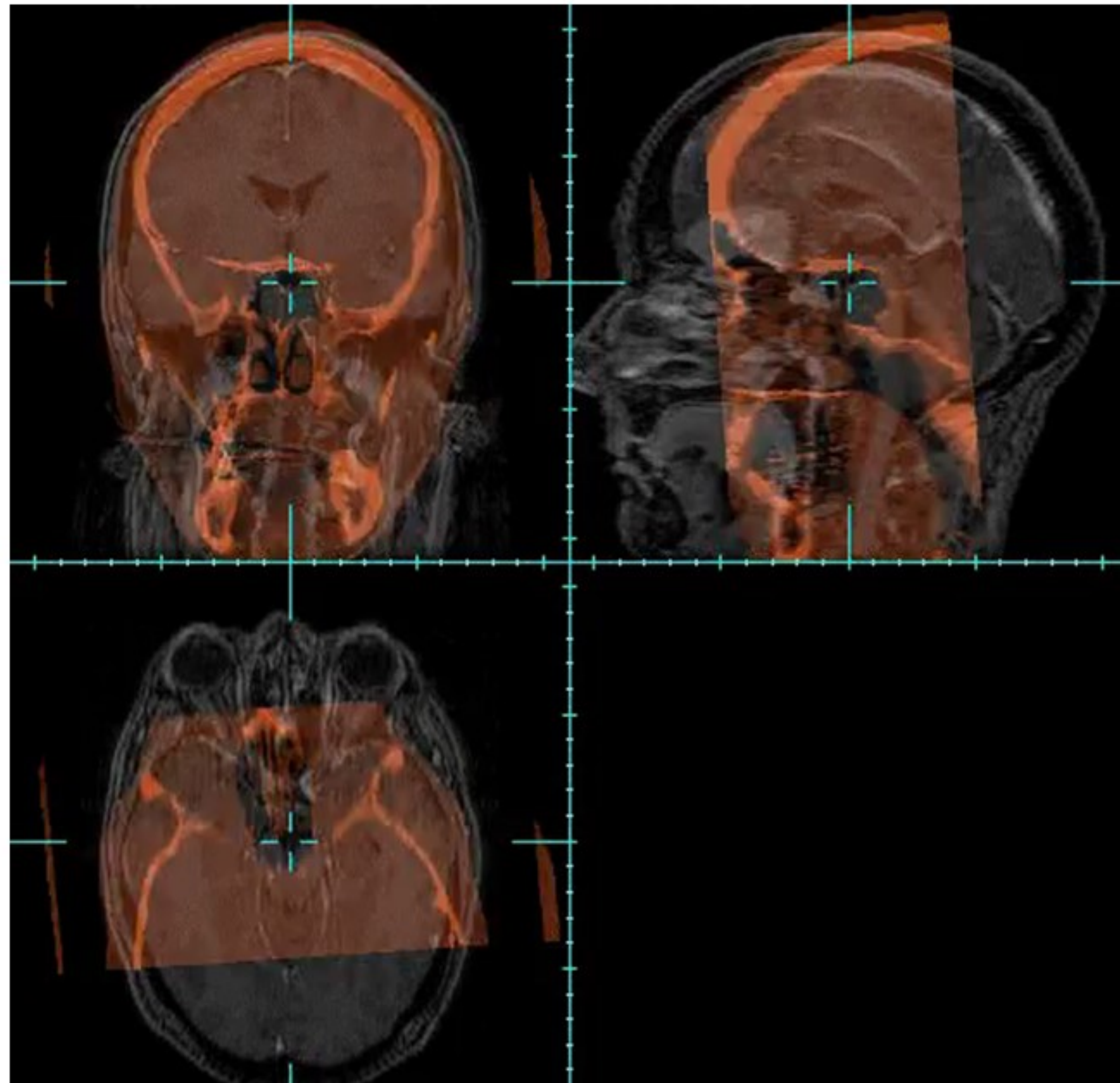
Iterative techniques

To optimize the search for the optimal geometric registration transformation and ensure its convergence, e.g. through:

- multiresolution/multiscale approaches
- additional (adaptative) constraints to restrict the search (parameter) space
- "intelligent" use of various optimization strategies and/or cost functions
- initialization close enough to the likely solution, e.g. via controlled positioning and/or processing/registration sequence
- etc.



Iterative registration (CT to MR):



Non-rigid registration:

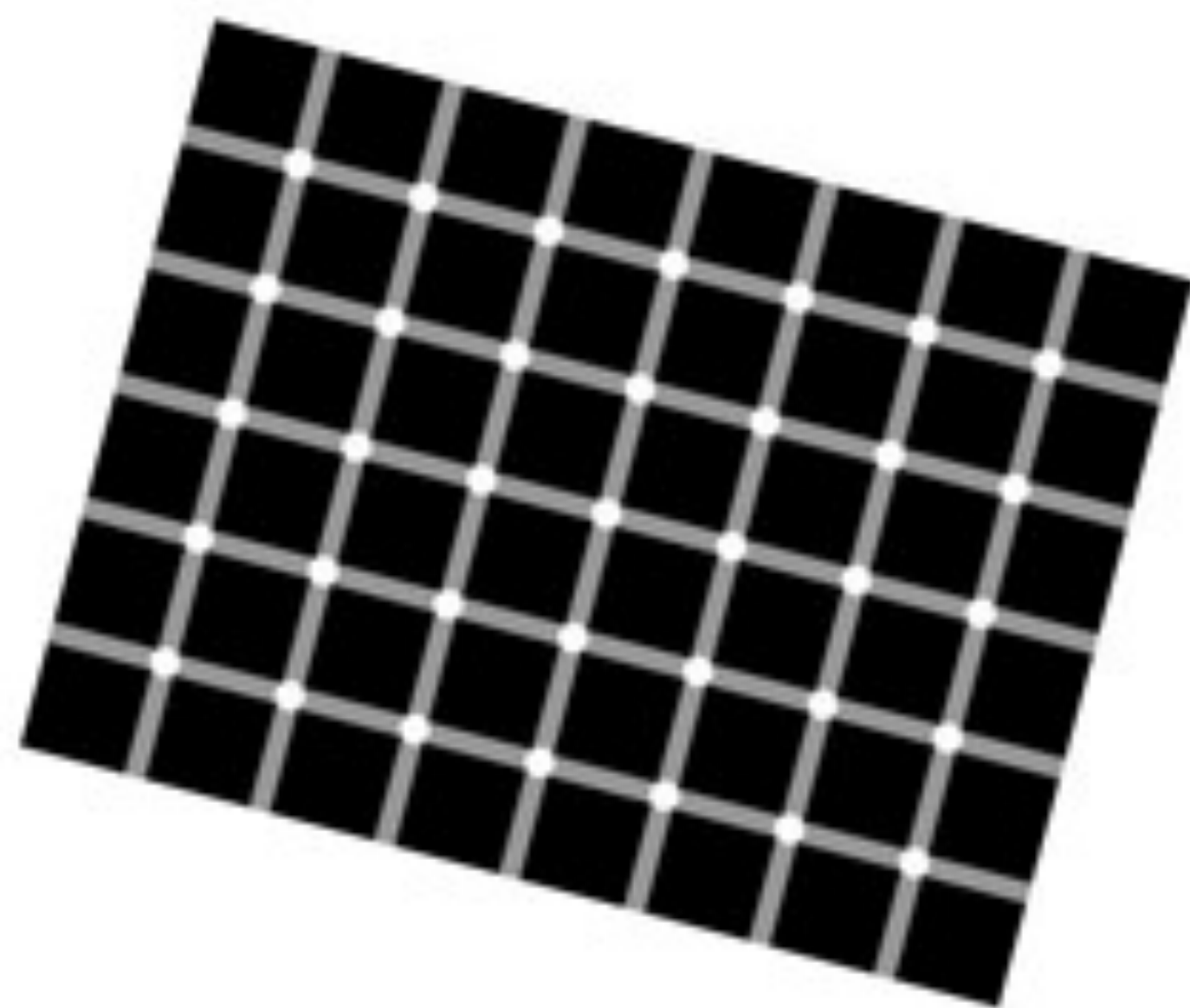
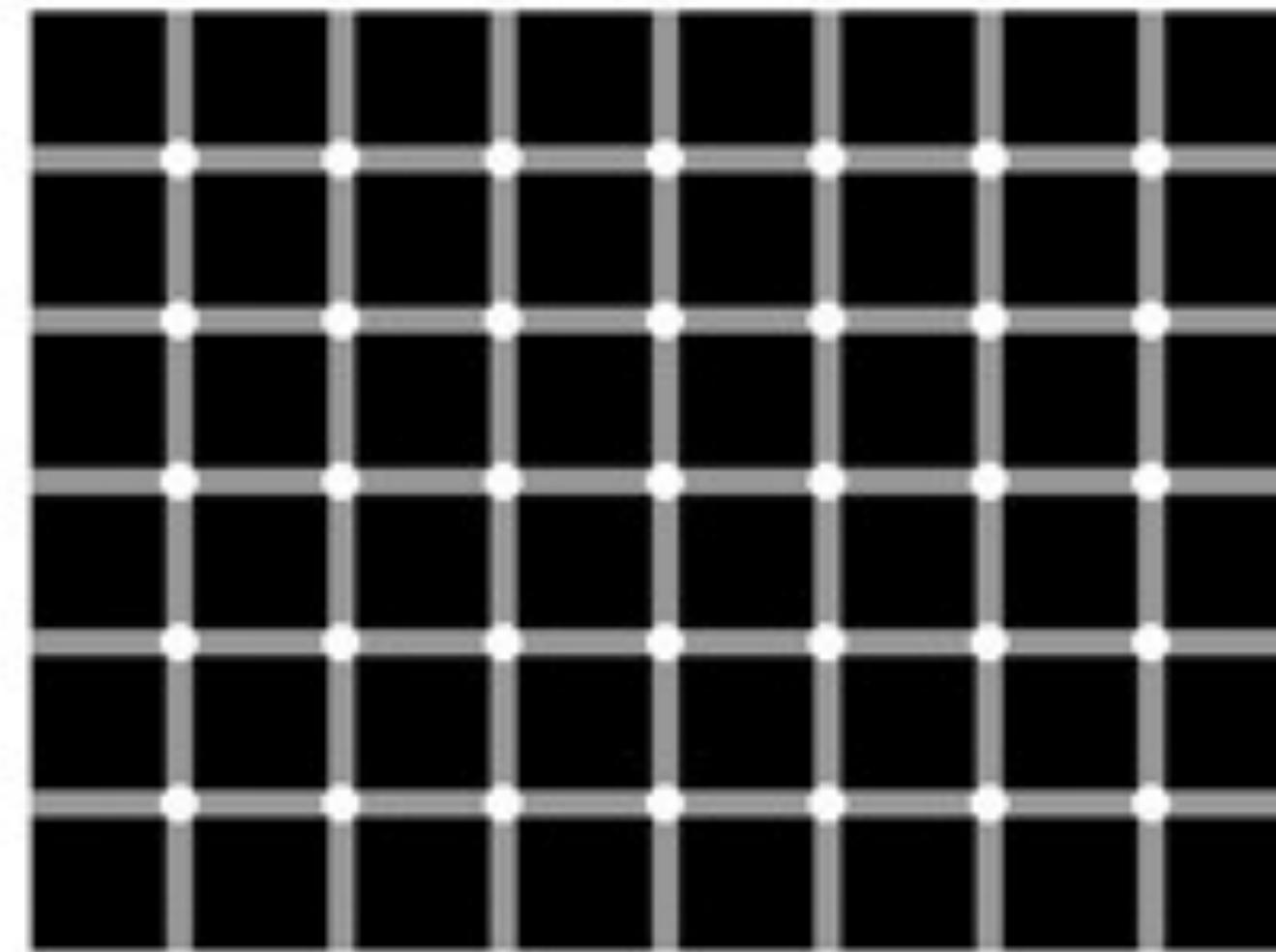
Two distinct approaches:

- linear combination of basis functions:
polynomials,
B-Splines,
Thin Plate Splines (TPS),
Radial Basis Functions (RBF)
 - modelling deformations as a physical process:
e.g., viscous fluids (or other models) that deform images
under the influence of external forces
- =>> estimates a 3D vector field describing the displacement or deformation at each voxel in the image

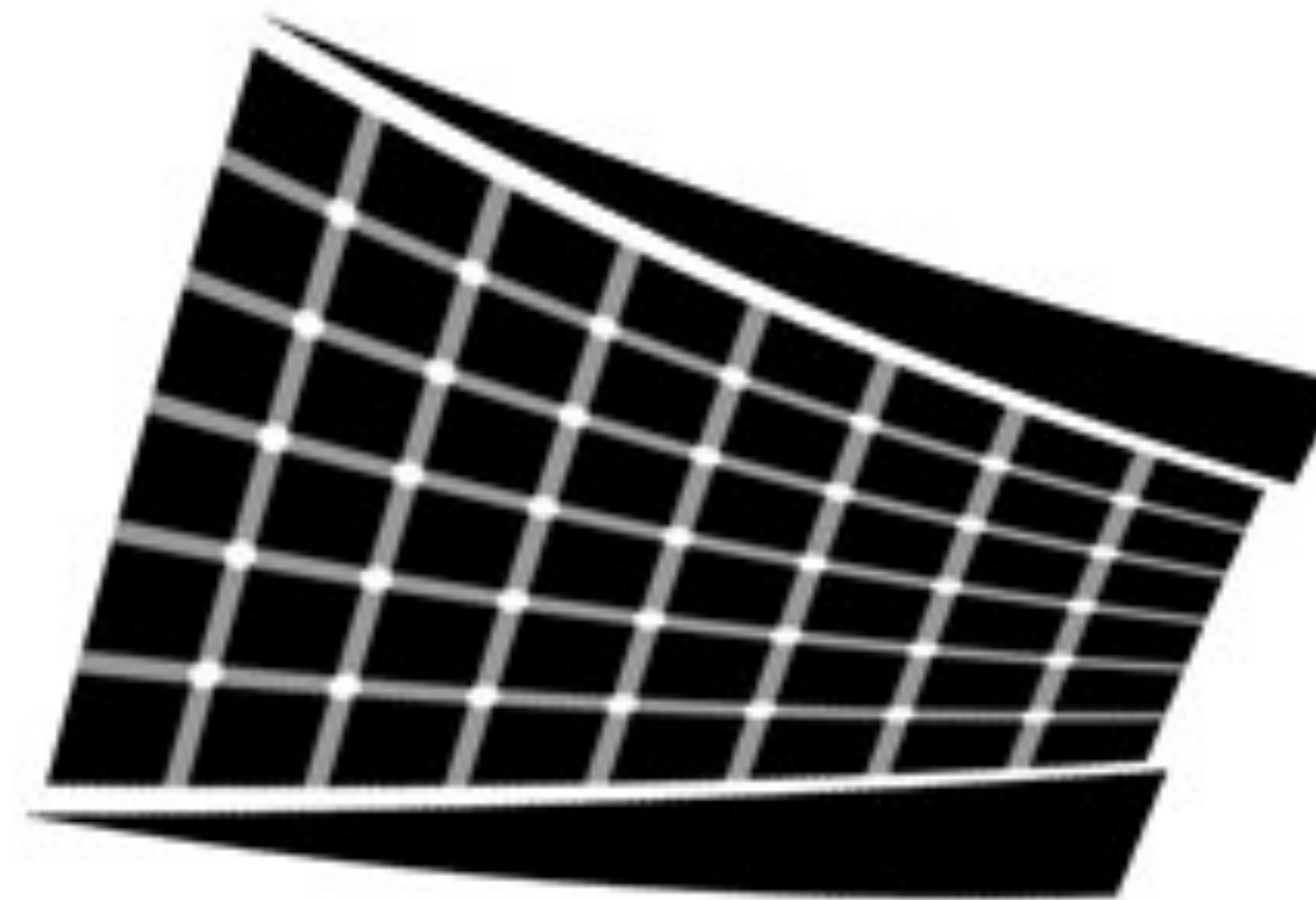


Non-rigid registration

Original



*Rotate and Shift
Rigid*



Affine

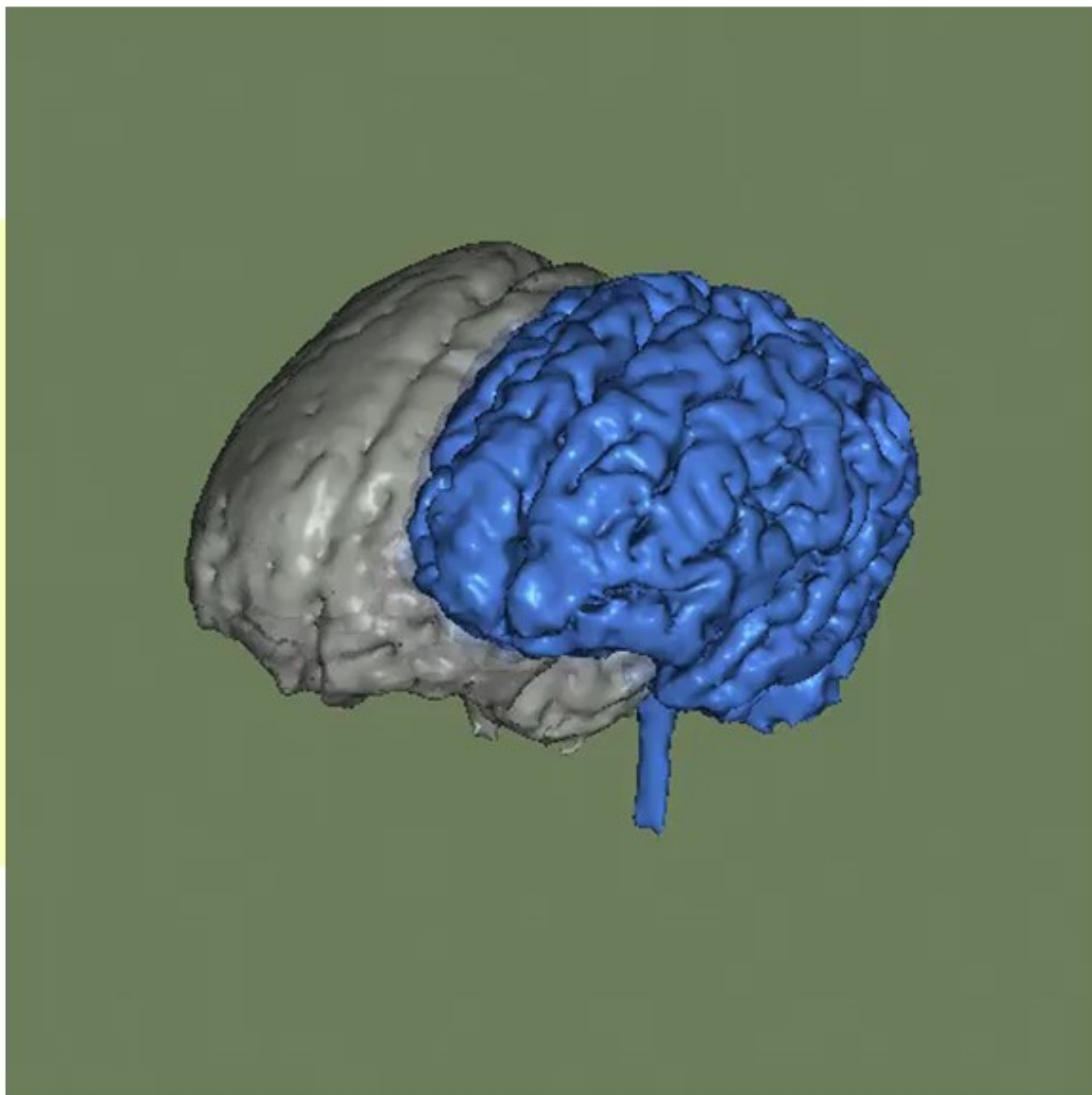
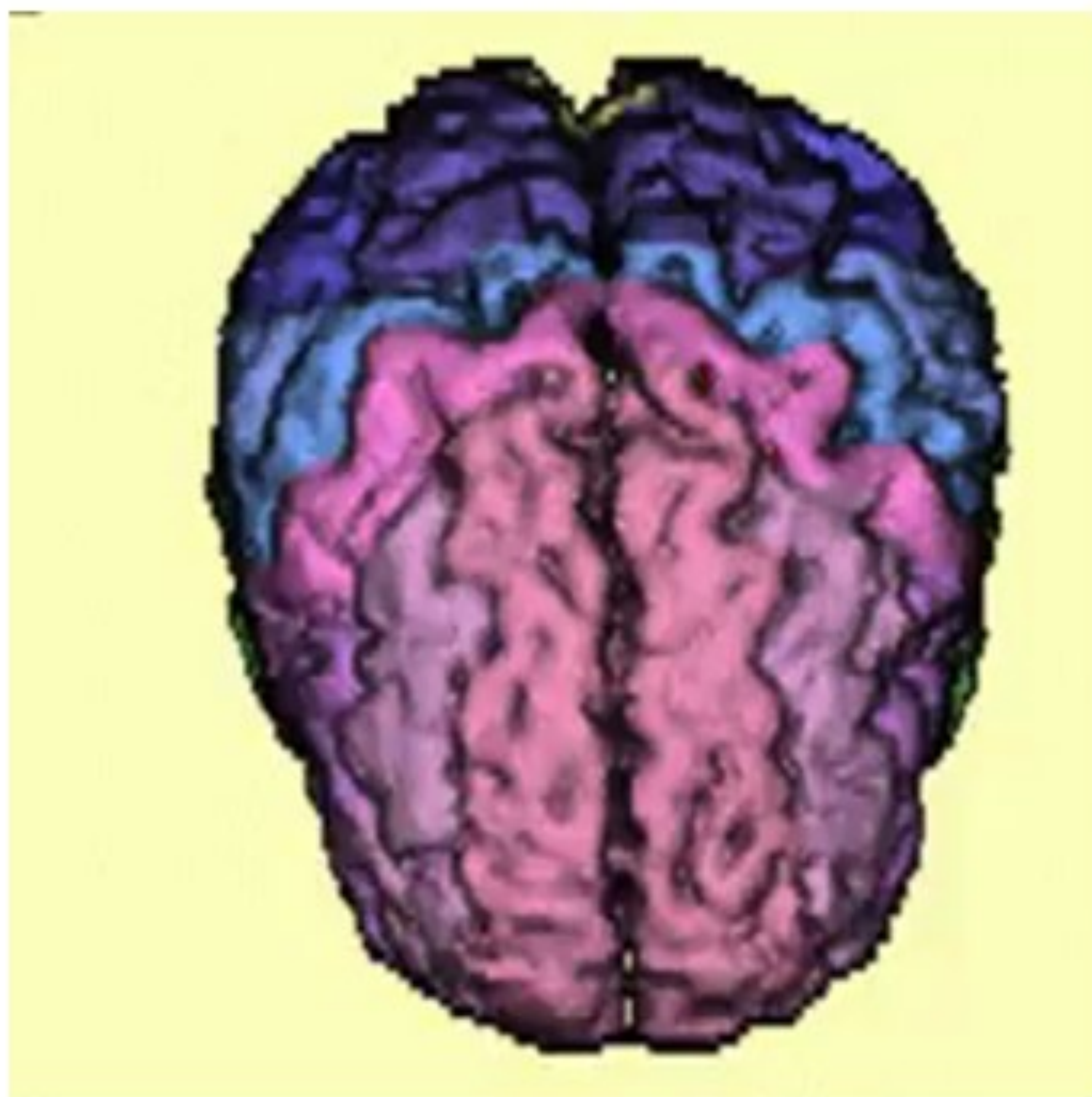


*Warp
Non Rigid*



Non-rigid iterative registration: brain(s)

(courtesy SPL)



Non-rigid registration: mouse

(mouse model/source data courtesy SKI;
post-proc LMB)

CTref

MRori

MRAff

MRnrC

MRnrF



sli



3D



Non-rigid registration

Need for:

- an interpolation/deformation model
 - Optic flow
 - Thin plate spline, etc.
- spatial constraints
- an optimization technique

=>> Much more computer intensive than rigid registration

+ validation?



Registration validation: rationale \Leftrightarrow how-to

To provide a measure of likely error for a given case:

- landmarks: Root Mean Square error (\approx distance)
- content: Least Square difference, correlation ratio

To provide an estimation of maximum/typical error for perfect and degraded cases:

- simulations:
 - numerical (total control),
 - realistic (closer to real life)
- phantoms, cadavers
- reference clinical data sets
- actual clinical data sets

*BUT think of the various/individual components
within the global registration system!!!*



Multimodality registration:

chain of process; most relevant items between parentheses

acquisition, transfer -> reference data set and source one

conversion (e.g., Dicom to other)

segmentation (various techniques; manual to semi-auto.)

estimate/initialize current geometric transformation (# of DOF: e.g., rigid vs non-rigid)

transform source and compare to ref. (geometric transformation: DOFs, interpolation, etc.)

compare source and ref. (cost definition, e.g., analog to a distance)

optimization (iterative techniques; multiscale, constraints, multicost, etc.
to avoid local extremas)

assess results (visual, stats, etc.)

visualize (various paradigms)

exploit ...



Multimodality alignment: you can never be too careful (or warned)

- any registration method is actually a mix of various components
- not one single approach will solve every problem
- potential shortcomings and pitfalls need to be understood and possibly addressed ahead of time
- all results need to be carefully reviewed



Multimodality Imaging: 2 complementary concepts

- **Registration**



$\langle = \rangle$ establishes the spatial relationship between two (or more) data sets

- **Fusion**

$\langle = \rangle$ combines data from different sources, times, procedures for extracting and exploiting complementary information



Visualization techniques:

slices: Multi-Planar Reconstruction (MPR)

side by side images with coupled cursors
colorwash/fusion

3D surfaces:

segmented structures
textured by another modality
or other (parametric) data

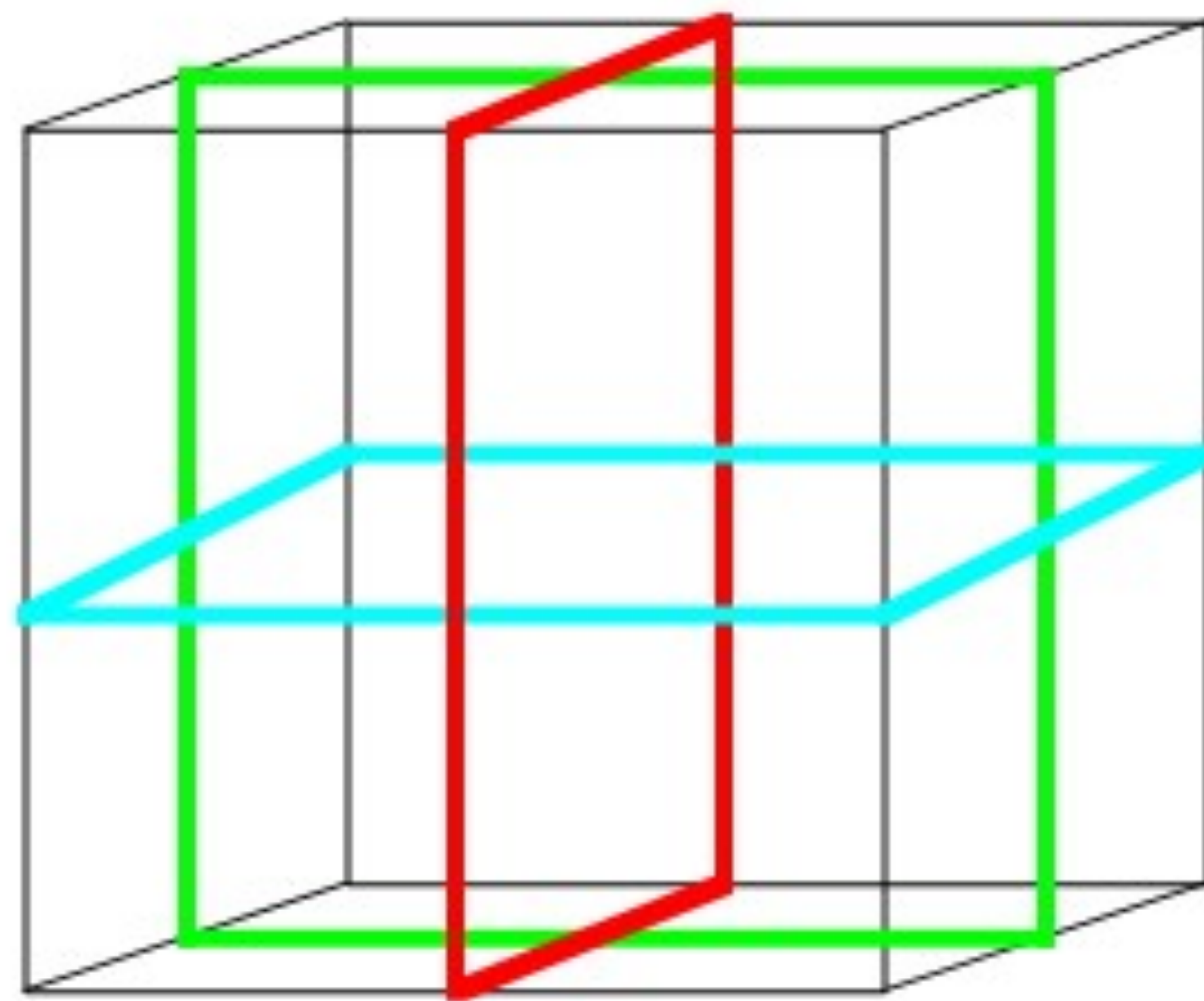
Volume Rendering:

Maximum Intensity Projection
ray-tracing/casting

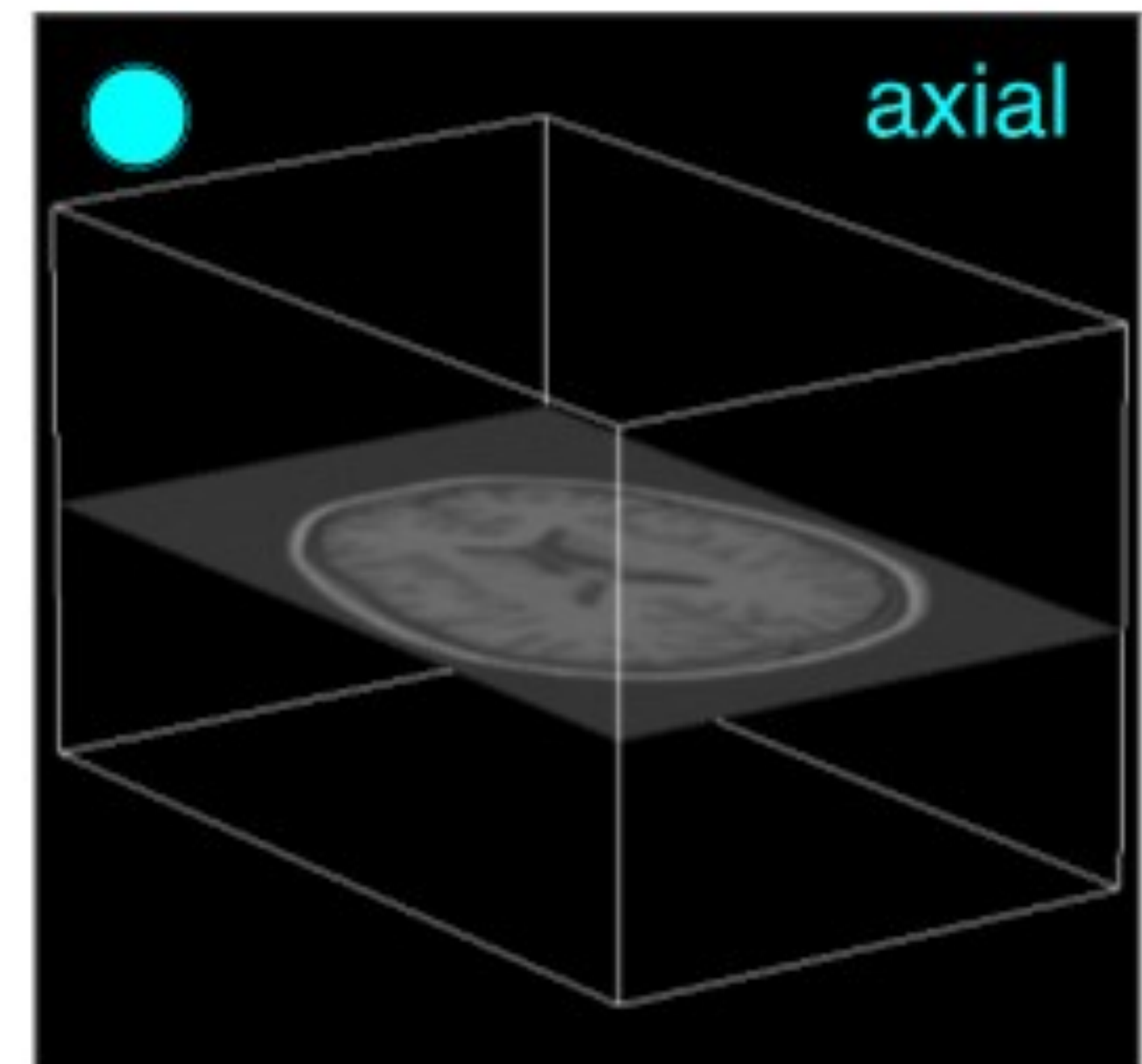
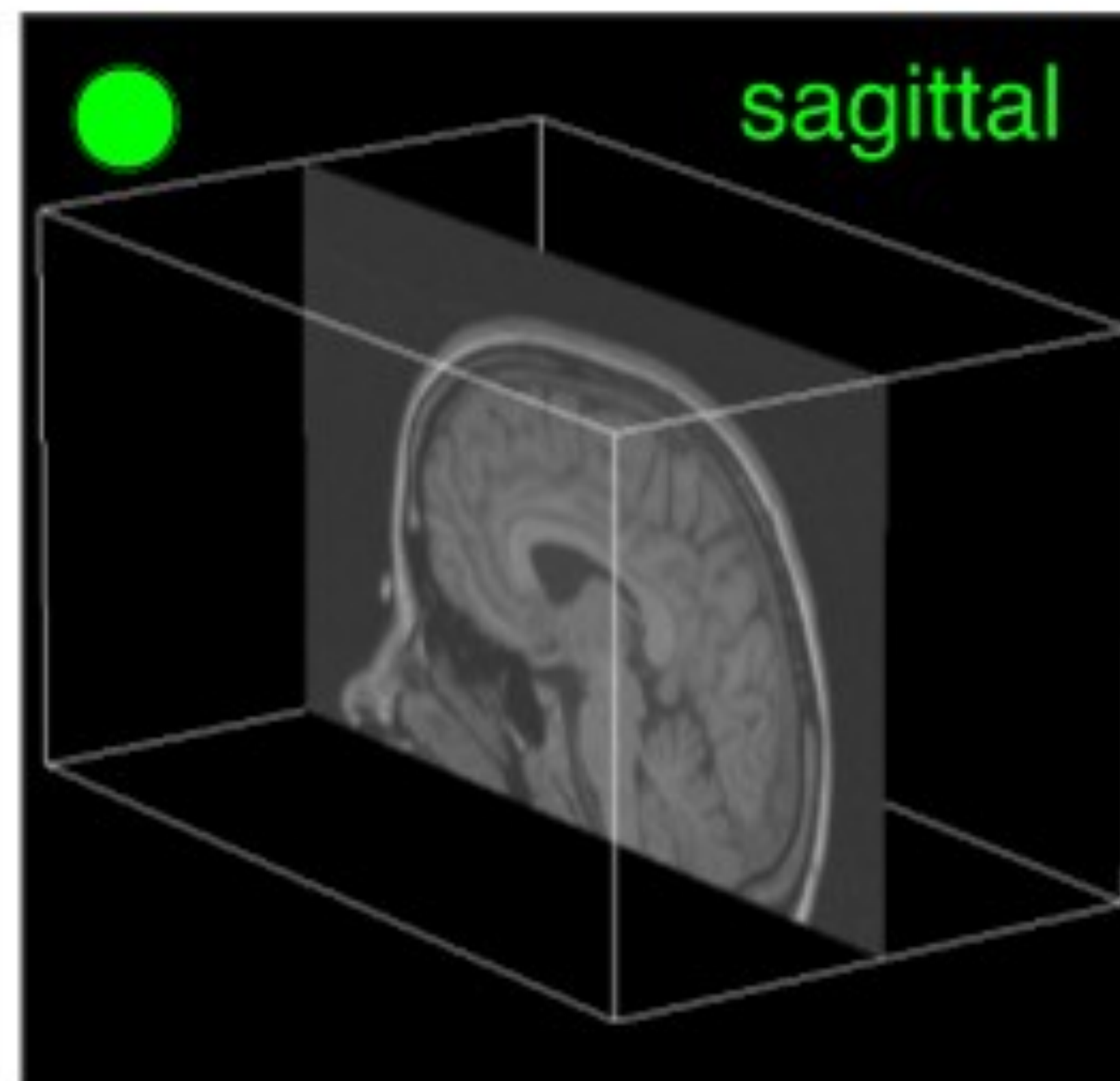
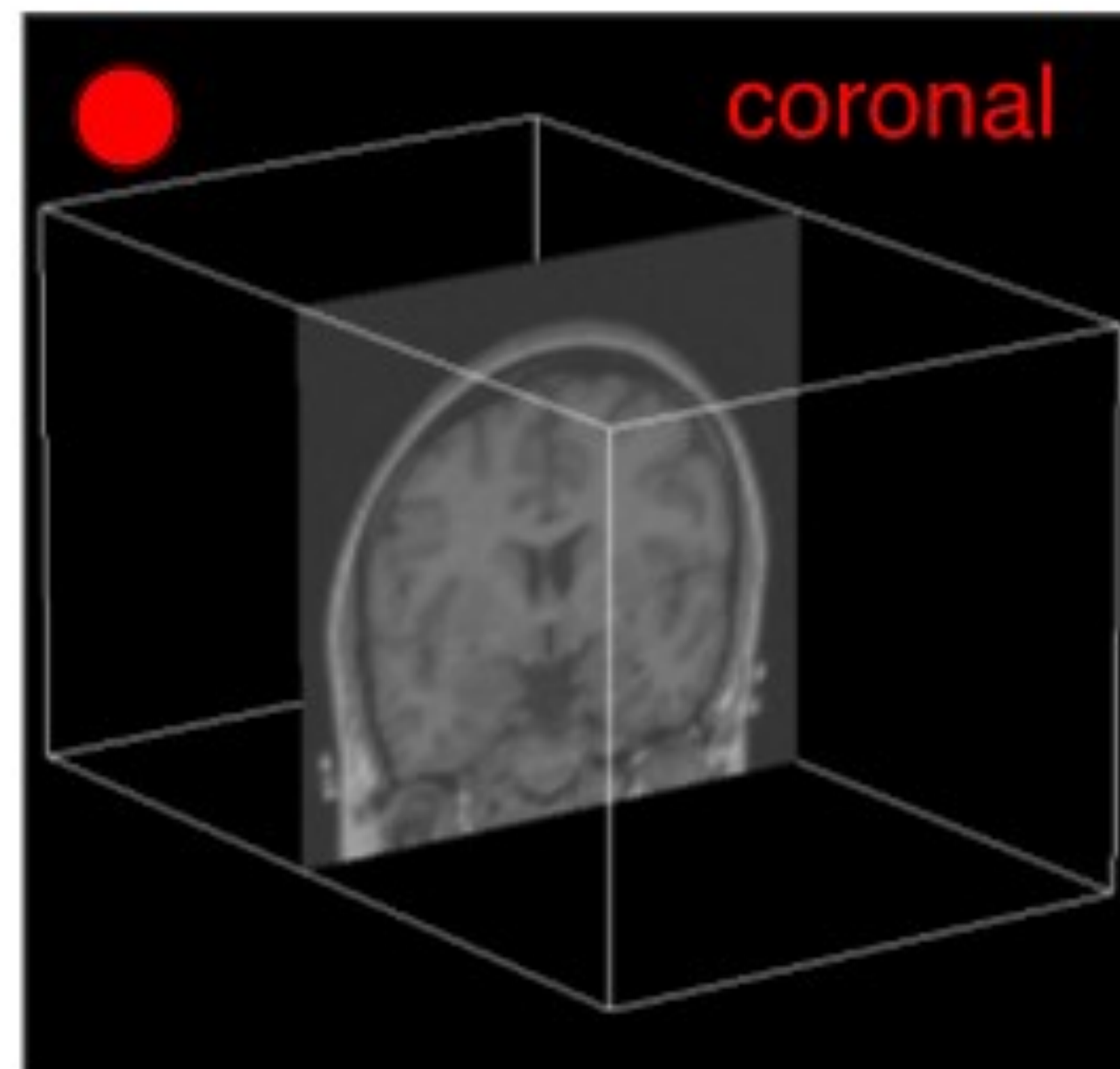
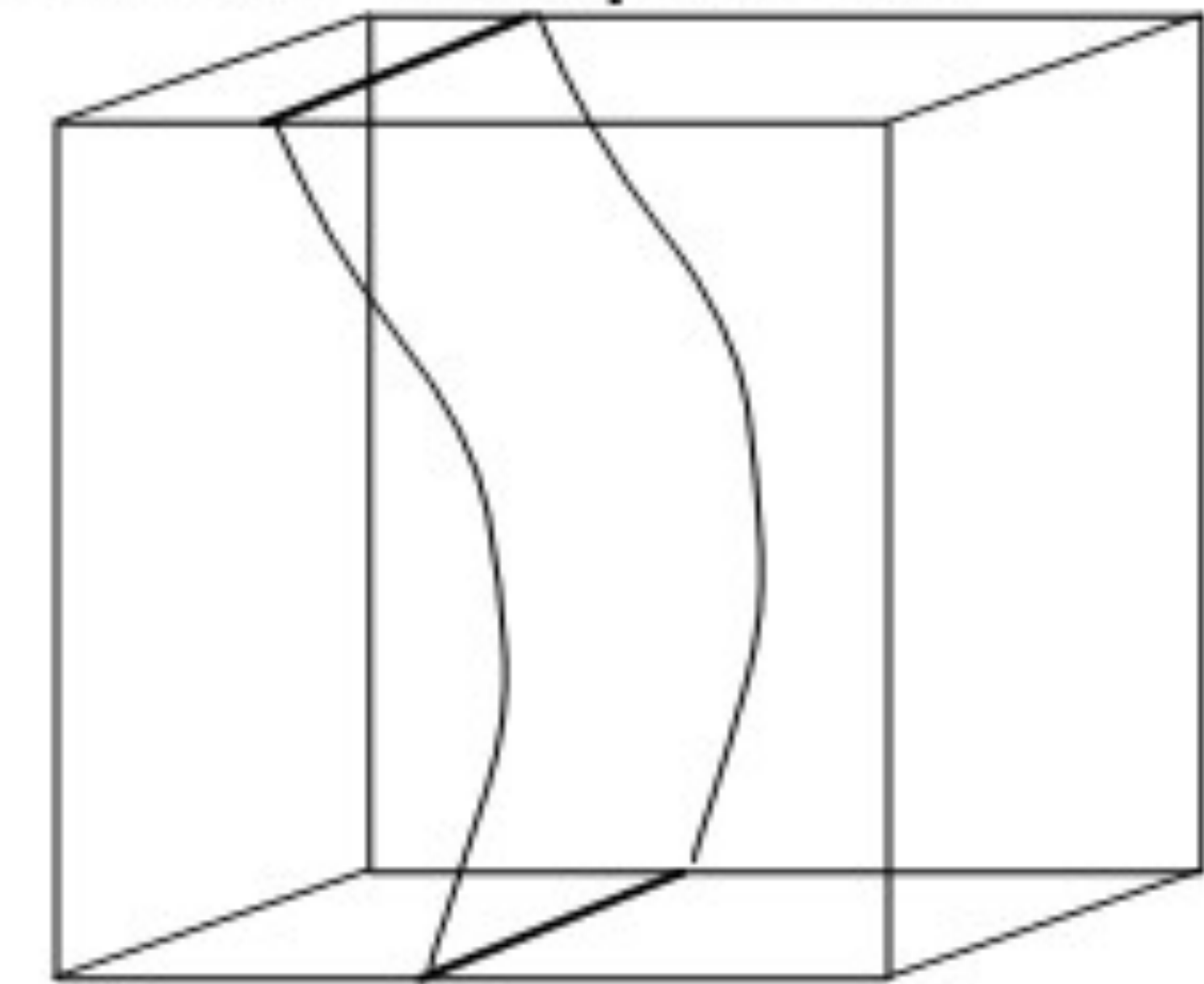
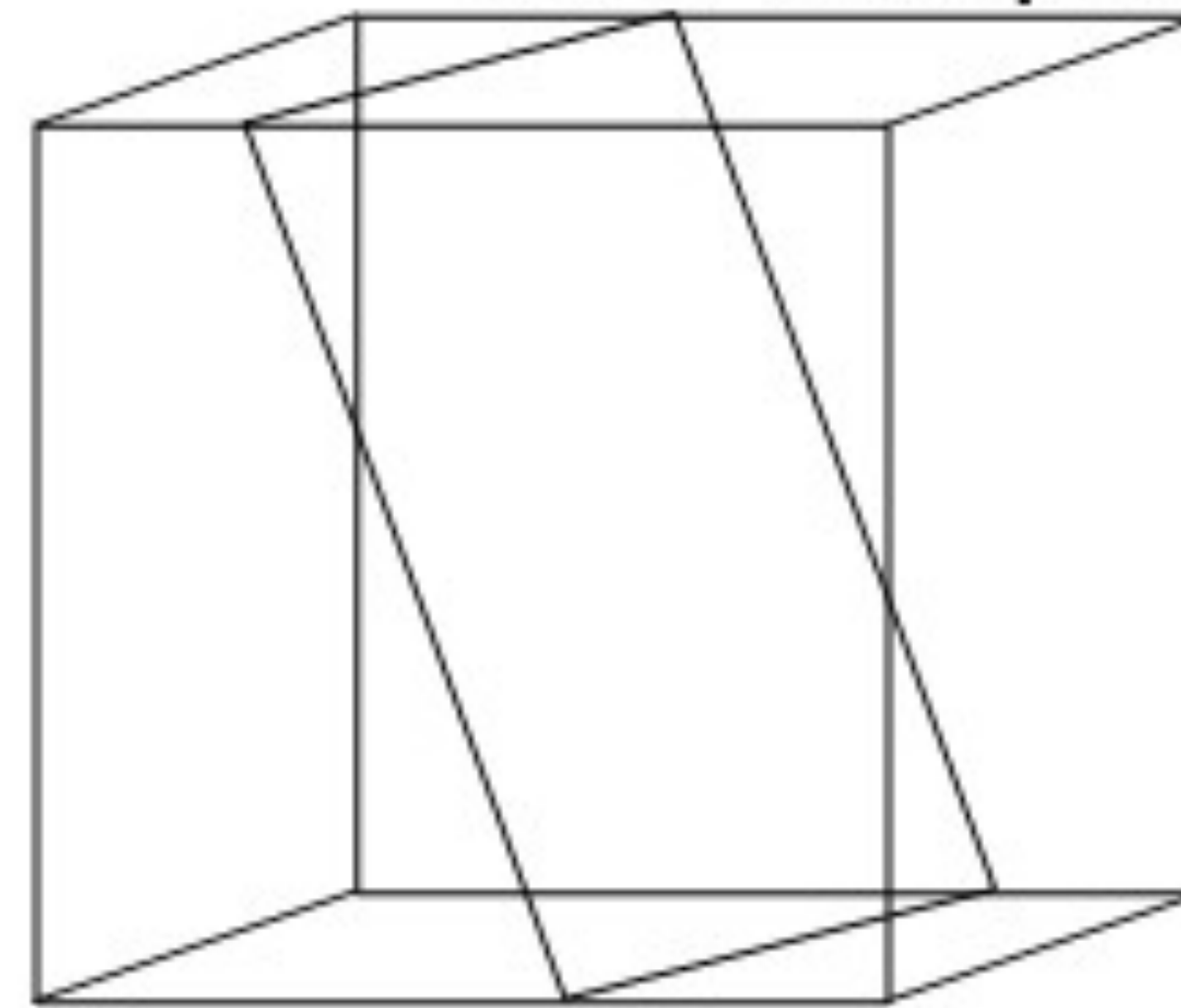


Multidimensional visualization techniques: slices and MPR

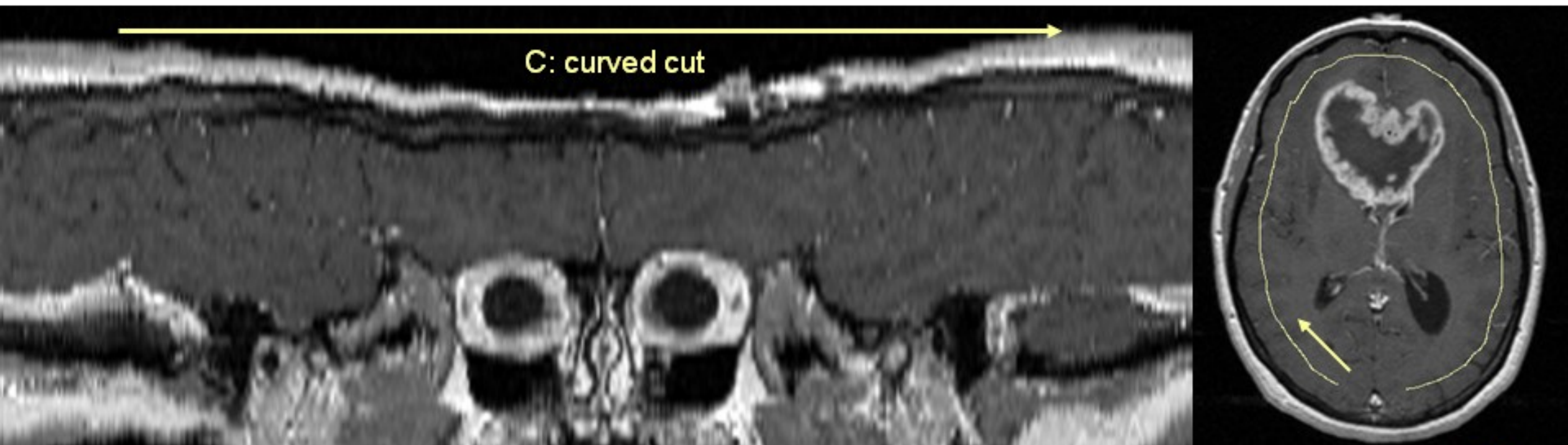
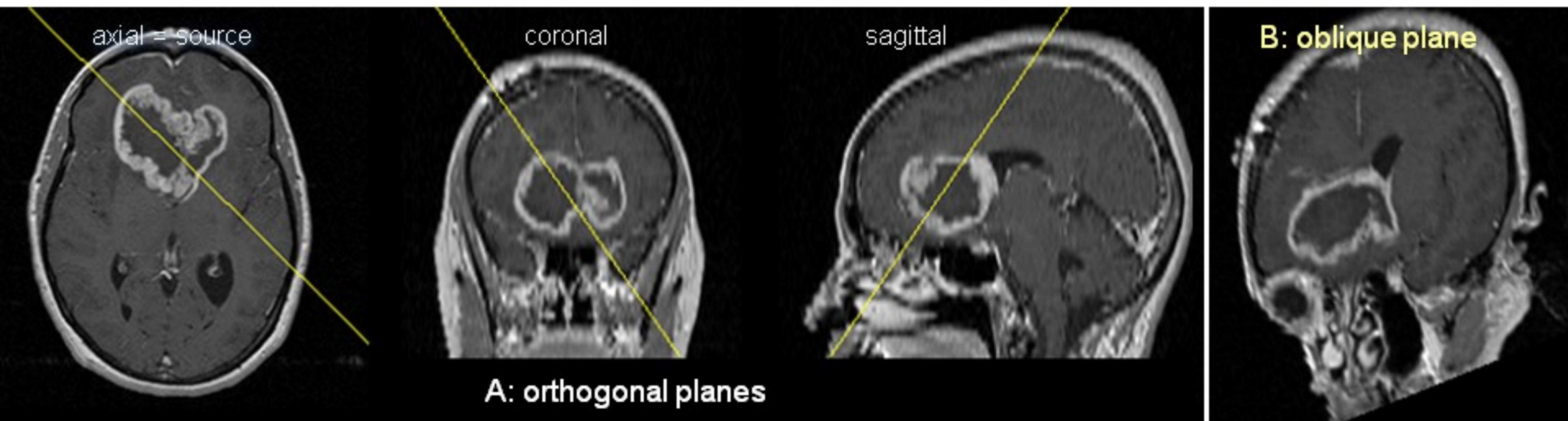
Multi-Planar Reconstruction (MPR)



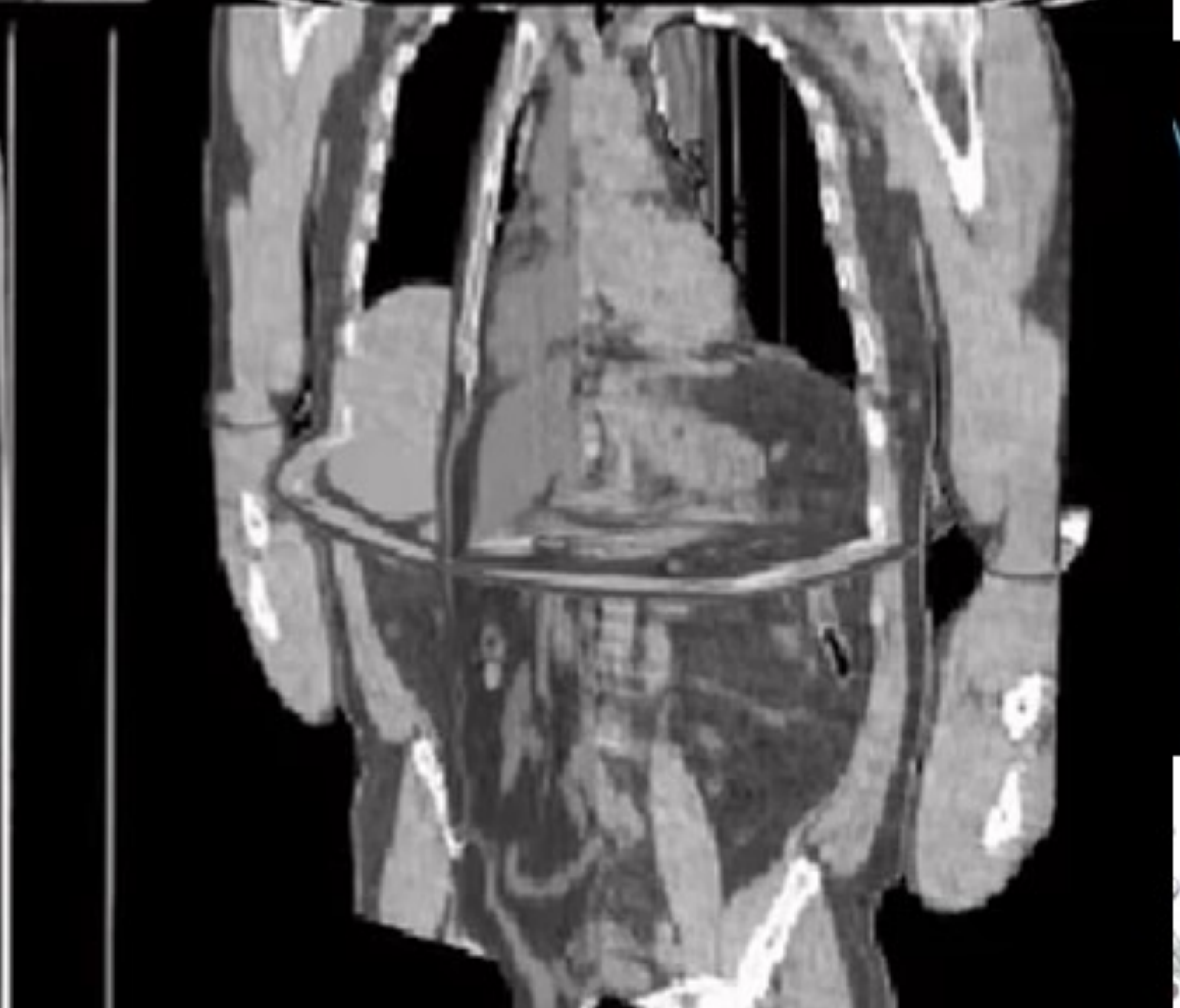
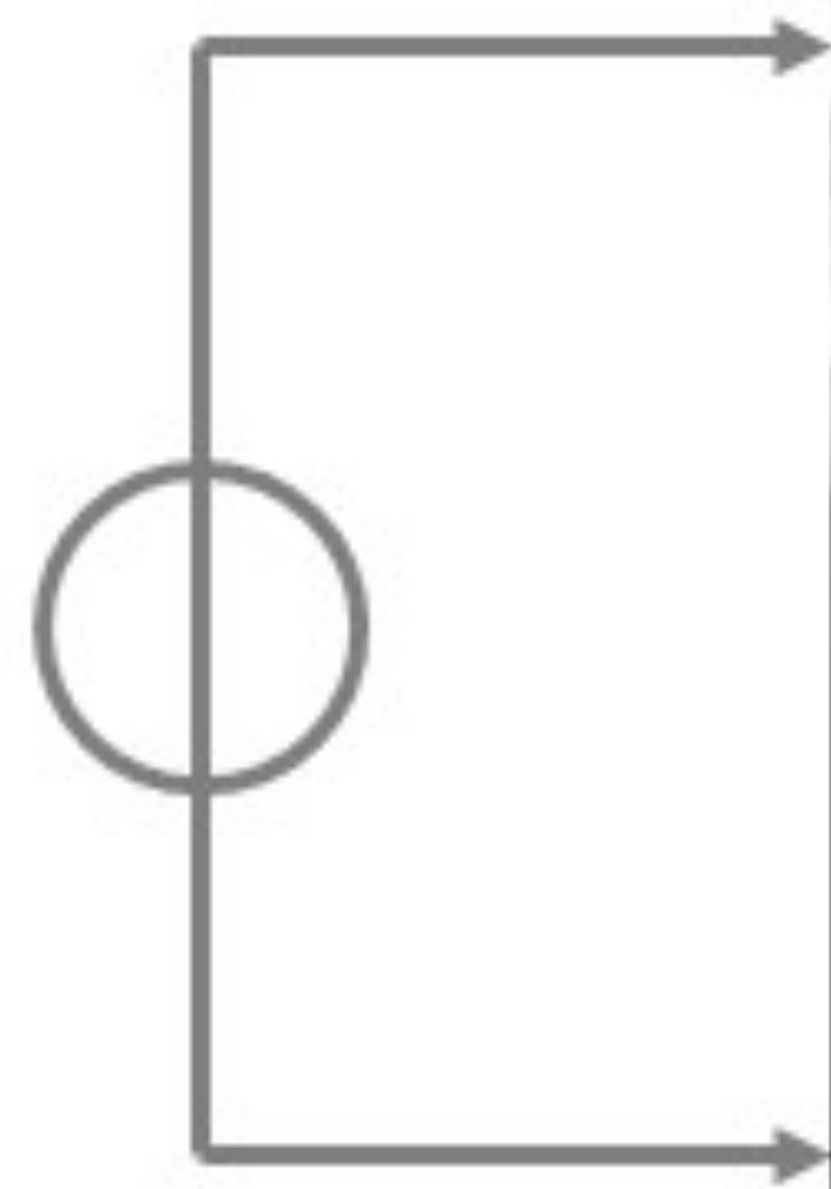
also oblique and curved planes



Multidimensional visualization techniques: slices and MPR



Multimodality
MPR:
colorwash



rot

Visualization techniques:

slices: Multi-Planar Reconstruction (MPR)
side by side images with coupled cursors
colorwash/fusion

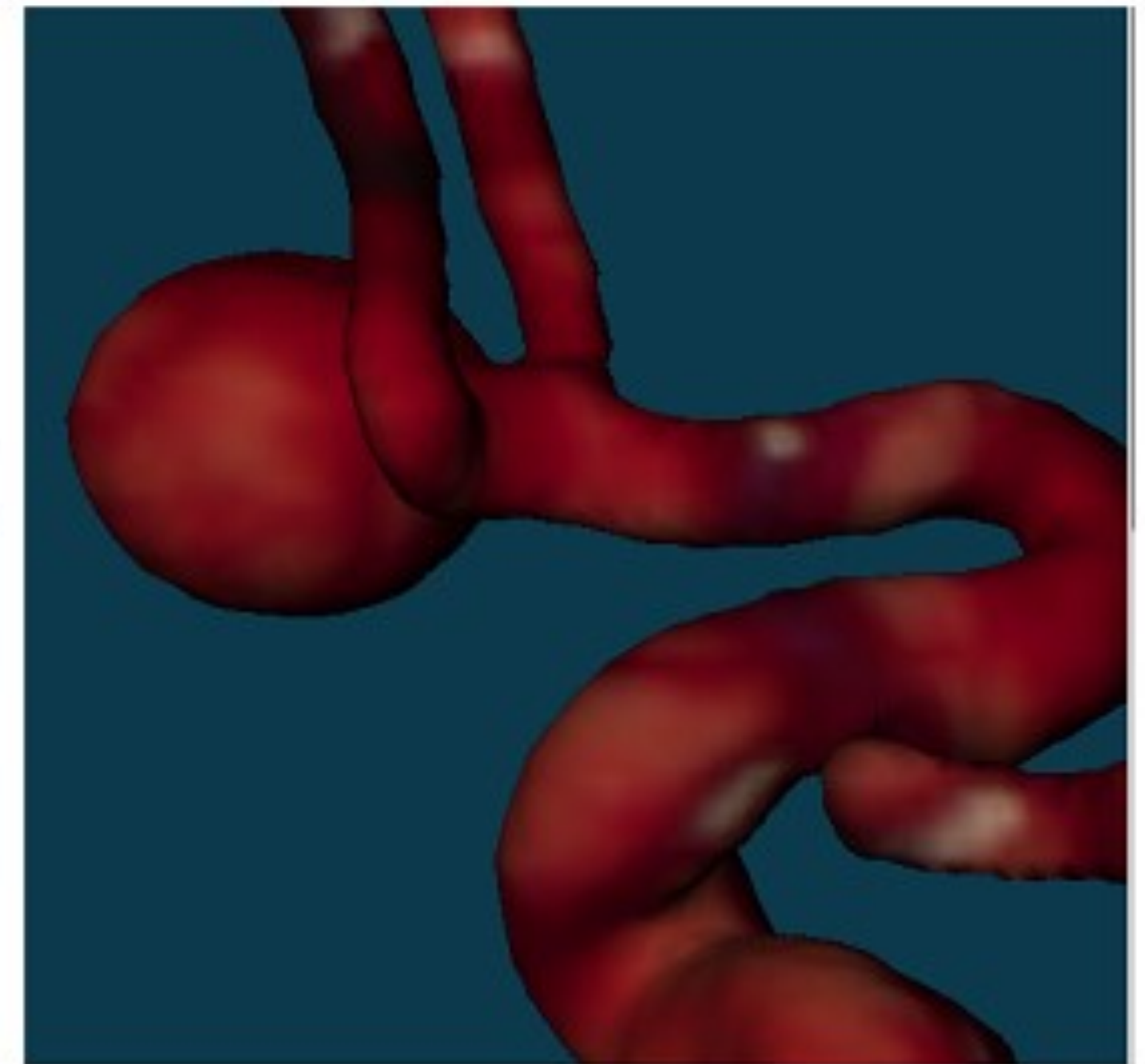
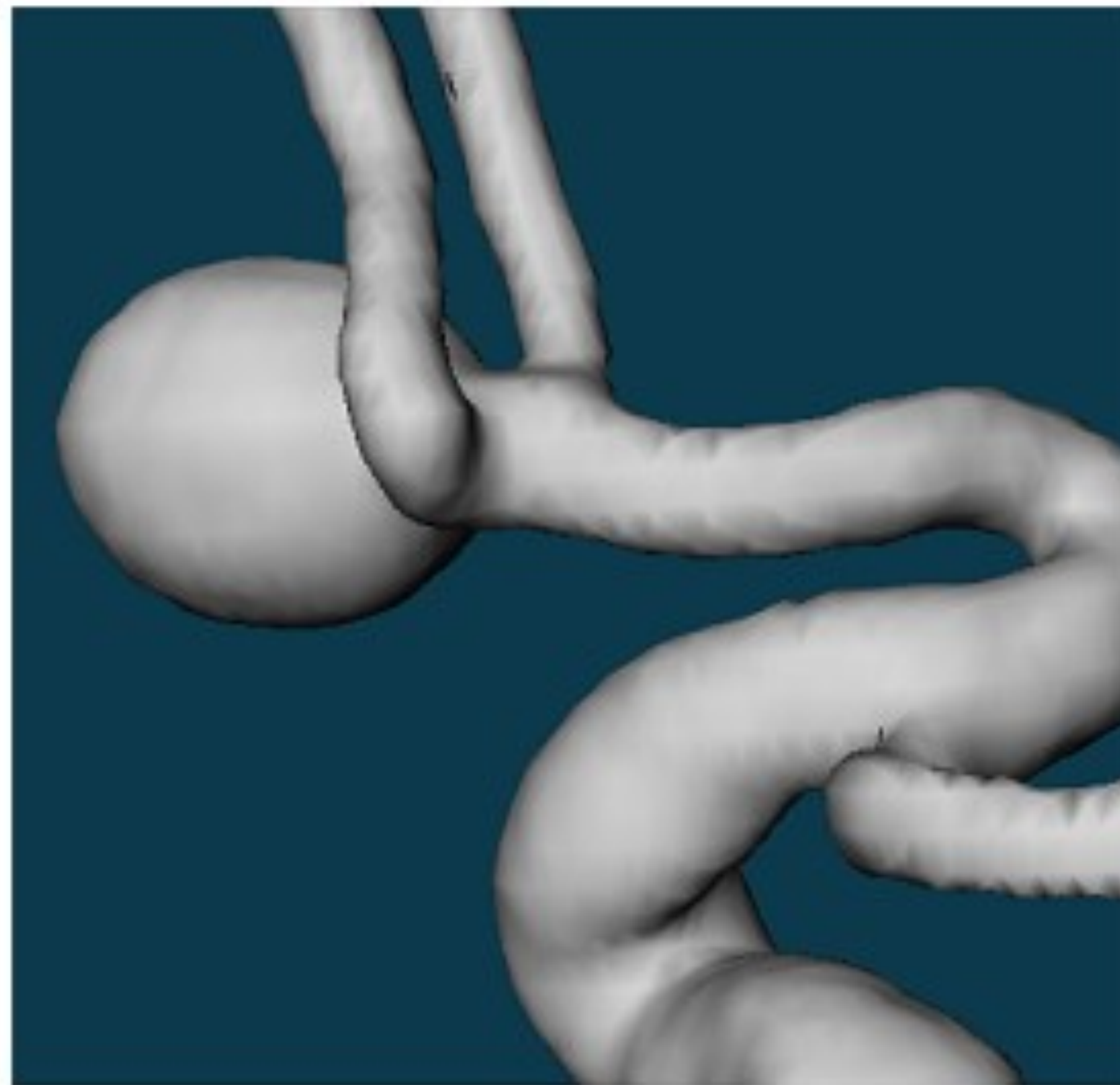
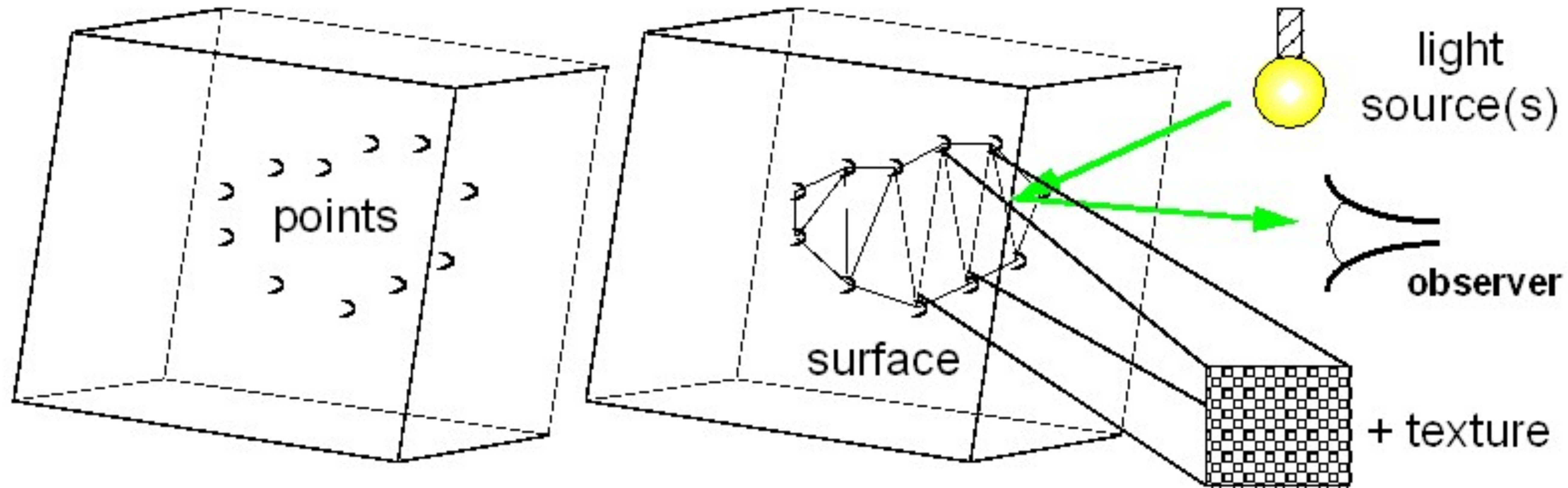
3D surfaces:
segmented structures
textured by another modality
or other (parametric) data

Volume Rendering:
Maximum Intensity Projection
ray-tracing/casting



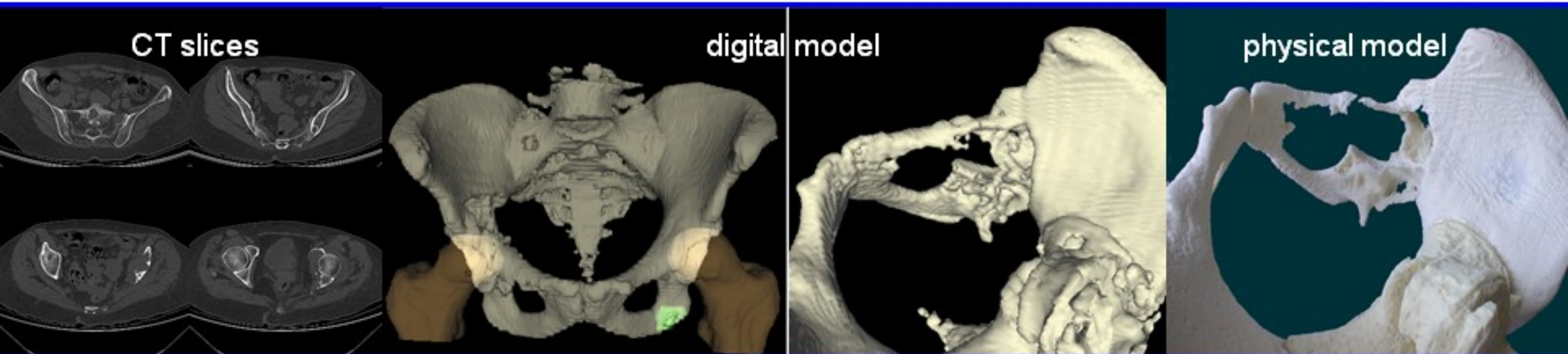
Multidimensional visualization techniques: surface rendering

Volume segmentation to surface

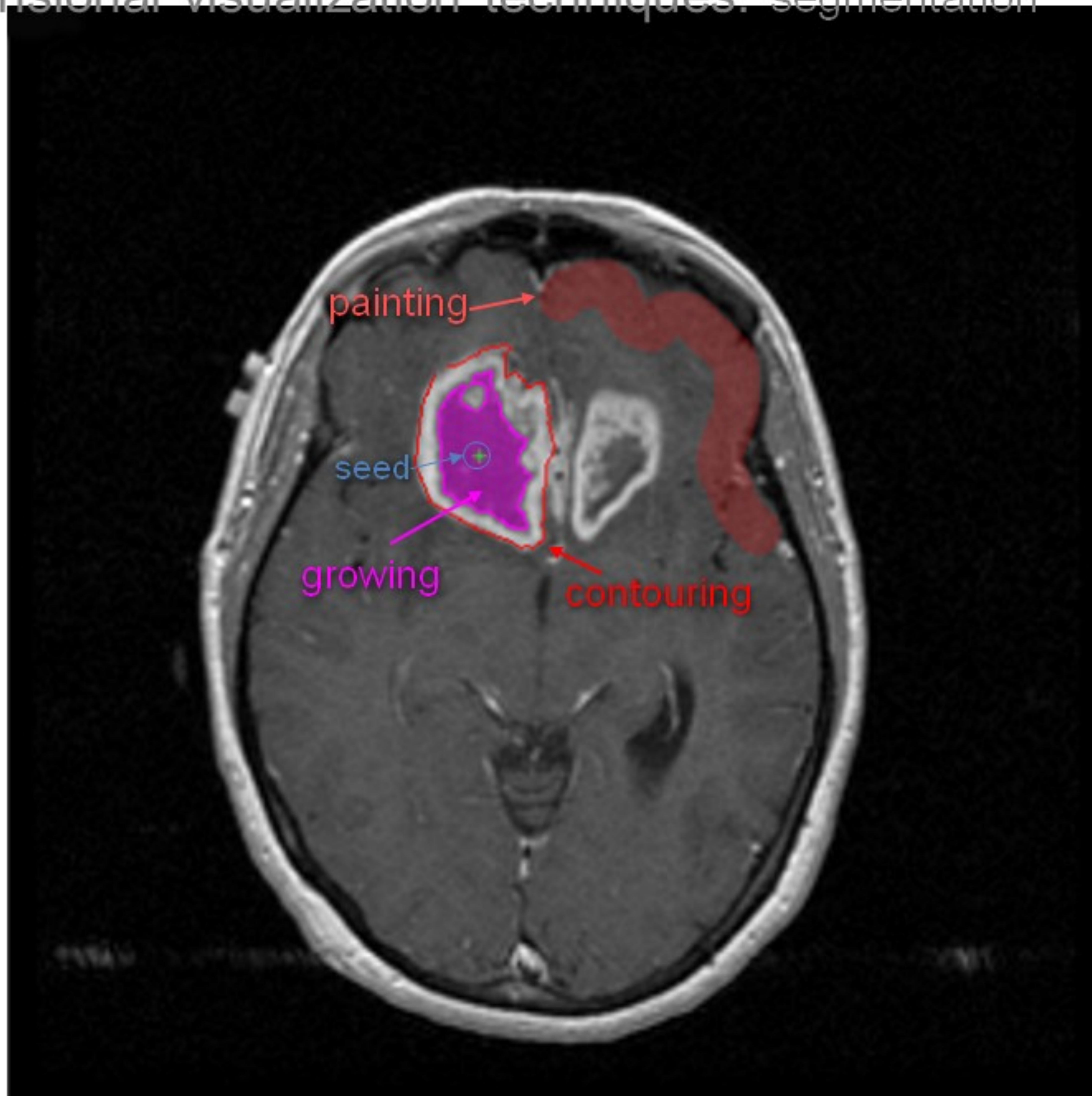


Rapid prototyping

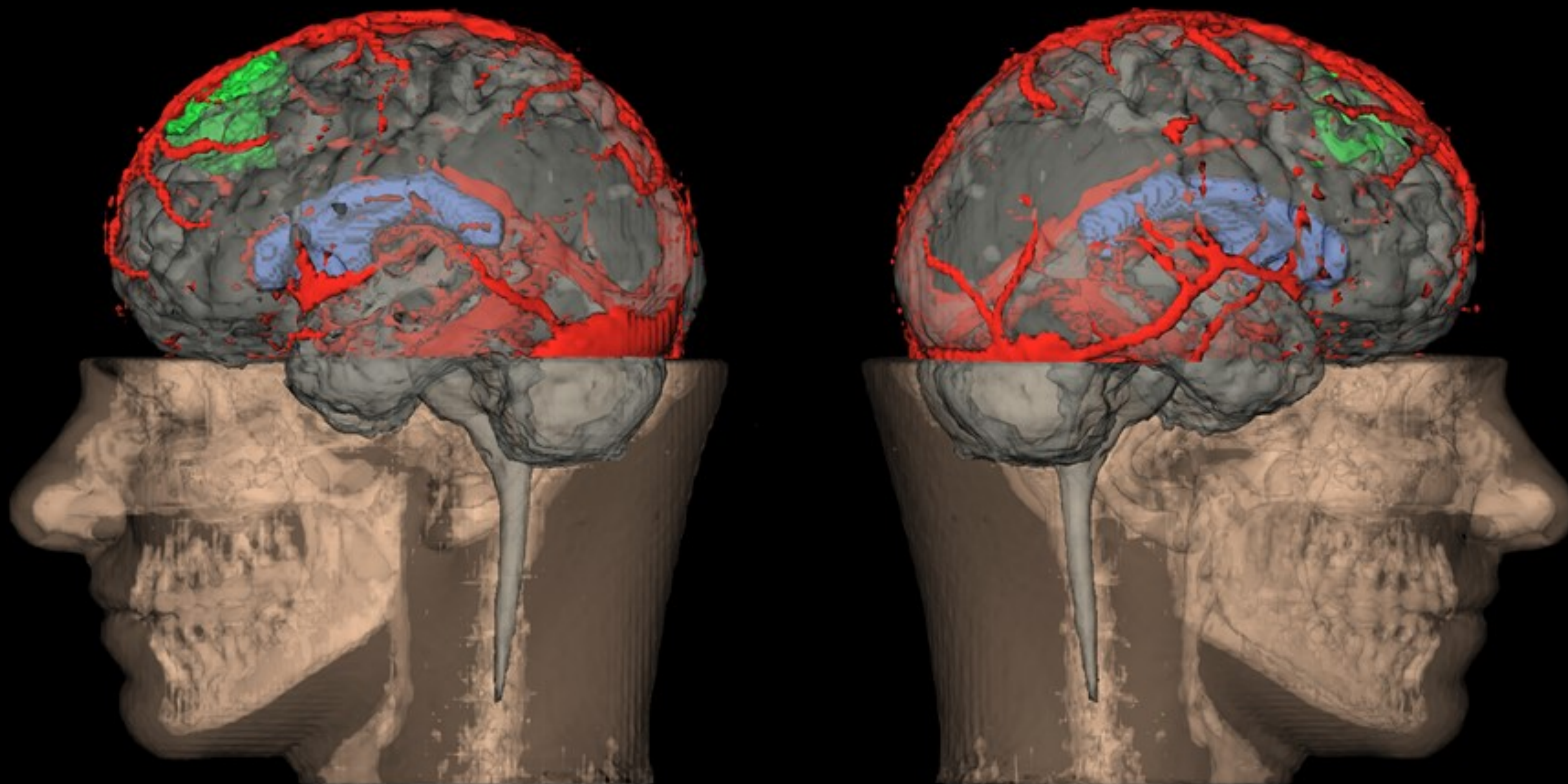
(source CT's courtesy HUG;
post-proc./3D printing LMB)



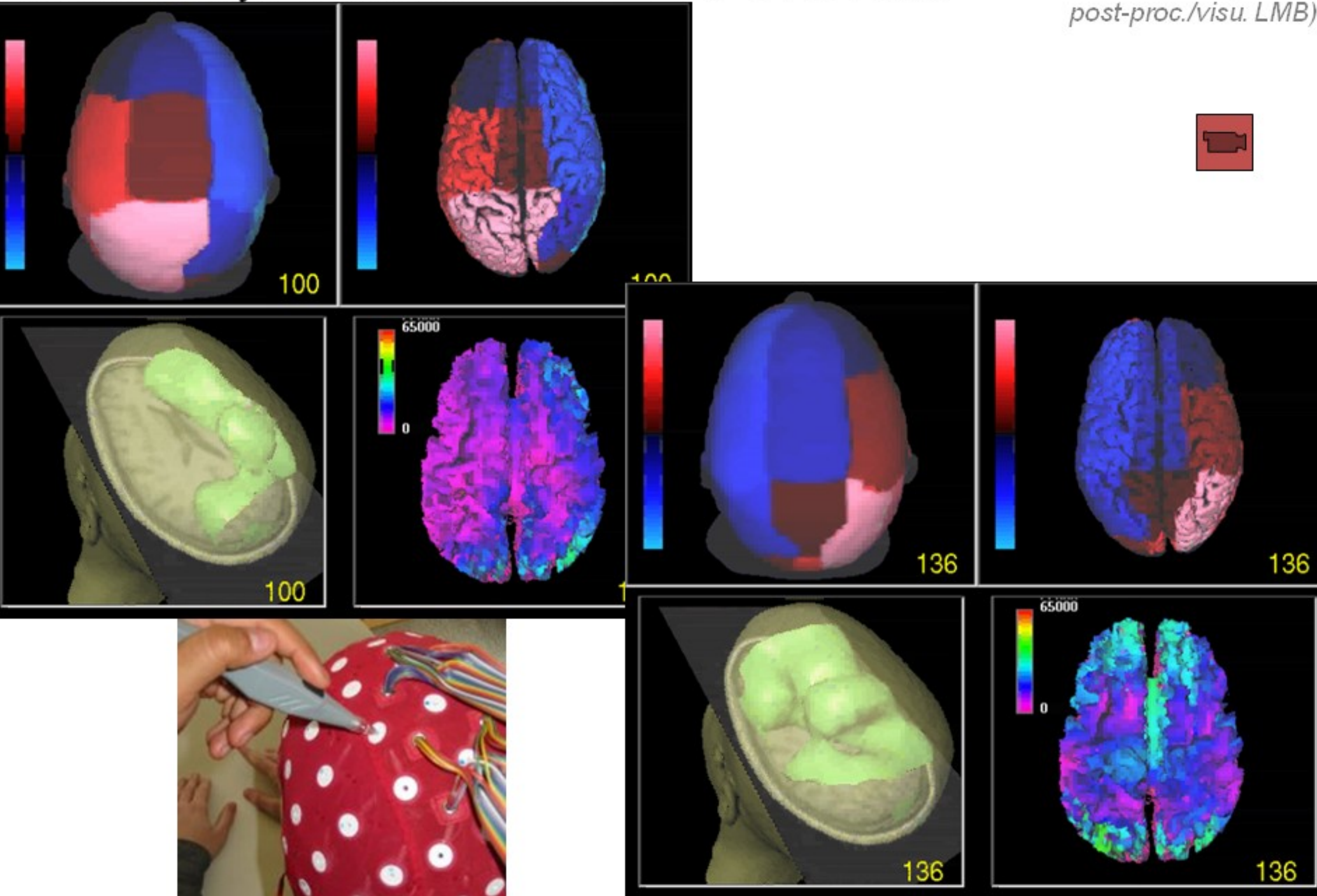
Multidimensional visualization techniques: segmentation



Brain tumor: *(source MR courtesy HUG; post-proc./visu. LMB)*



Multimodality/Multisensor: MR + ERP (EEG+EMT) (source MR/EMT courtesy HUG; post-proc./visu. LMB)



Visualization techniques:

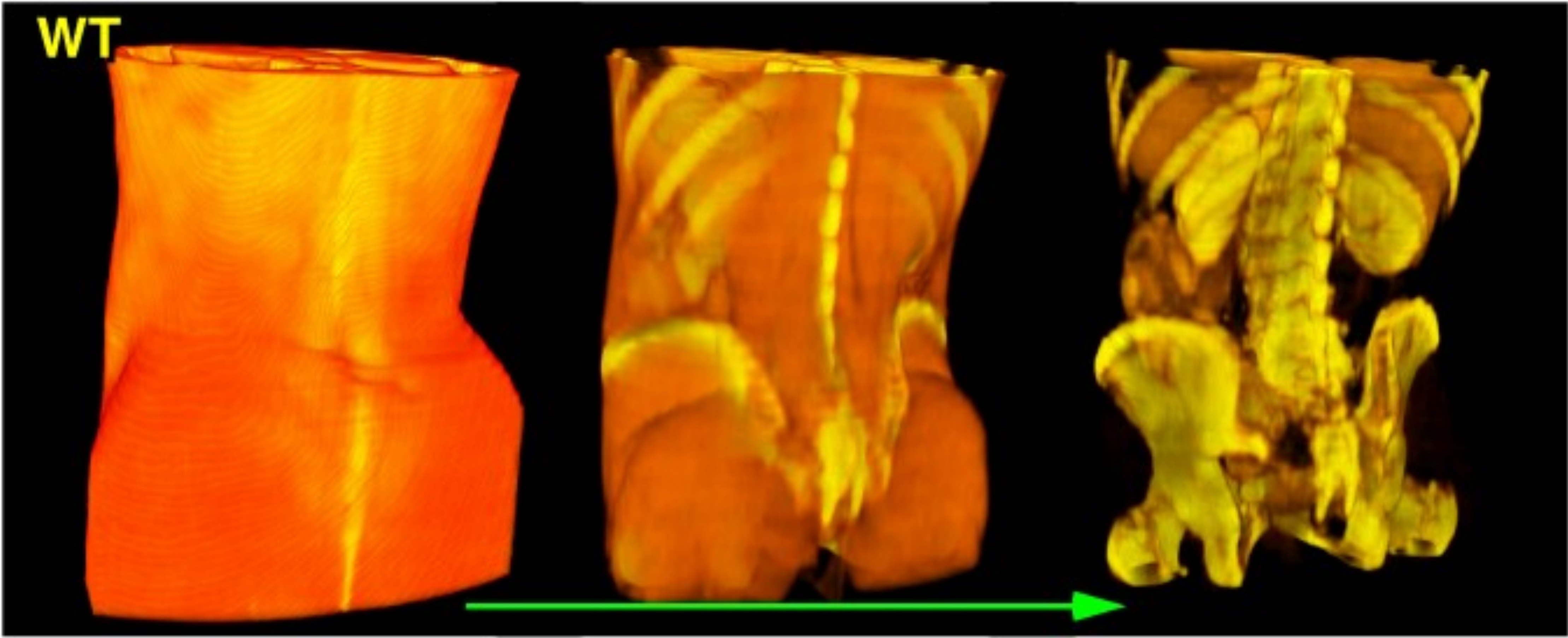
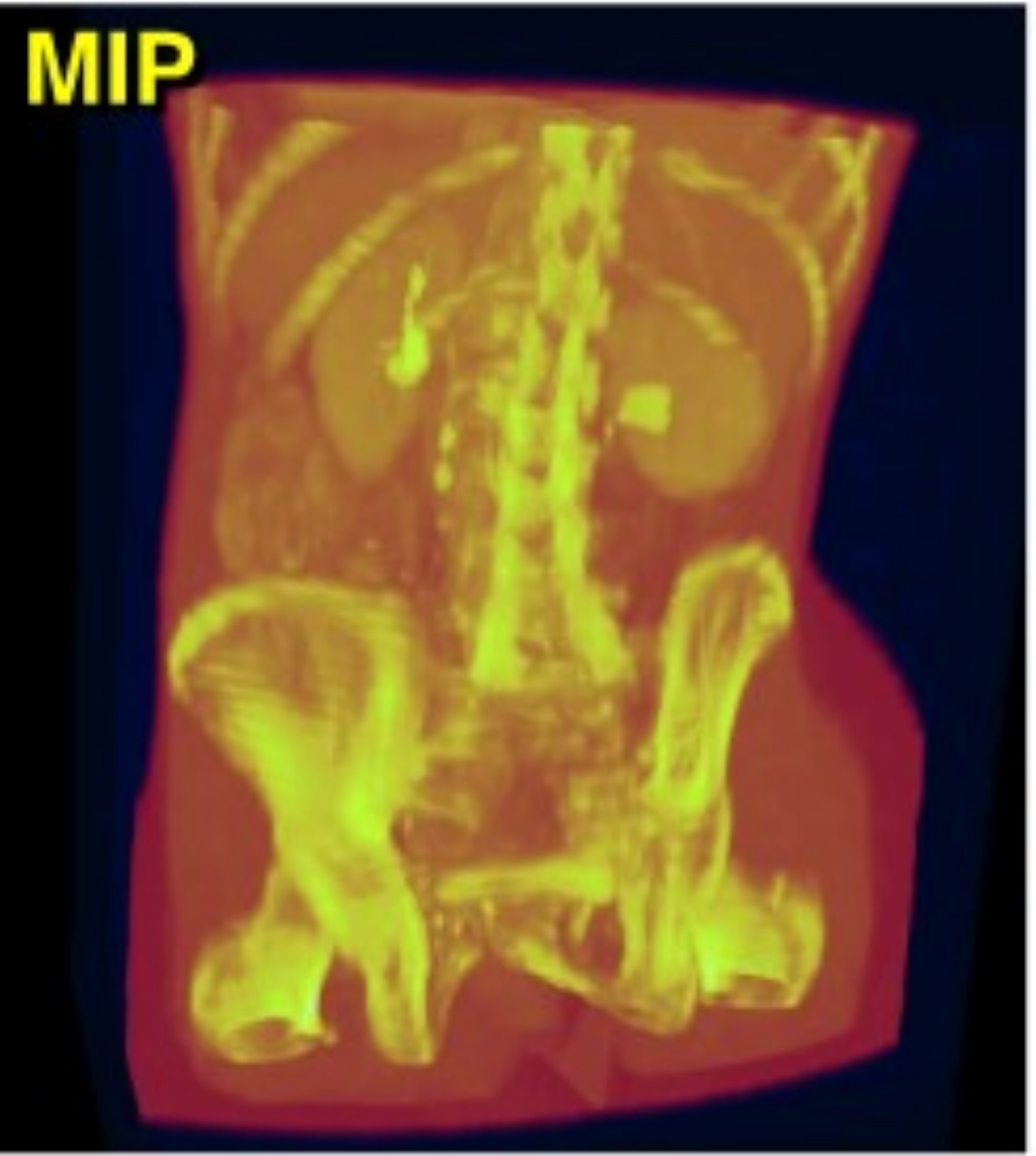
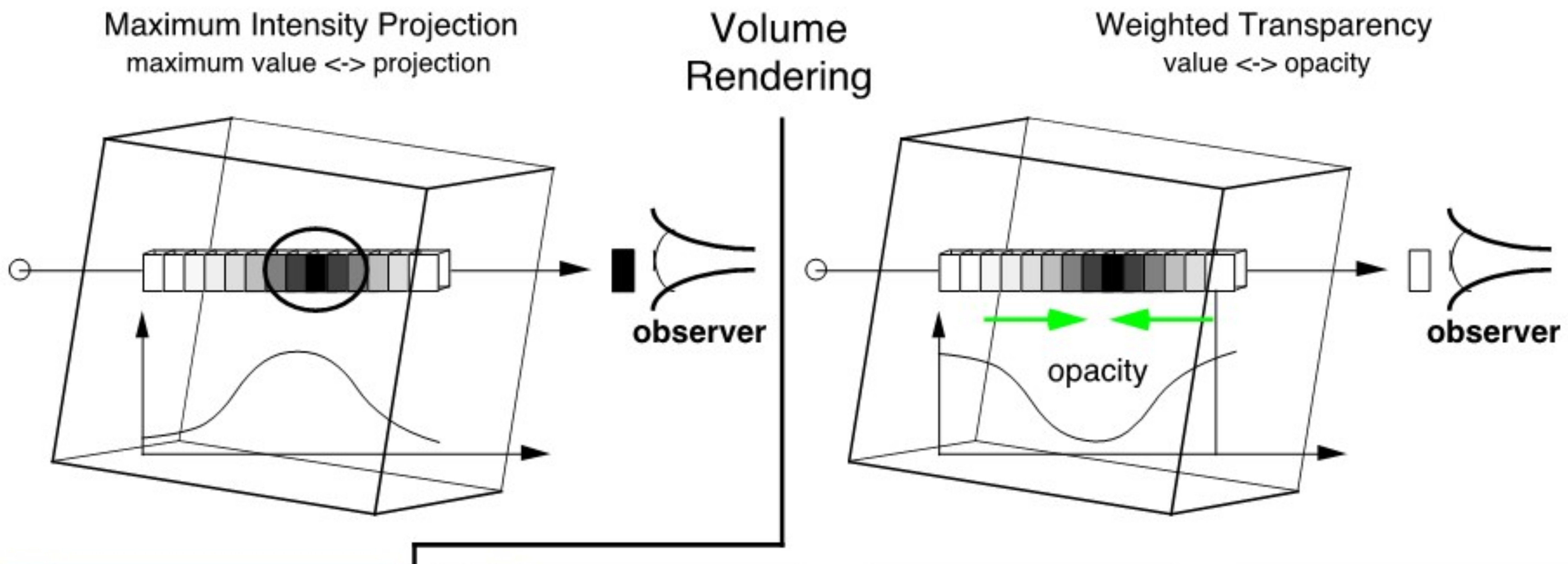
slices: Multi-Planar Reconstruction (MPR)
side by side images with coupled cursors
colorwash/fusion

3D surfaces:
segmented structures
textured by another modality
or other (parametric) data

Volume Rendering:
Maximum Intensity Projection
ray-tracing/casting

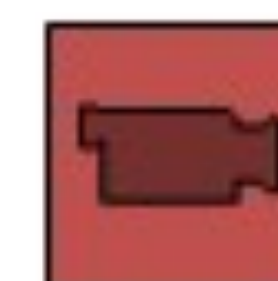
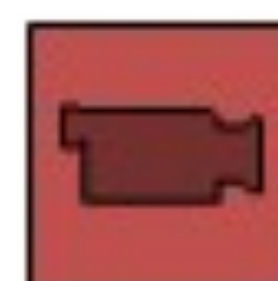
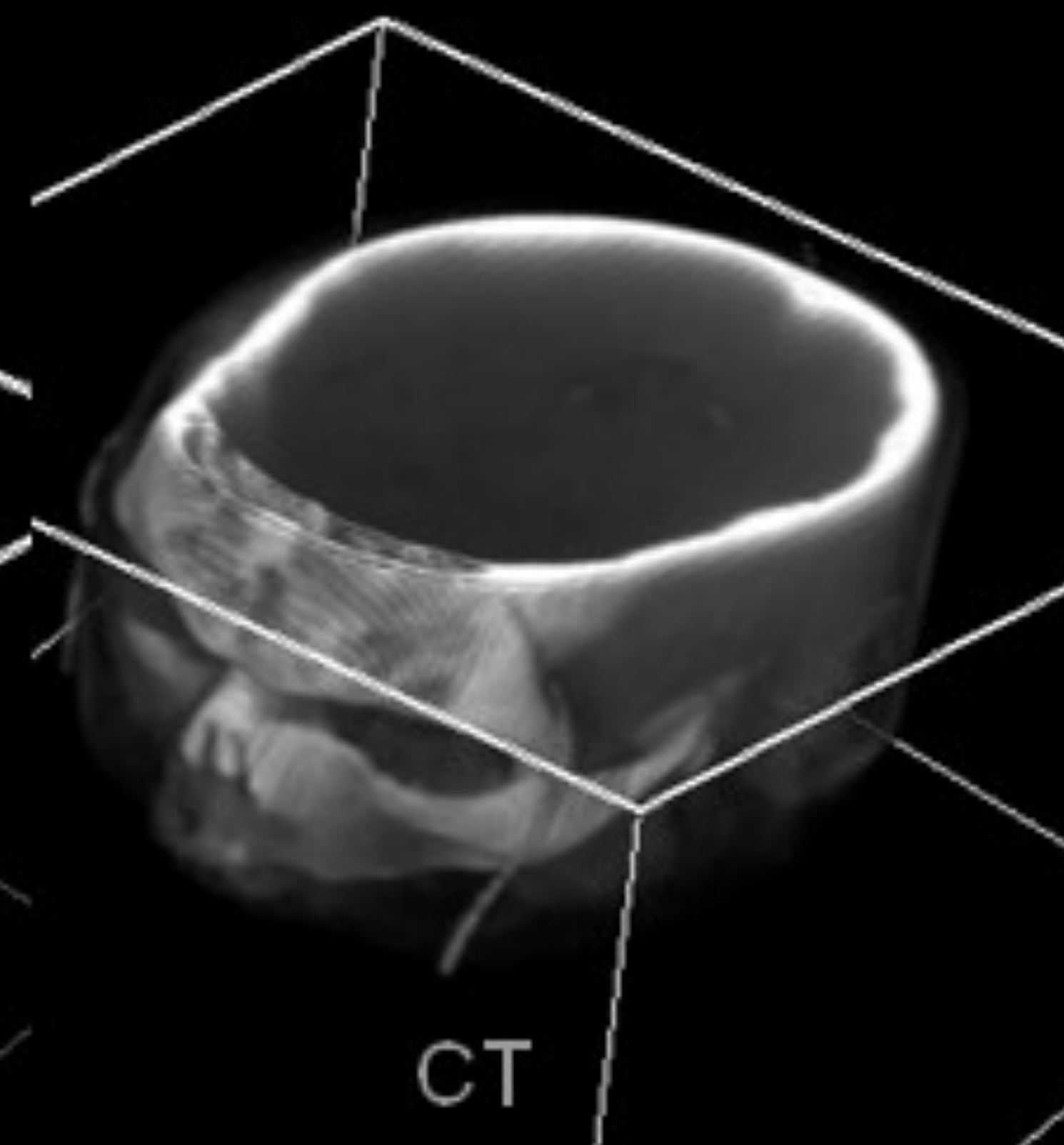
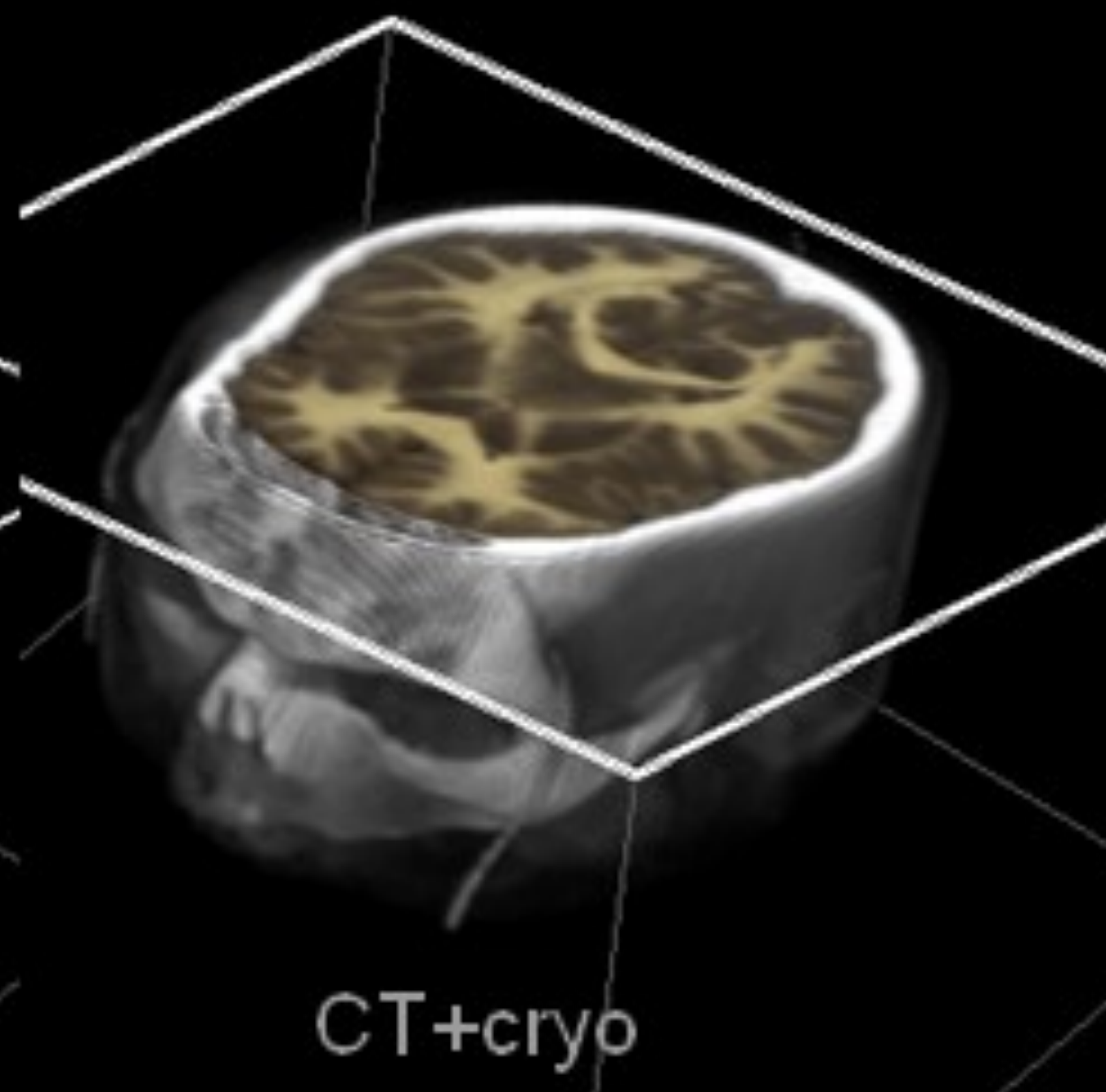
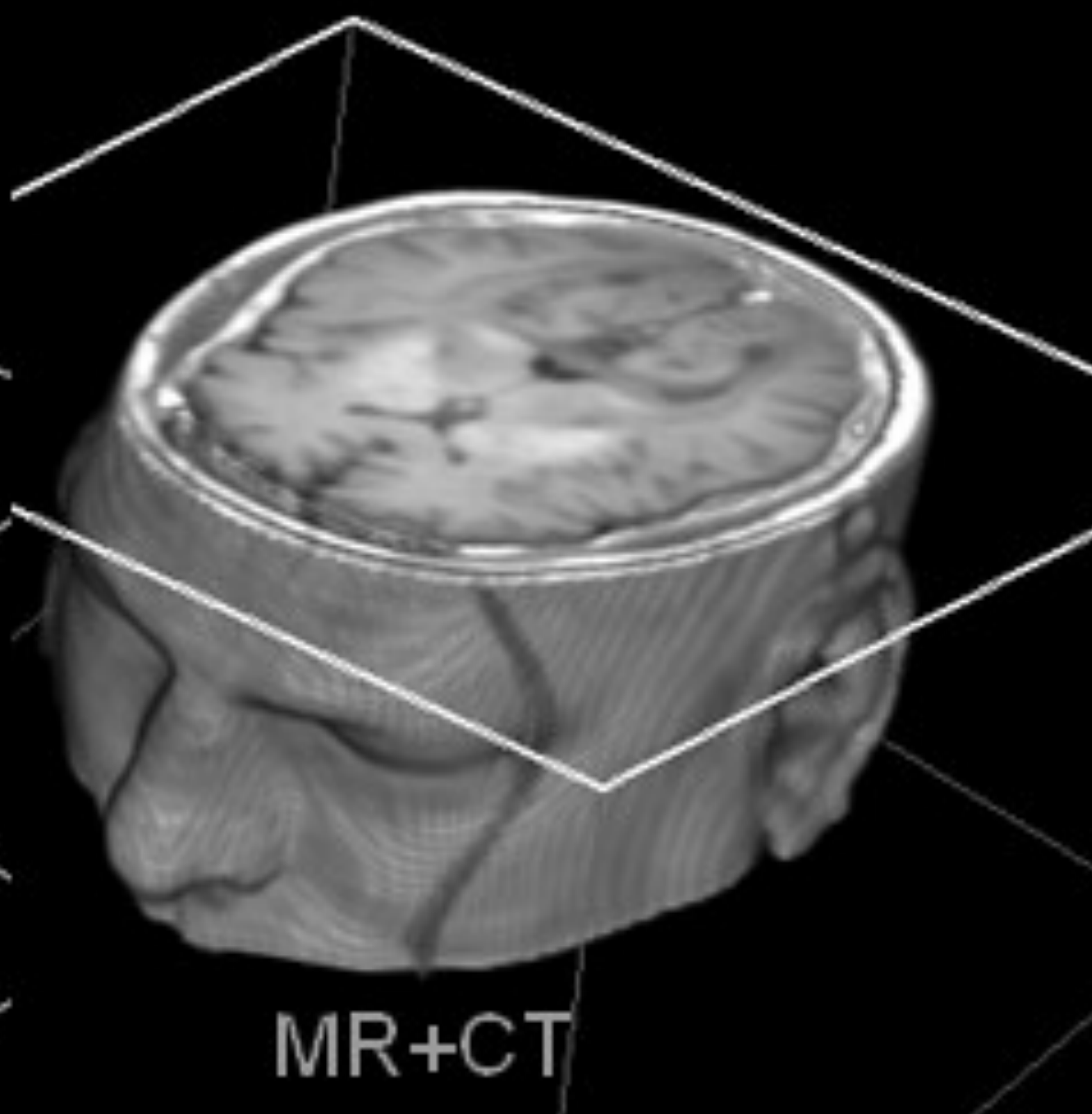
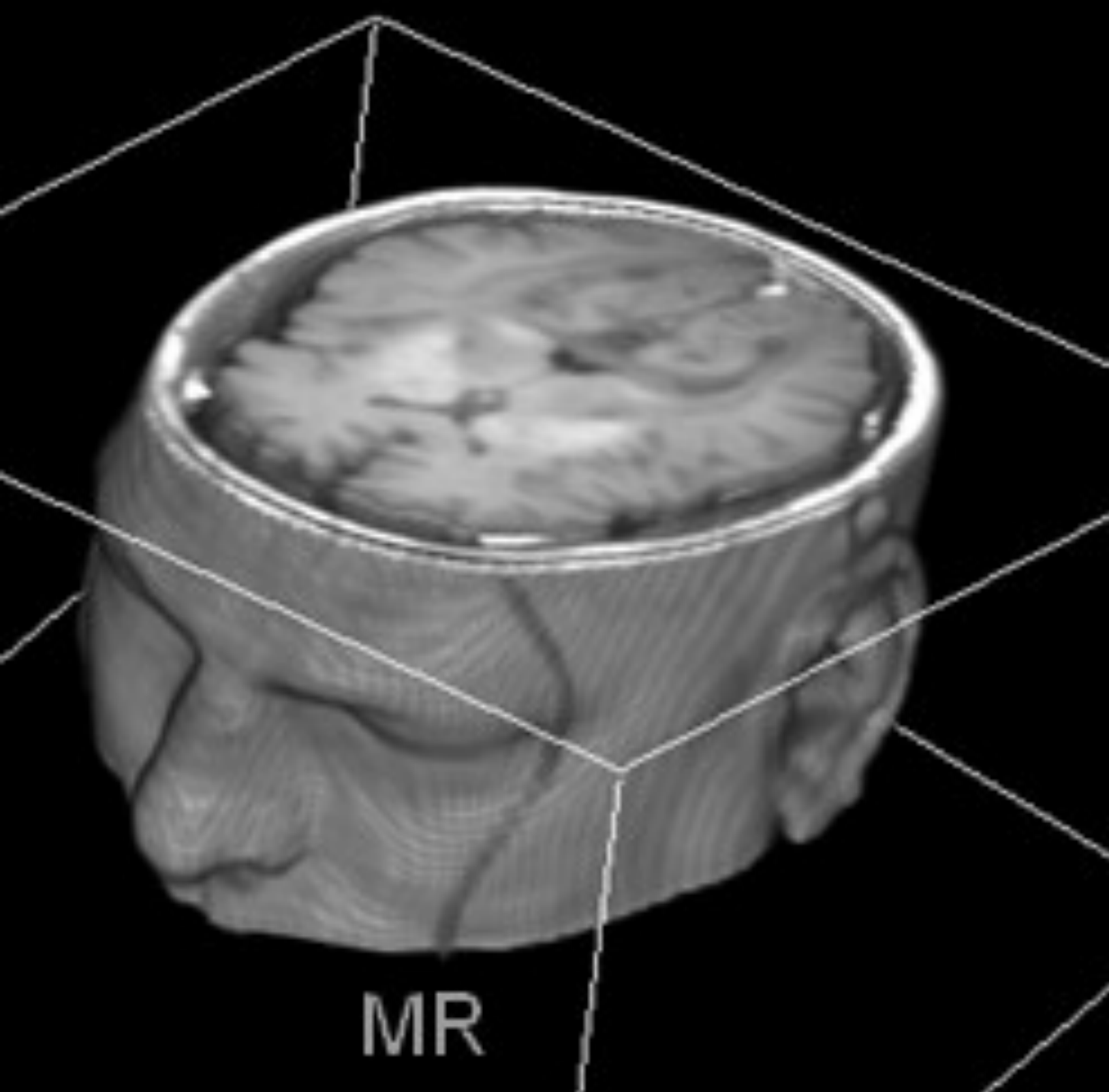
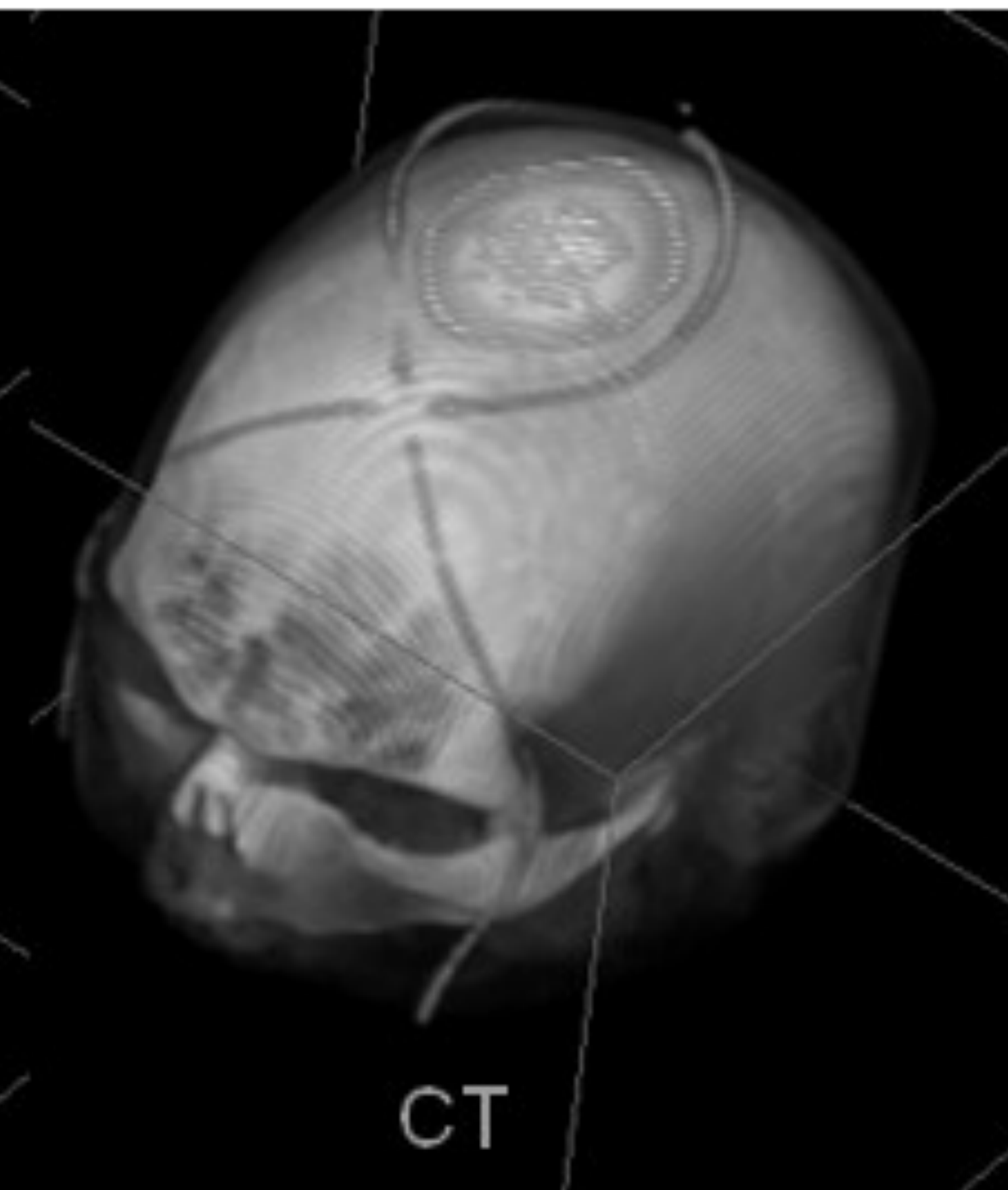
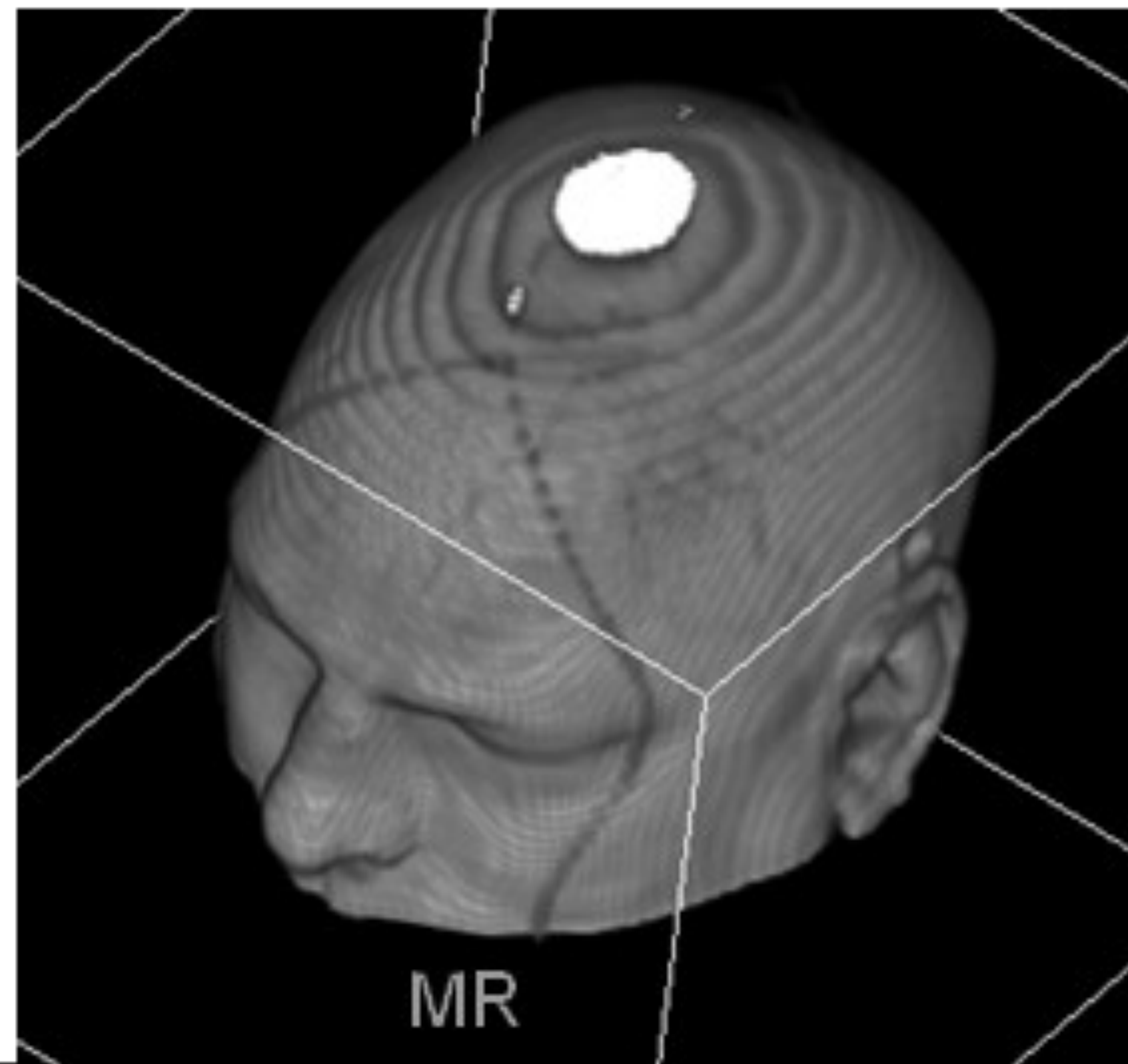


Multidimensional visualization techniques: volume rendering



Multimodality visualization techniques: volume rendering (VH)

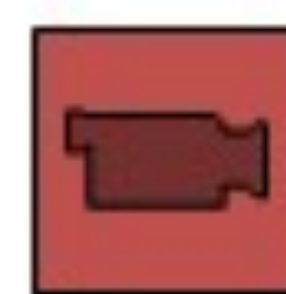
(source VH datasets courtesy NLM;
post-proc./visu. LMB)



ali.slices

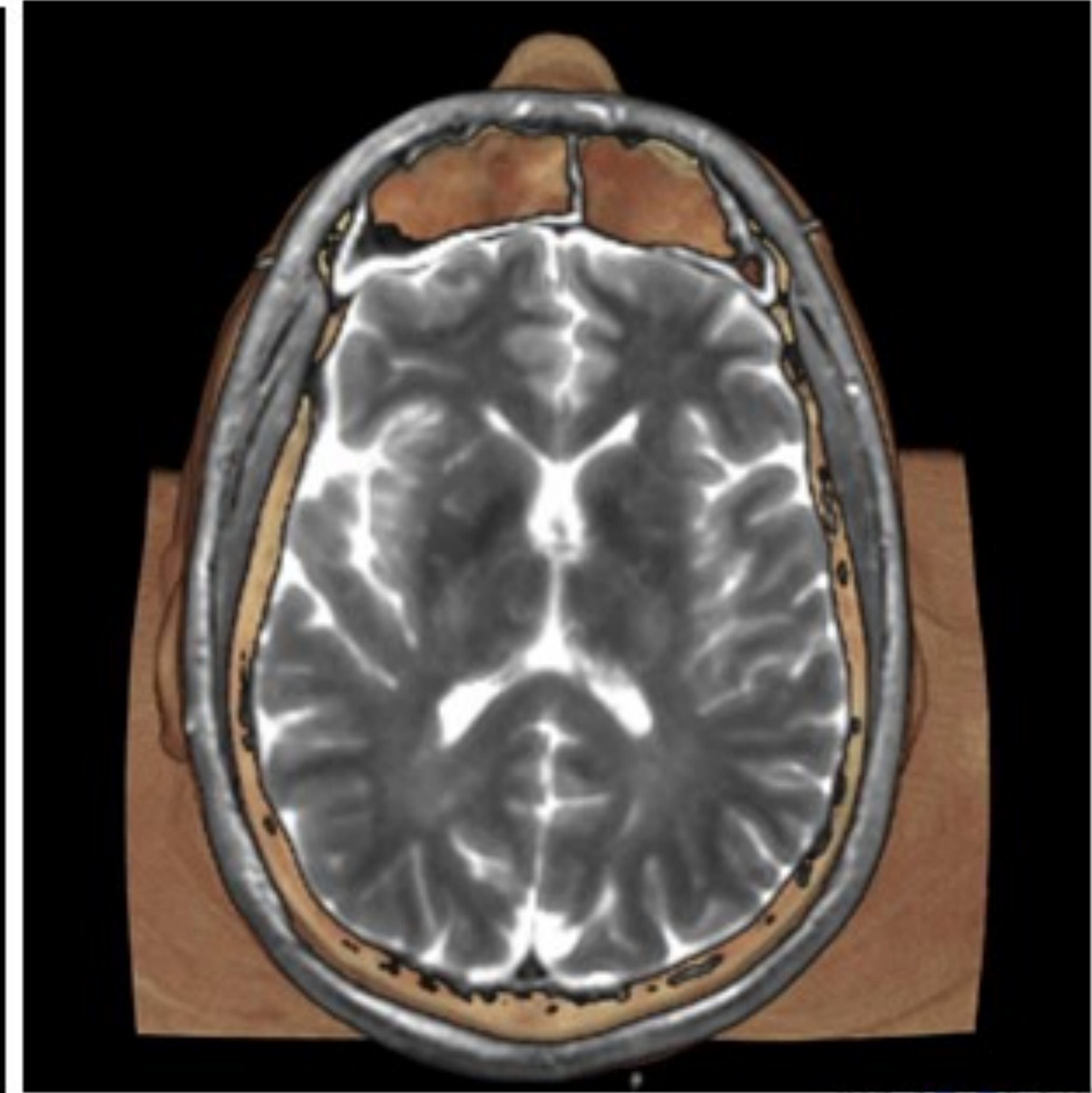
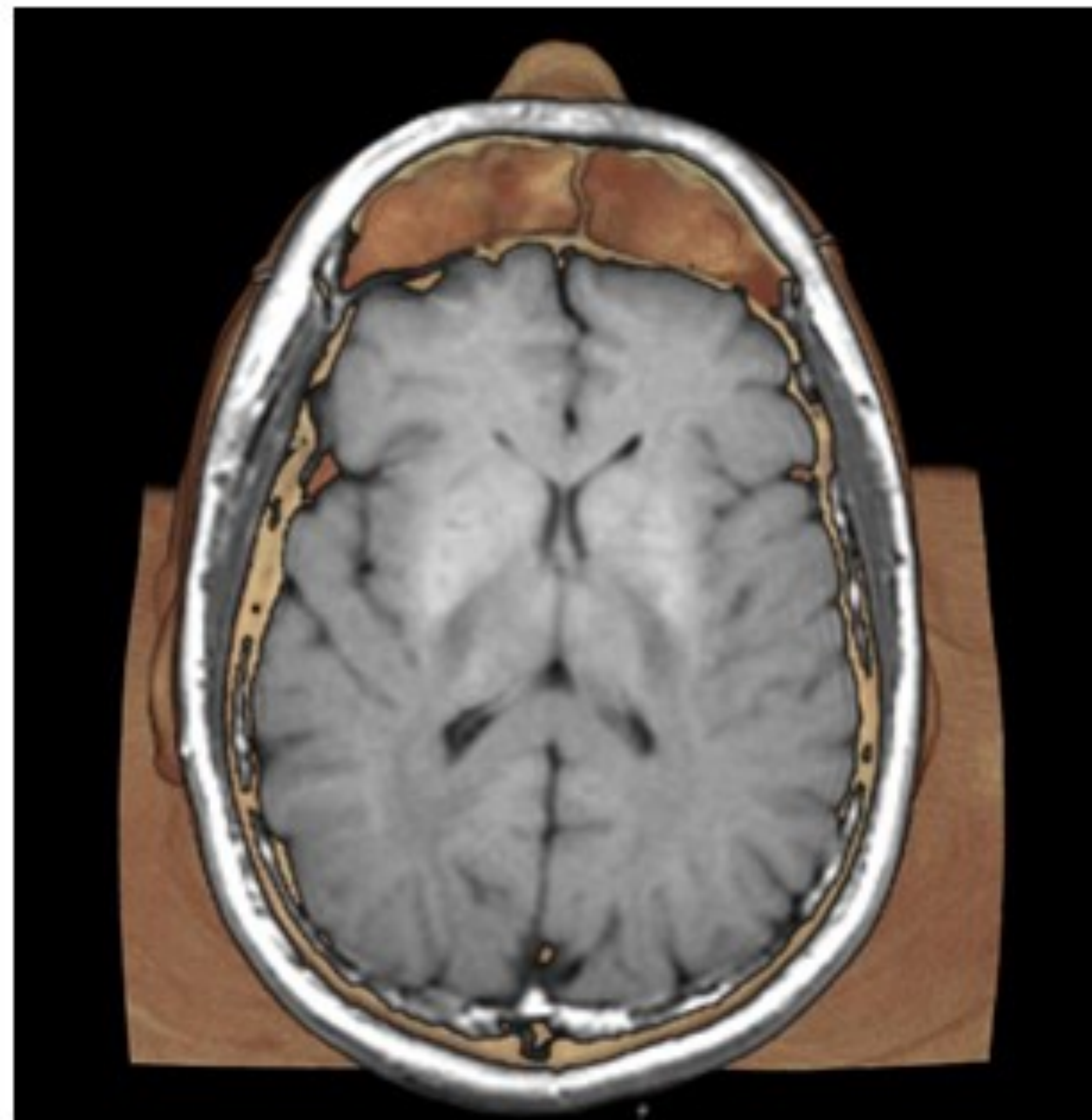
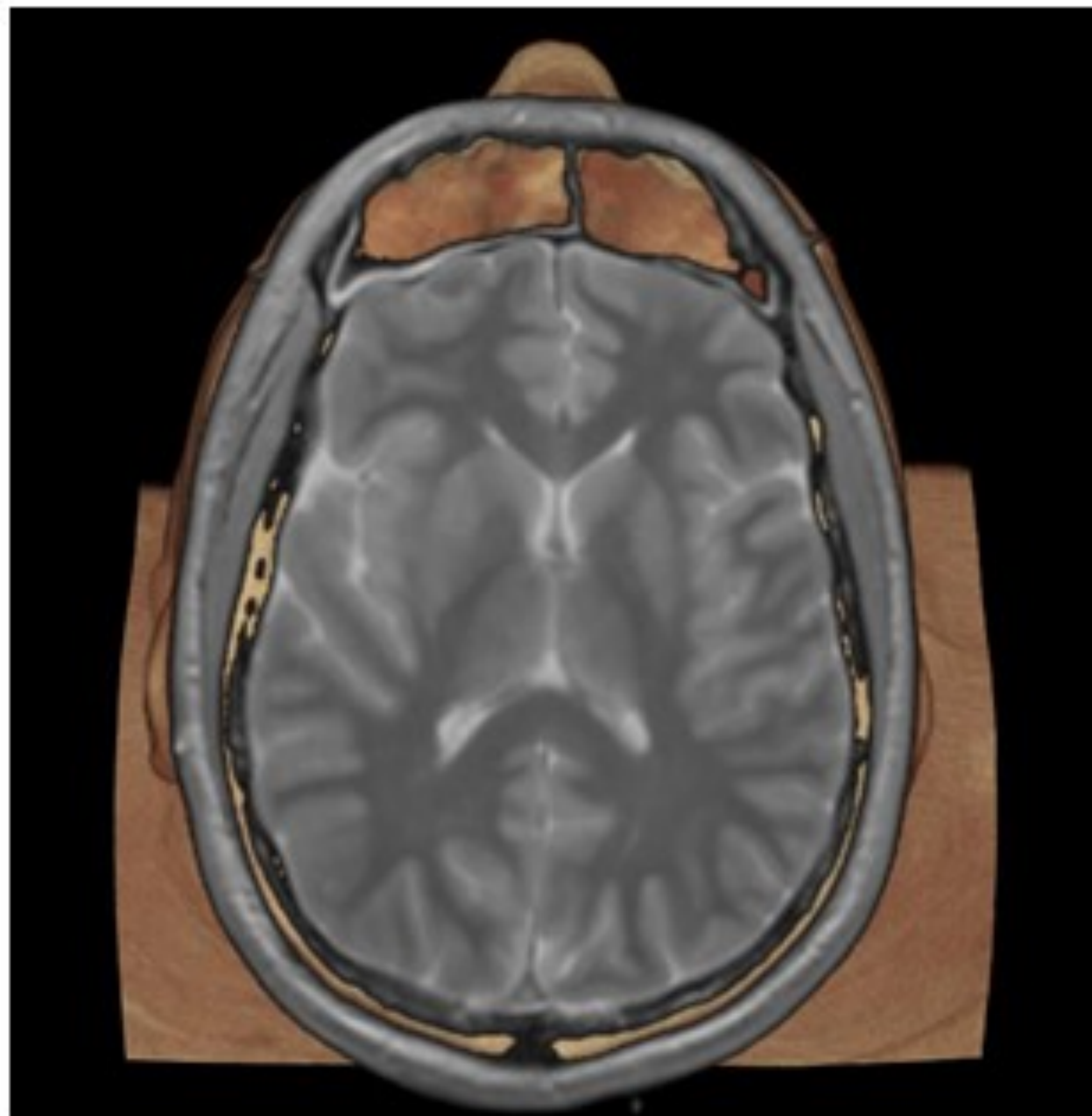
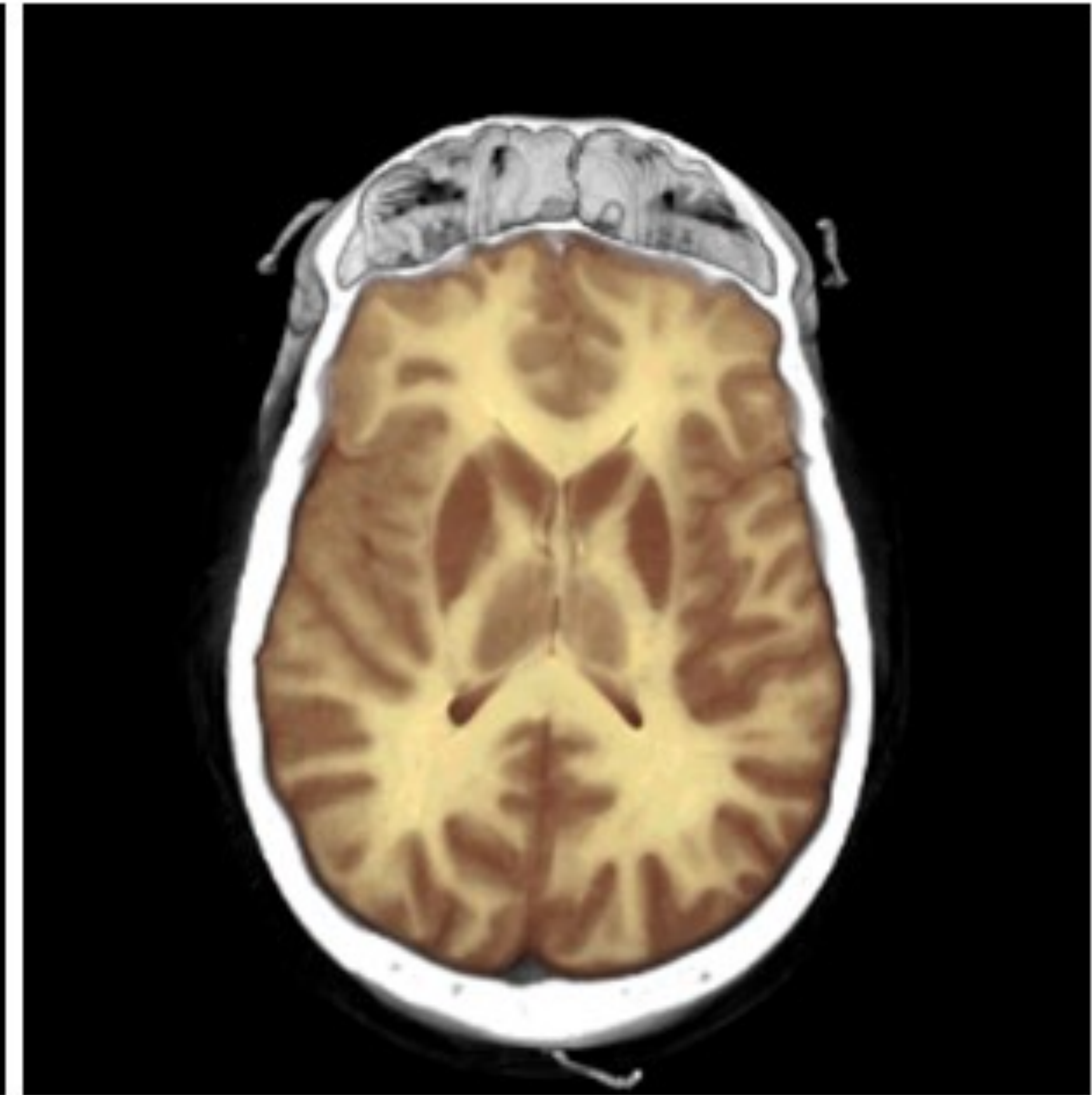
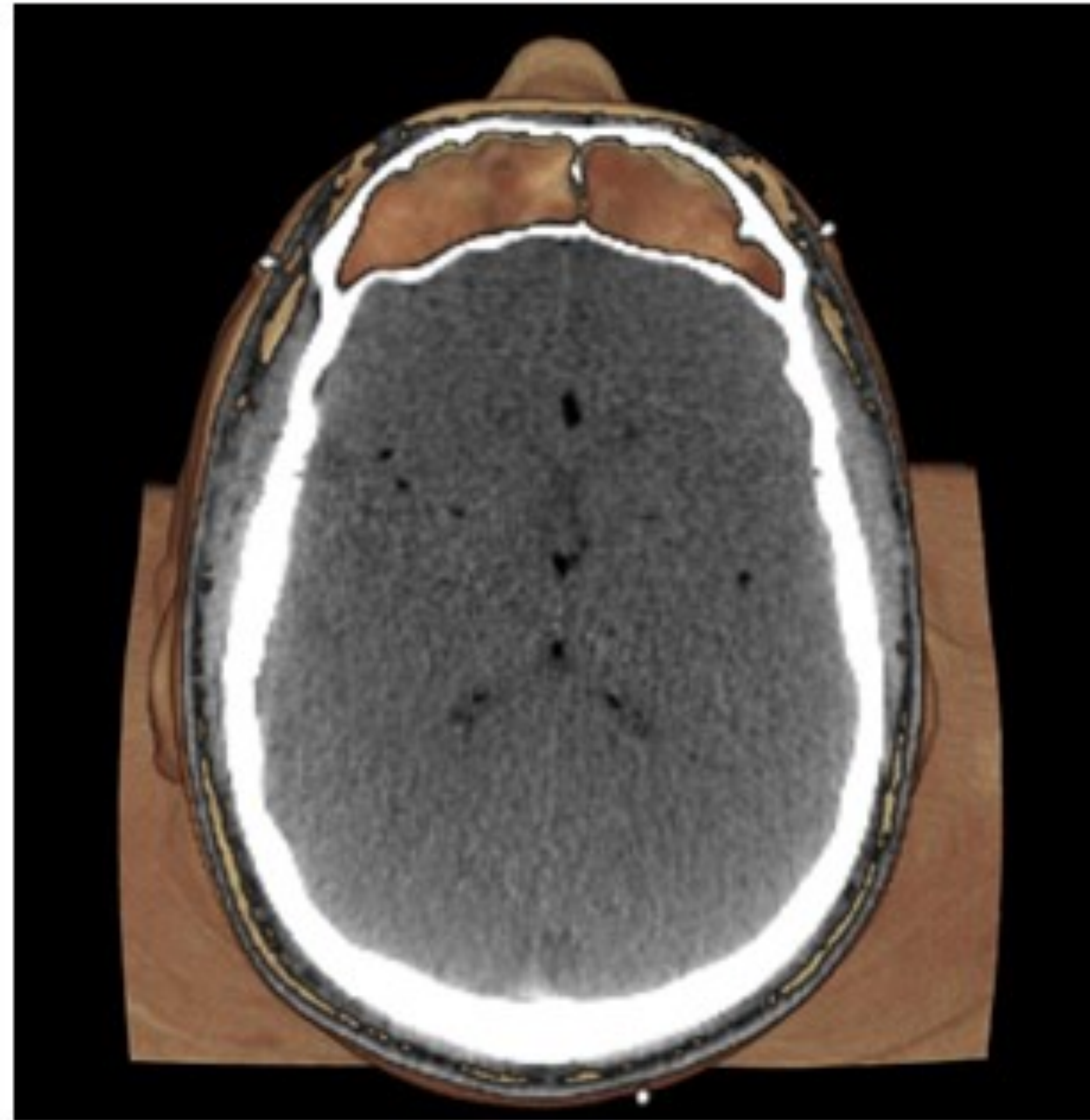


VH multimodality



rot

(source VH datasets courtesy NLM; post-proc./visu. LMB)



Imaging uses:



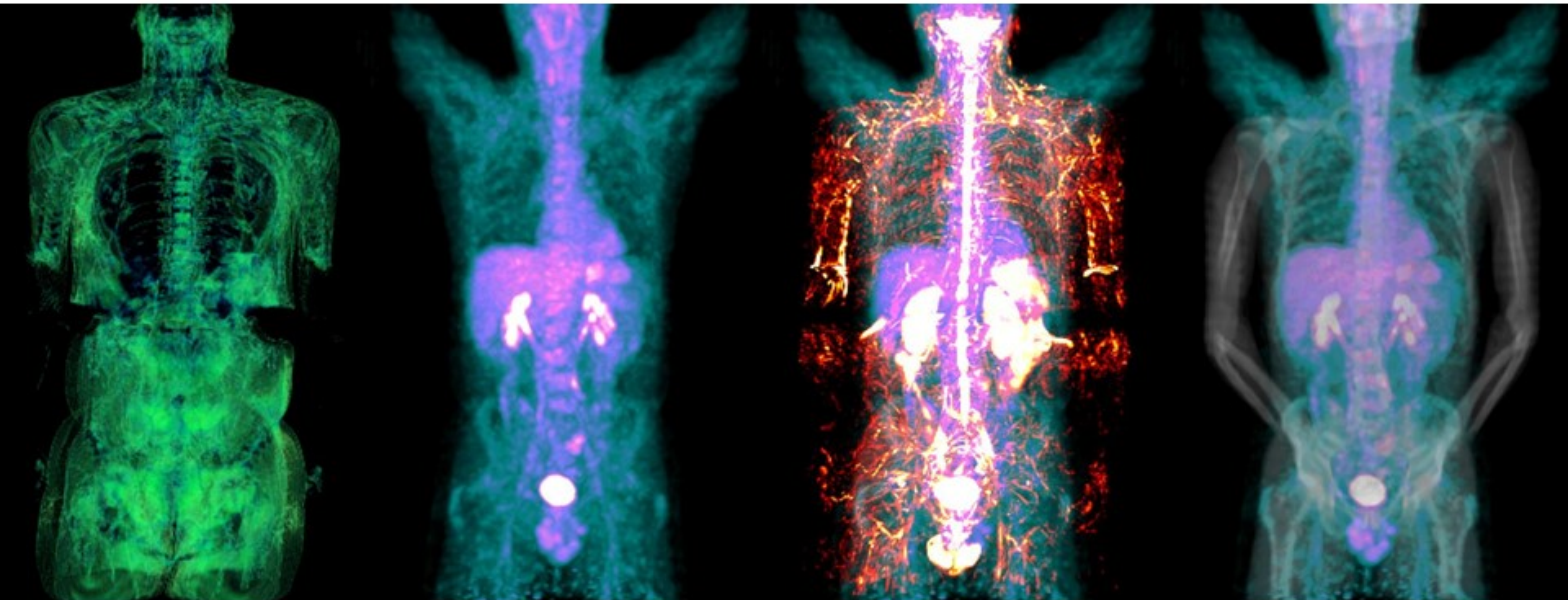
- diagnosis/staging
- treatment selection
- <- planning/targeting
- <- guidance
- response monitoring
- follow-up and restaging

Image-Guided Therapy...



Whole body MultiModality imaging

(source MR,PT(CT) courtesy MSKCC;
post-proc./visu. LMB)



MRt1

PET

PET+MRir

PET+CT



vr



Whole body MultiModality imaging

(source 3Dus,PT/CT courtesy MSKCC;
post-proc./visu. LMB)



MSKCC

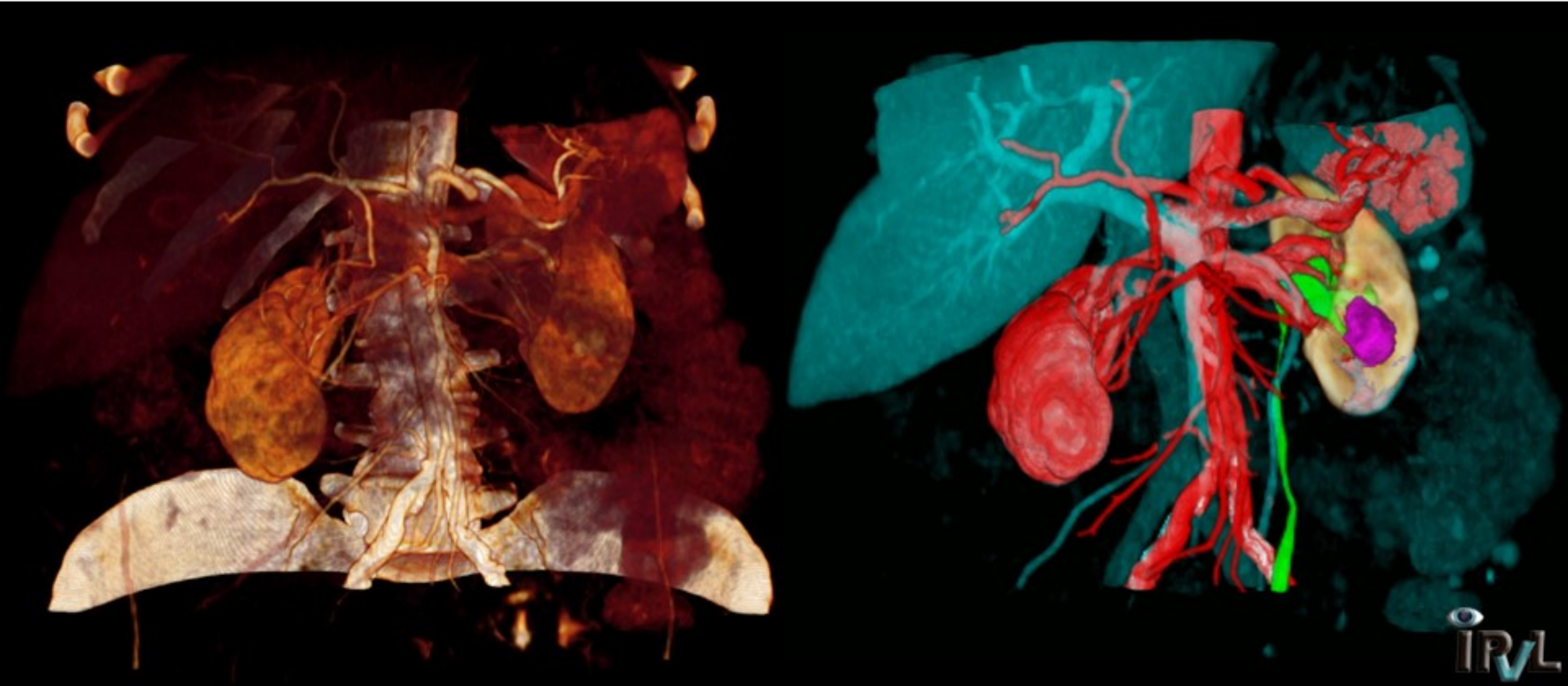


mov



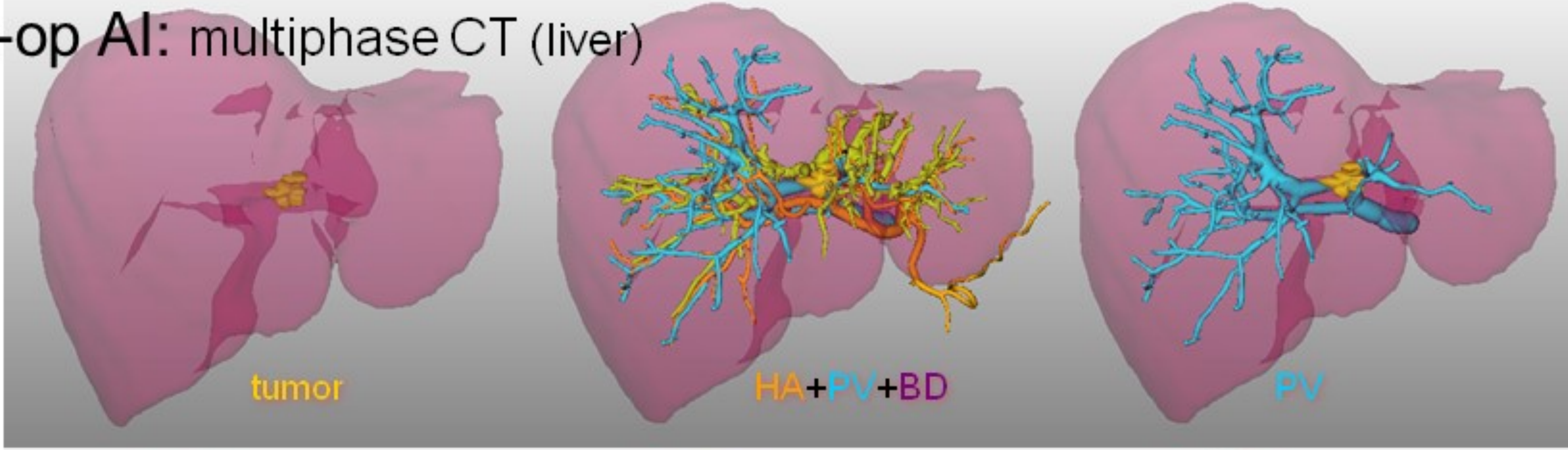
Pre-op AI: multiphase CT (kidney)

(source CTs courtesy MDACC;
post-proc./visu. LMB)



Pre-op AI: multiphase CT (liver)

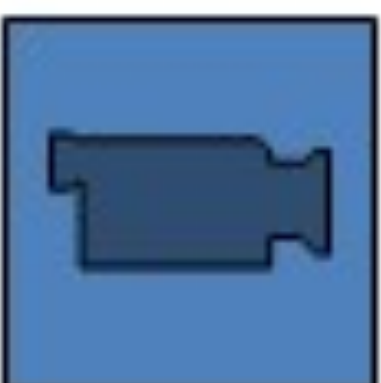
(collab.
MeVis)



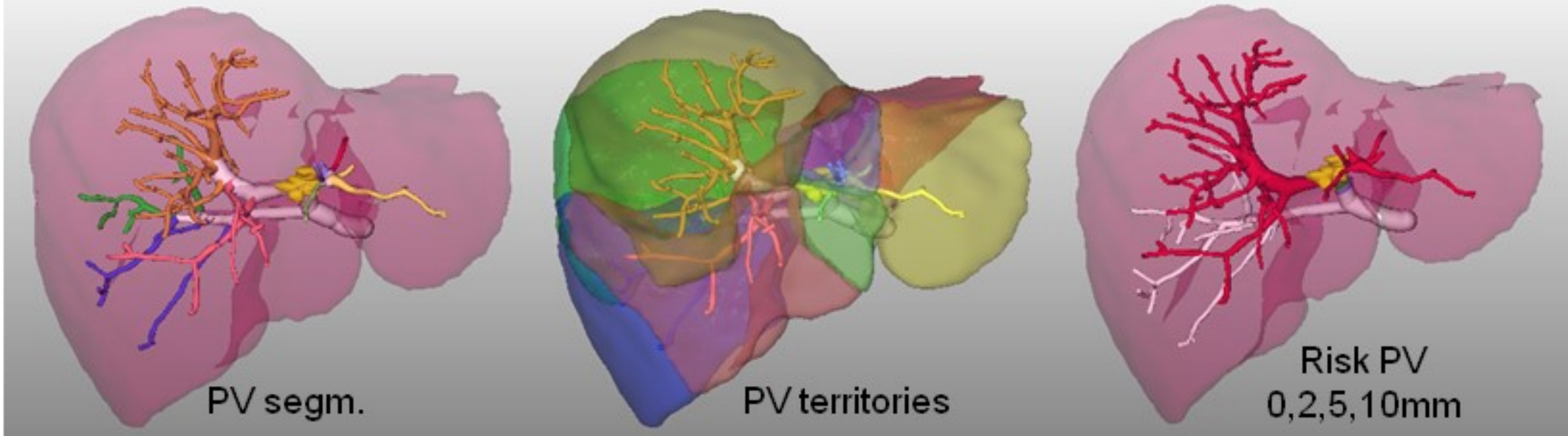
tumor

HA+PV+BD

PV



mos



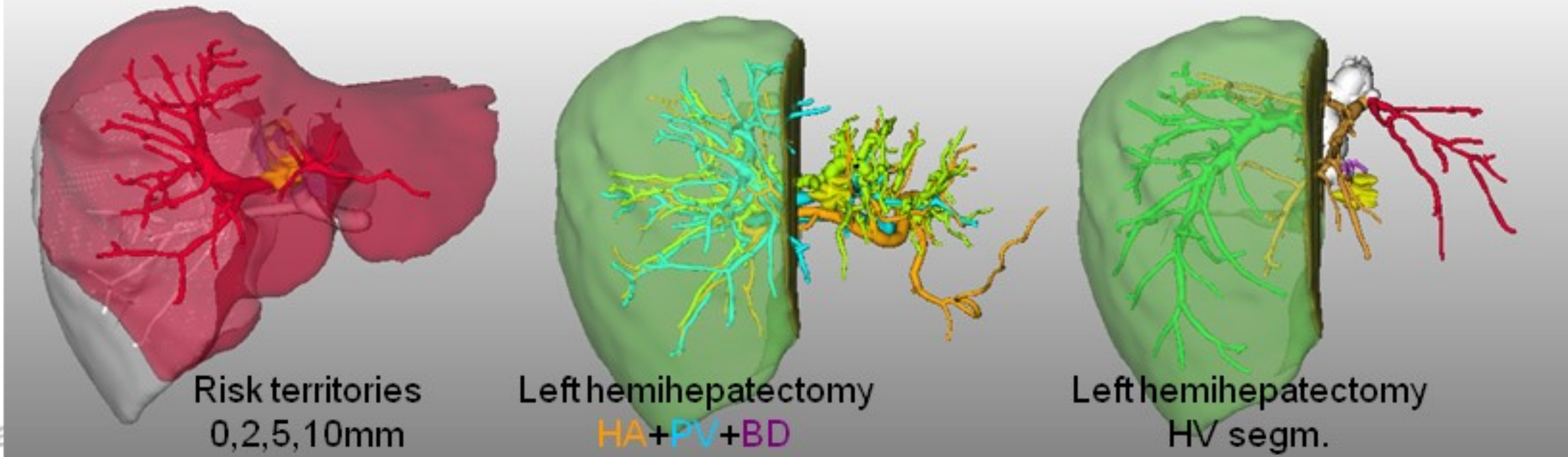
PV segm.

PV territories

Risk PV
0,2,5,10mm



vr



Risk territories
0,2,5,10mm

Left hemihepatectomy
HA+PV+BD

Left hemihepatectomy
HV segm.

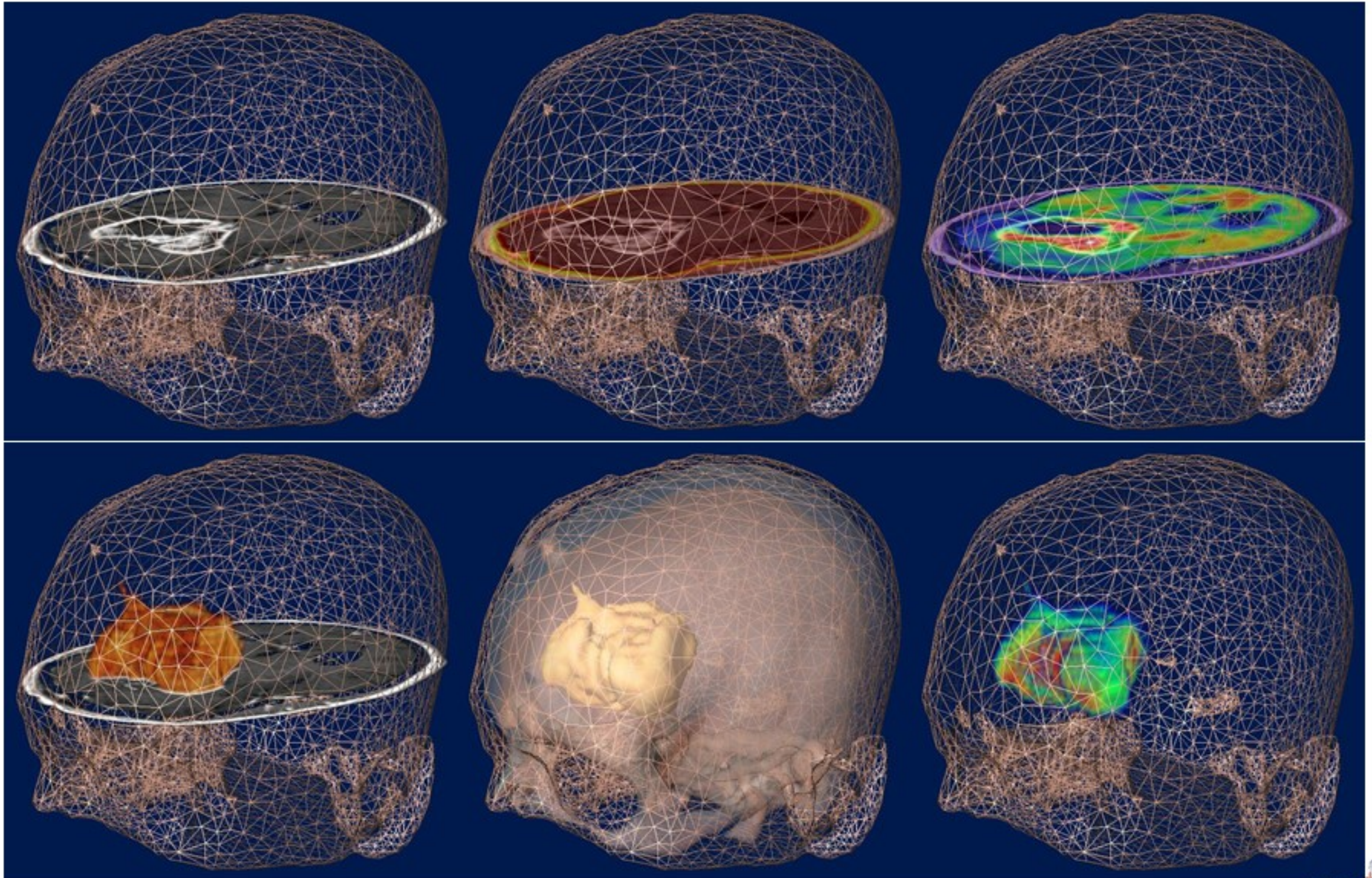


rep



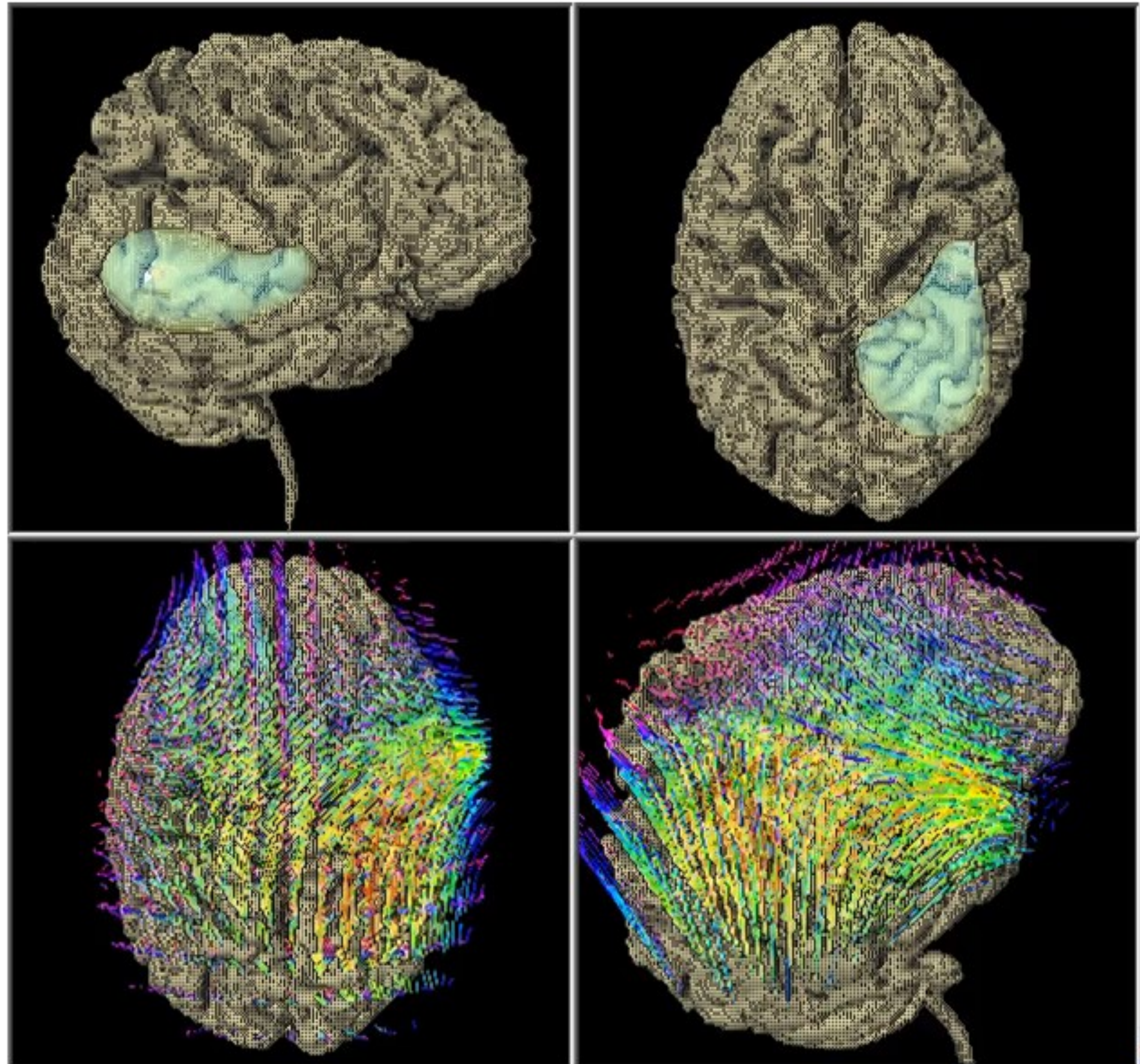
Brain tumor: MR + FDG-PET

(source data + collab. P. Gutin et al., MSKCC;
post-proc./visu. LMB)



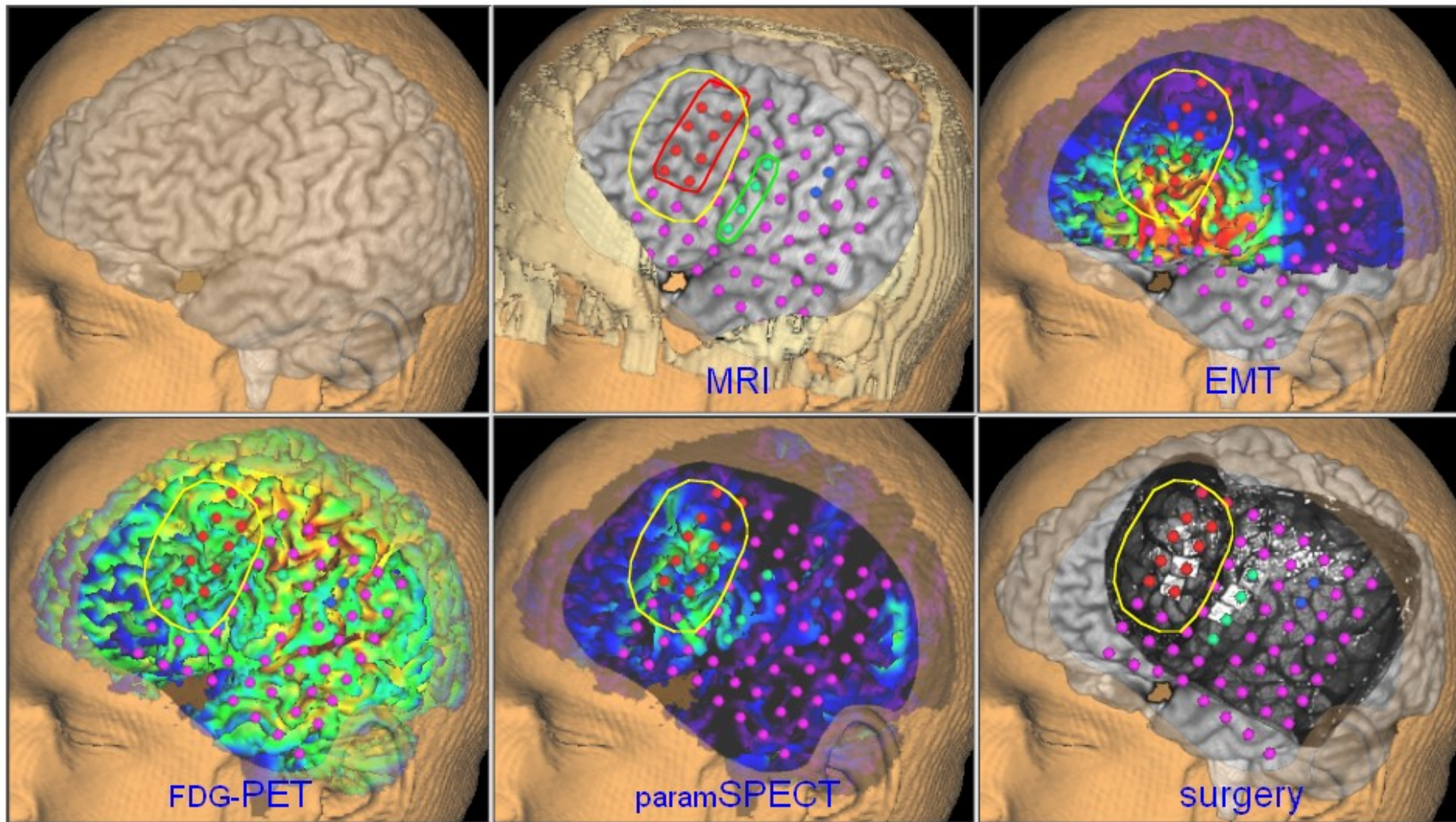
Brain function: EMT in epilepsy (from 1kHz EEG)

(source data + collabs.
HUG & CHUV; post-proc./visu. LMB)



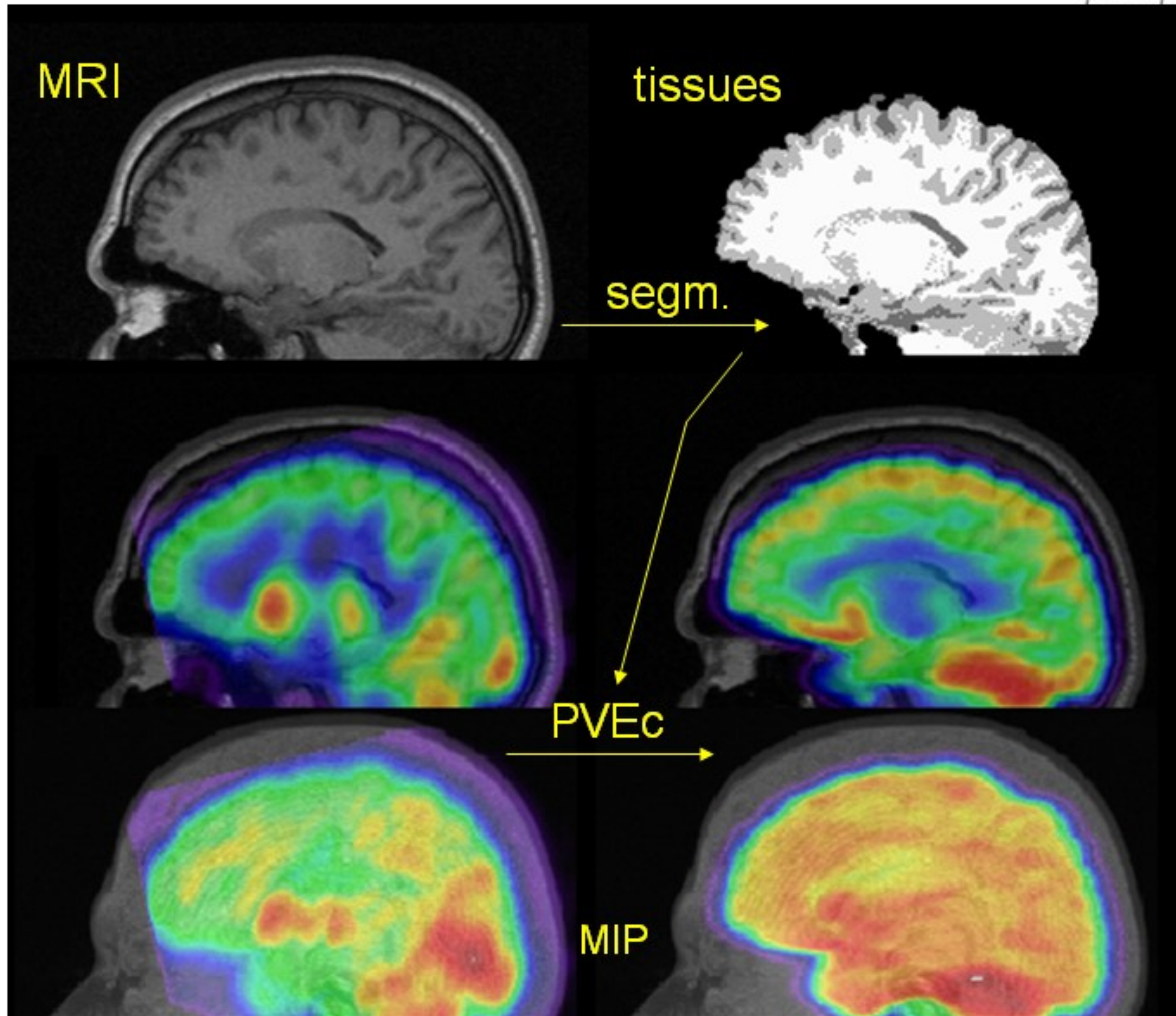
Pre/Post-op AI: neuro-epilepsy (MR + PET + SPECT + EMT + grid)

(source data + collabs.
HUG & CHUV; post-proc./visu. LMB)



Refined 3D segmentation: Partial Volume Effect correction

(source MR/SPECT/PET courtesy HUG;
post-proc./visu. LMB)



Inter subject registration:

Registering:

acquisition

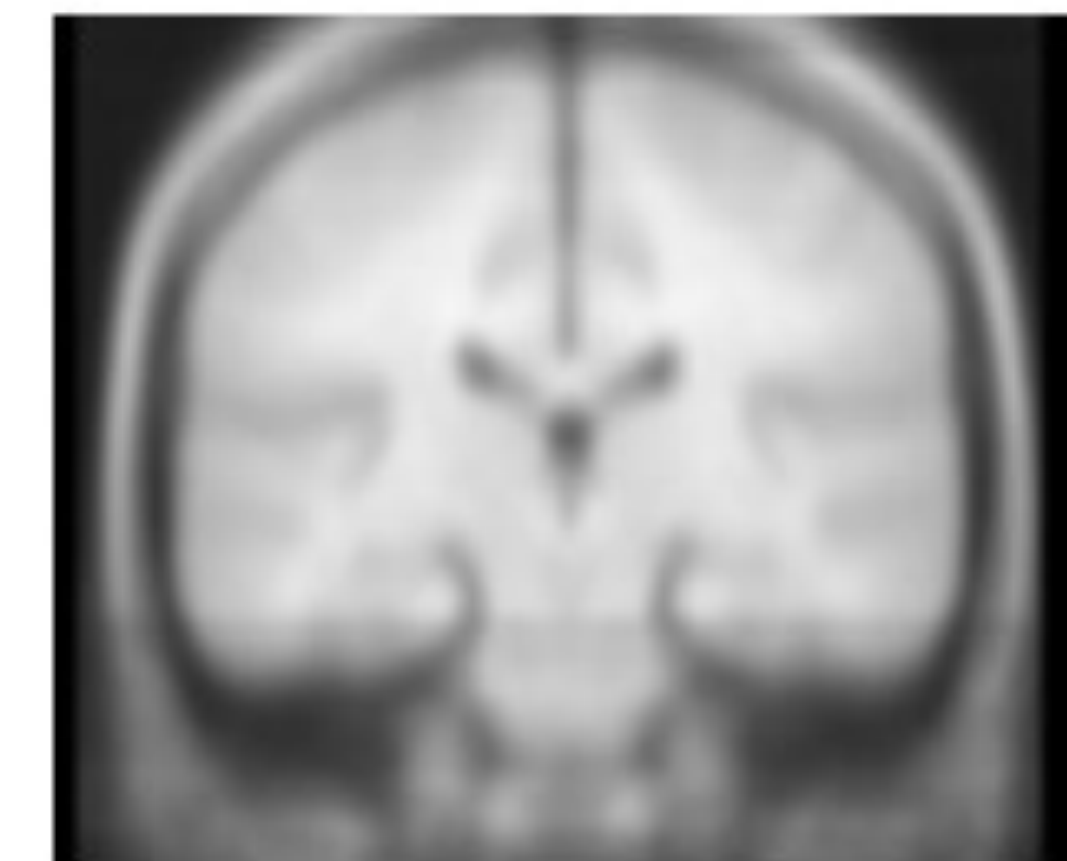
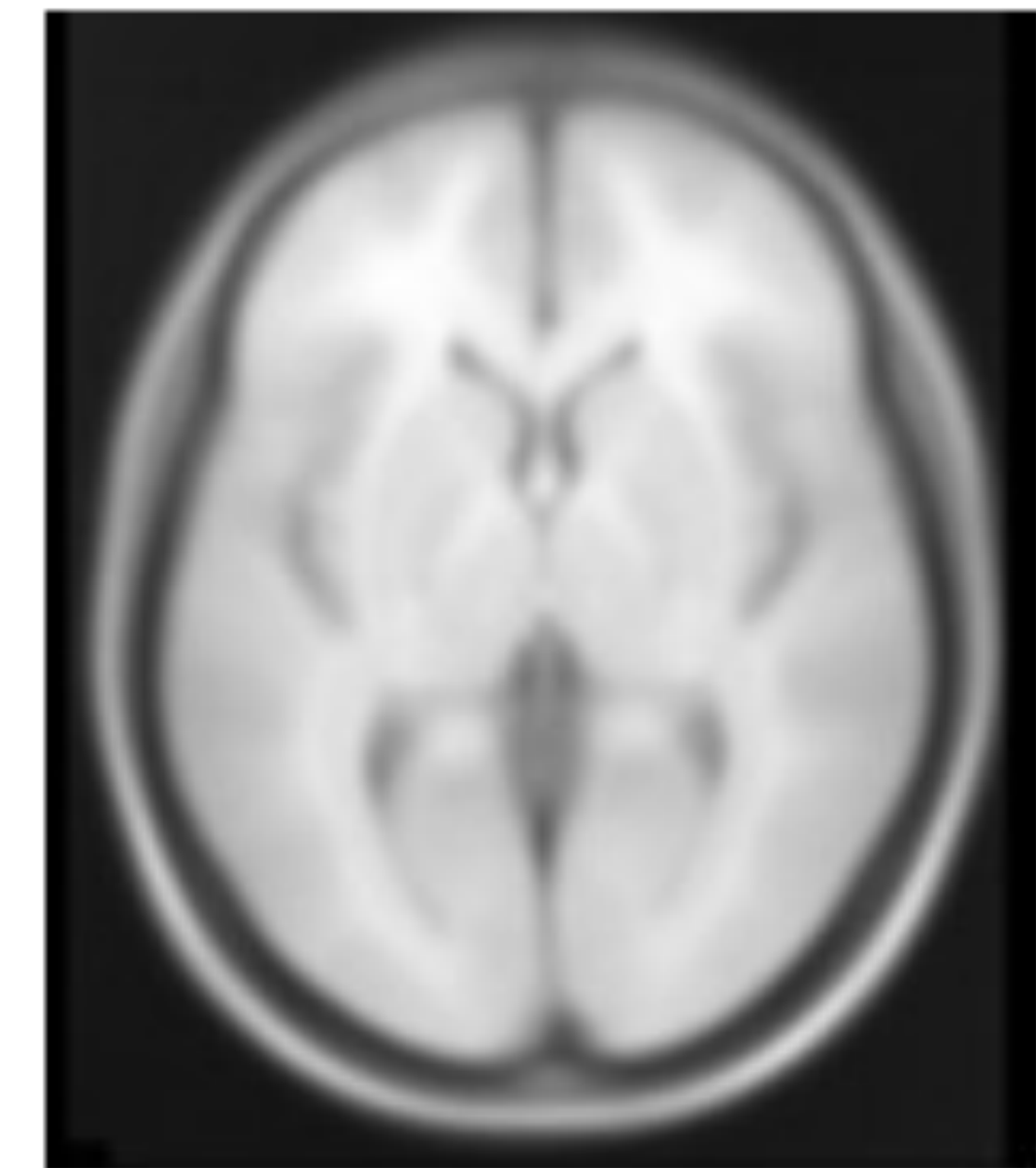
registration

rigid

affine

averaging

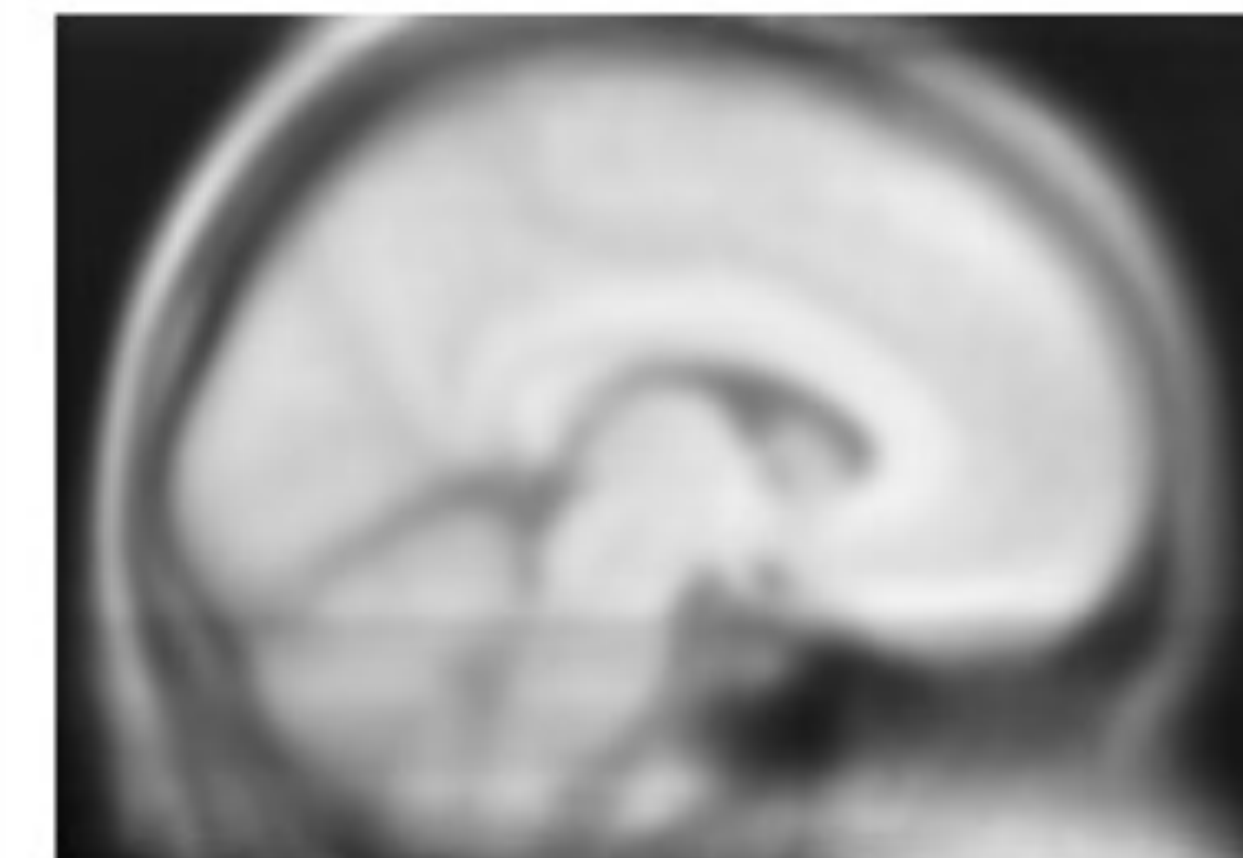
average MR brain



Exploitation ('à la' SPM):

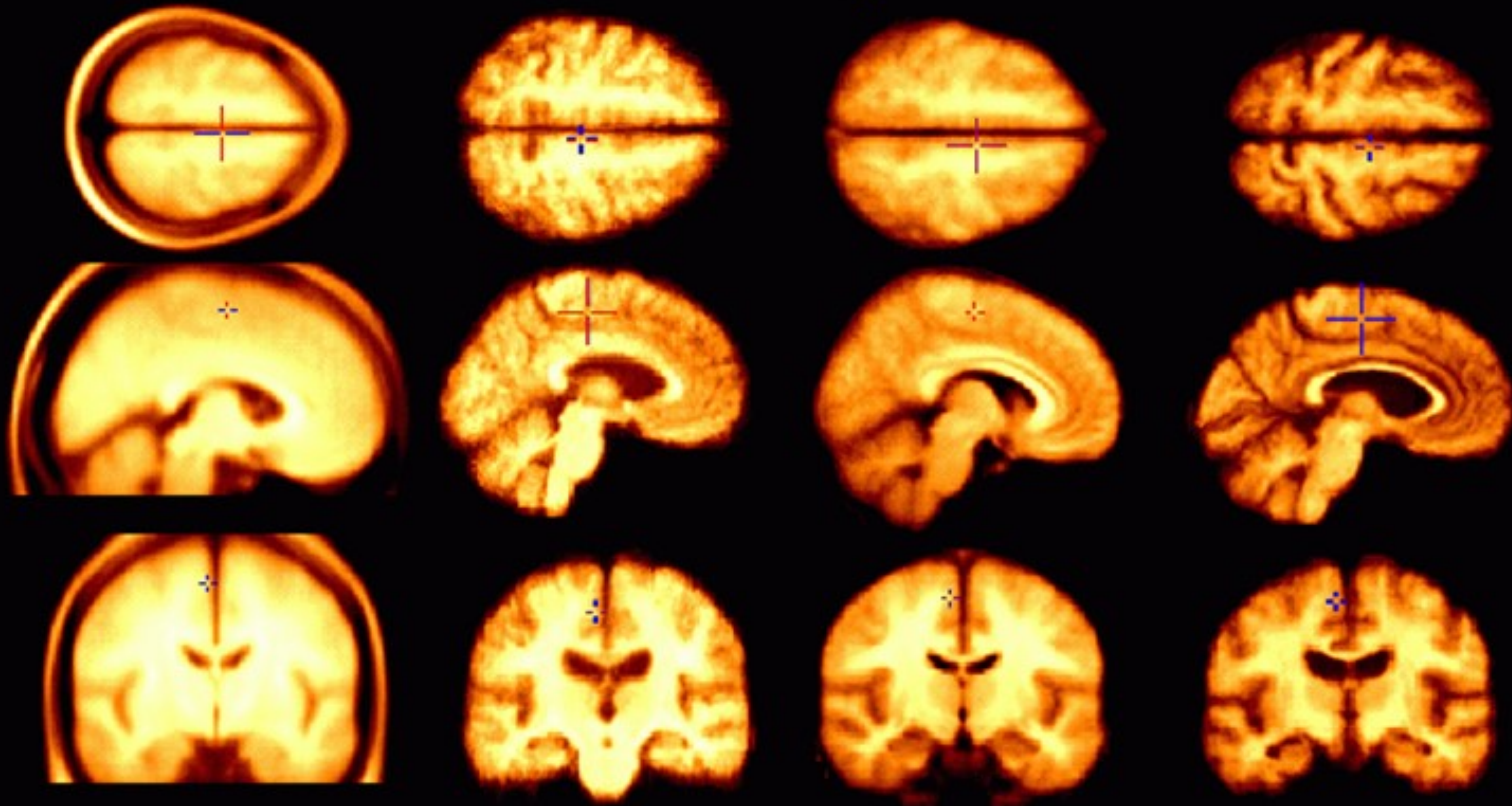
high (spatial) frequency smoothing

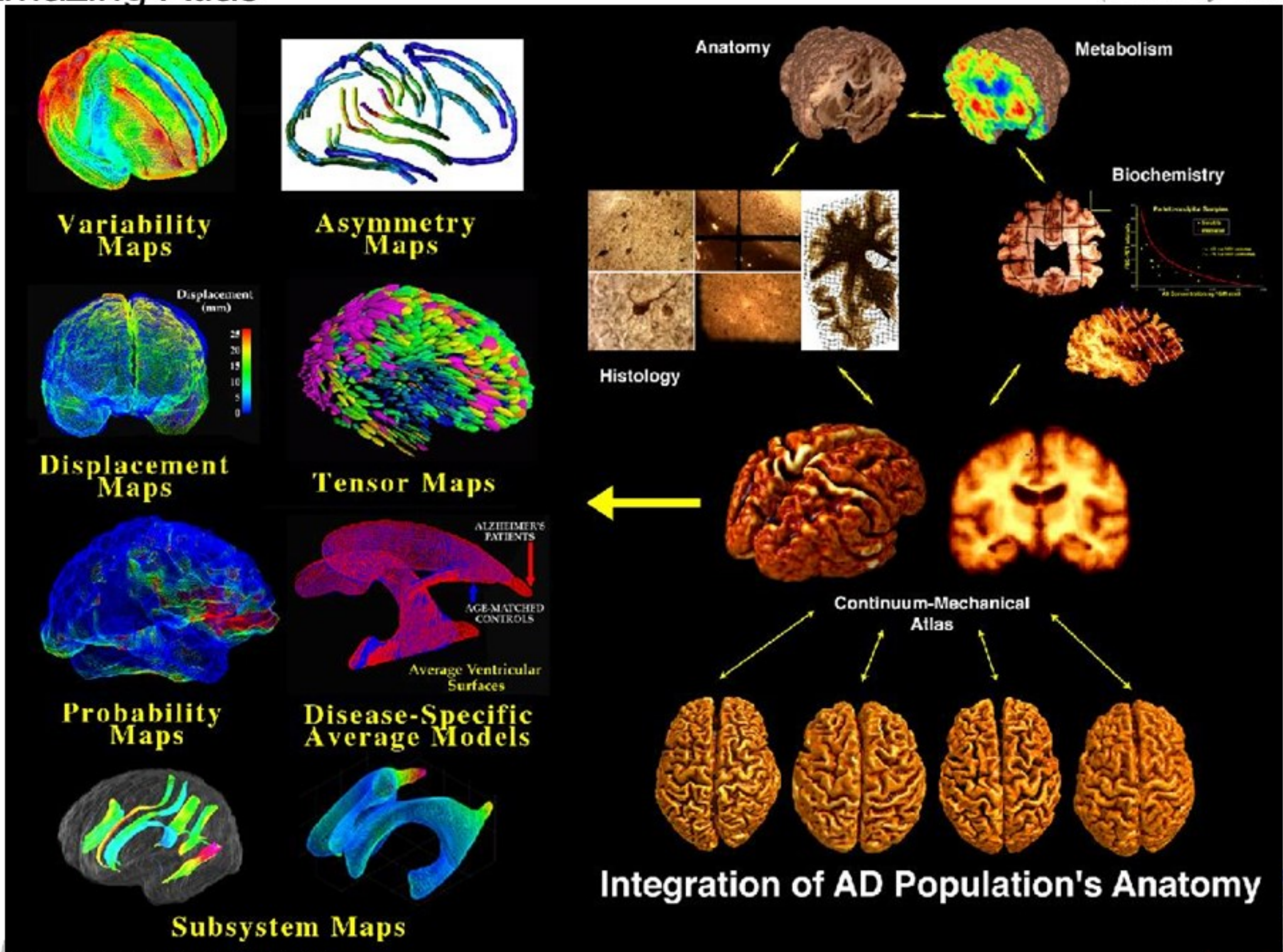
statistical analysis



Average Brain Templates

ICBM305 **Affine** **8th order Polynomial** **Continuum-Mechanical**



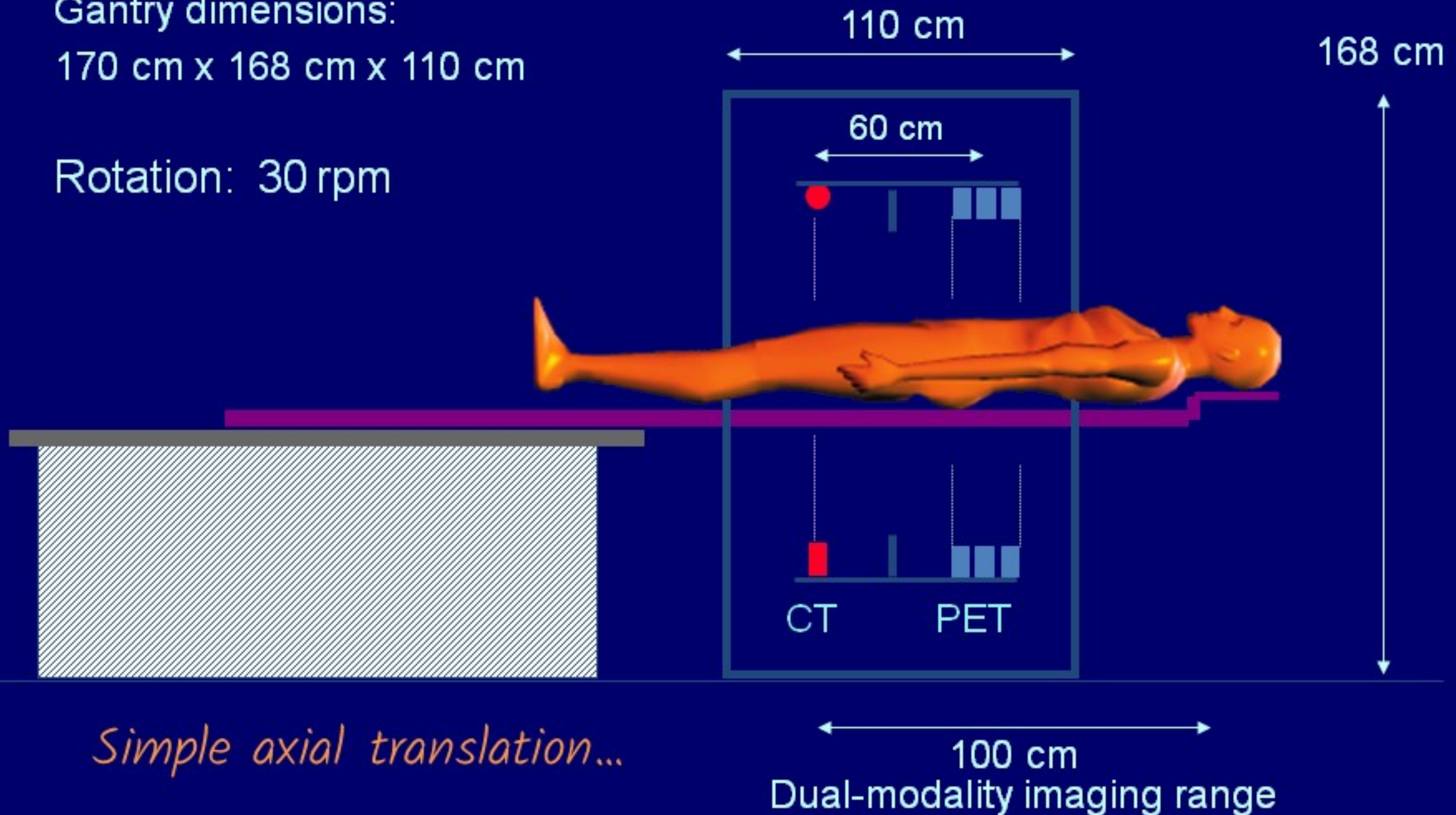


Multimodality/sensor: hybrid machines: PET/CTs, etc... mark II design

(derived from D. Townsend)

Gantry dimensions:
170 cm x 168 cm x 110 cm

Rotation: 30 rpm

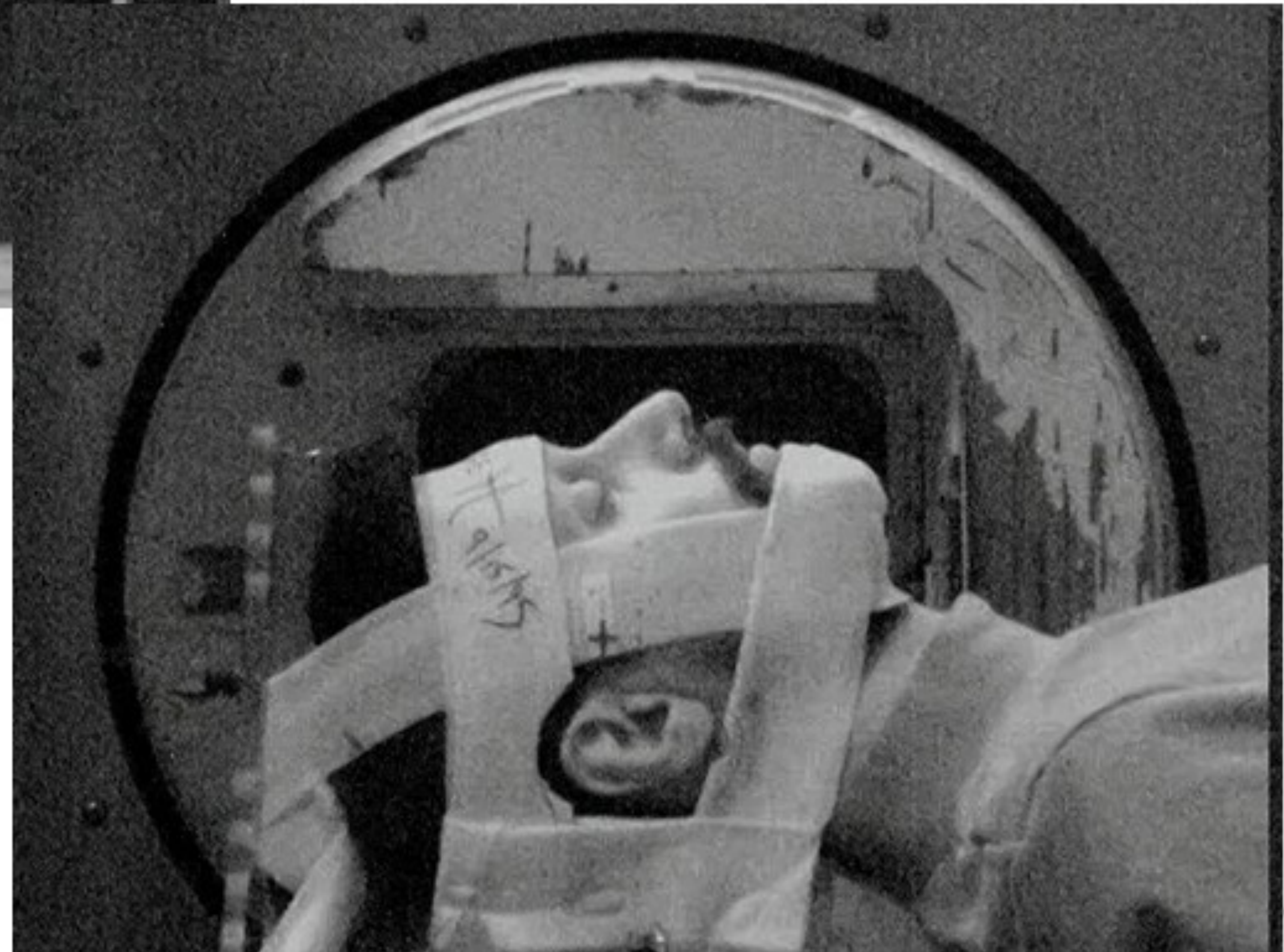
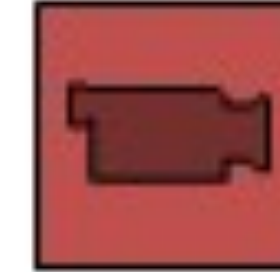
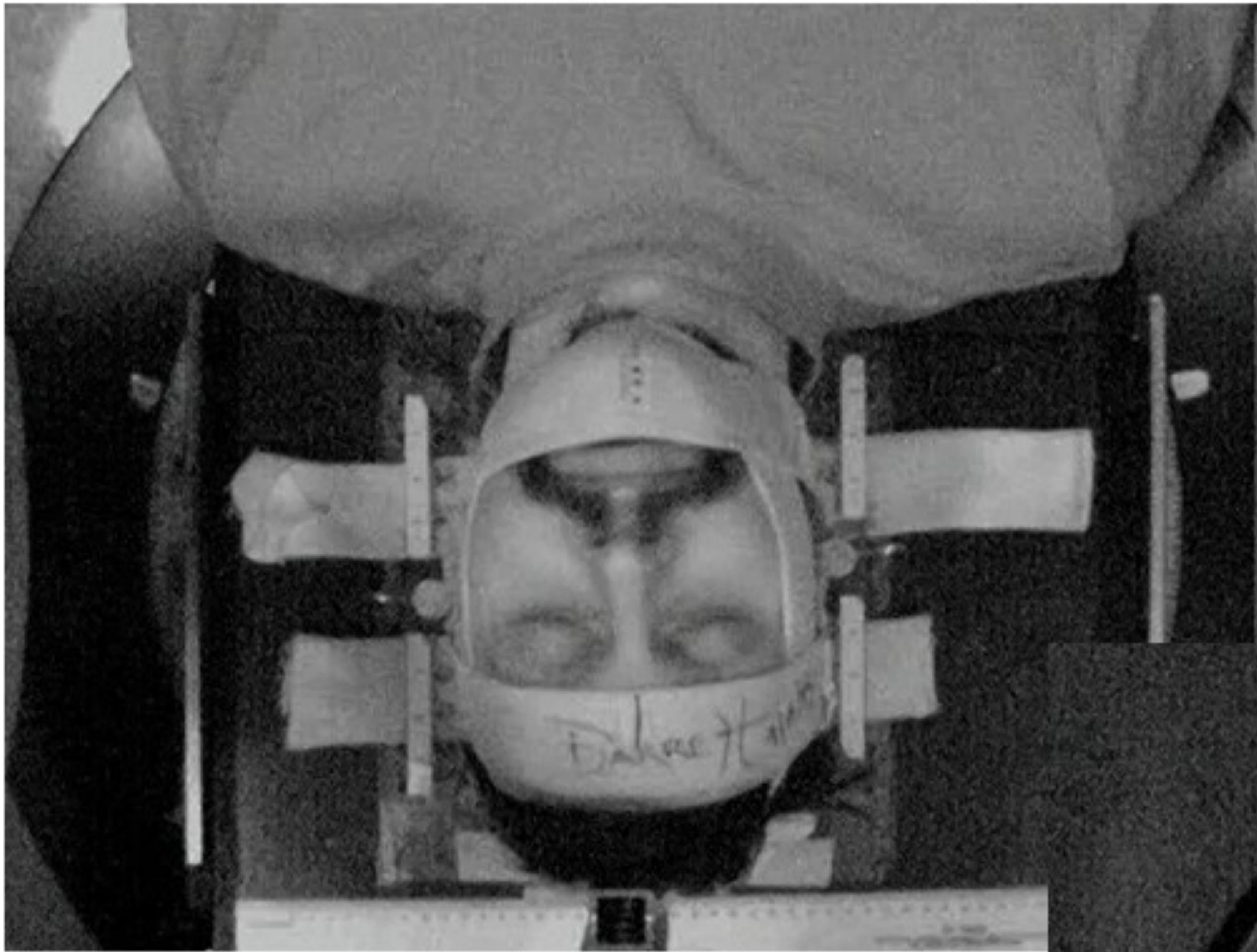


- non simultaneous
- different time scale (CT: few secs.; PT: several mins.)

and motions...

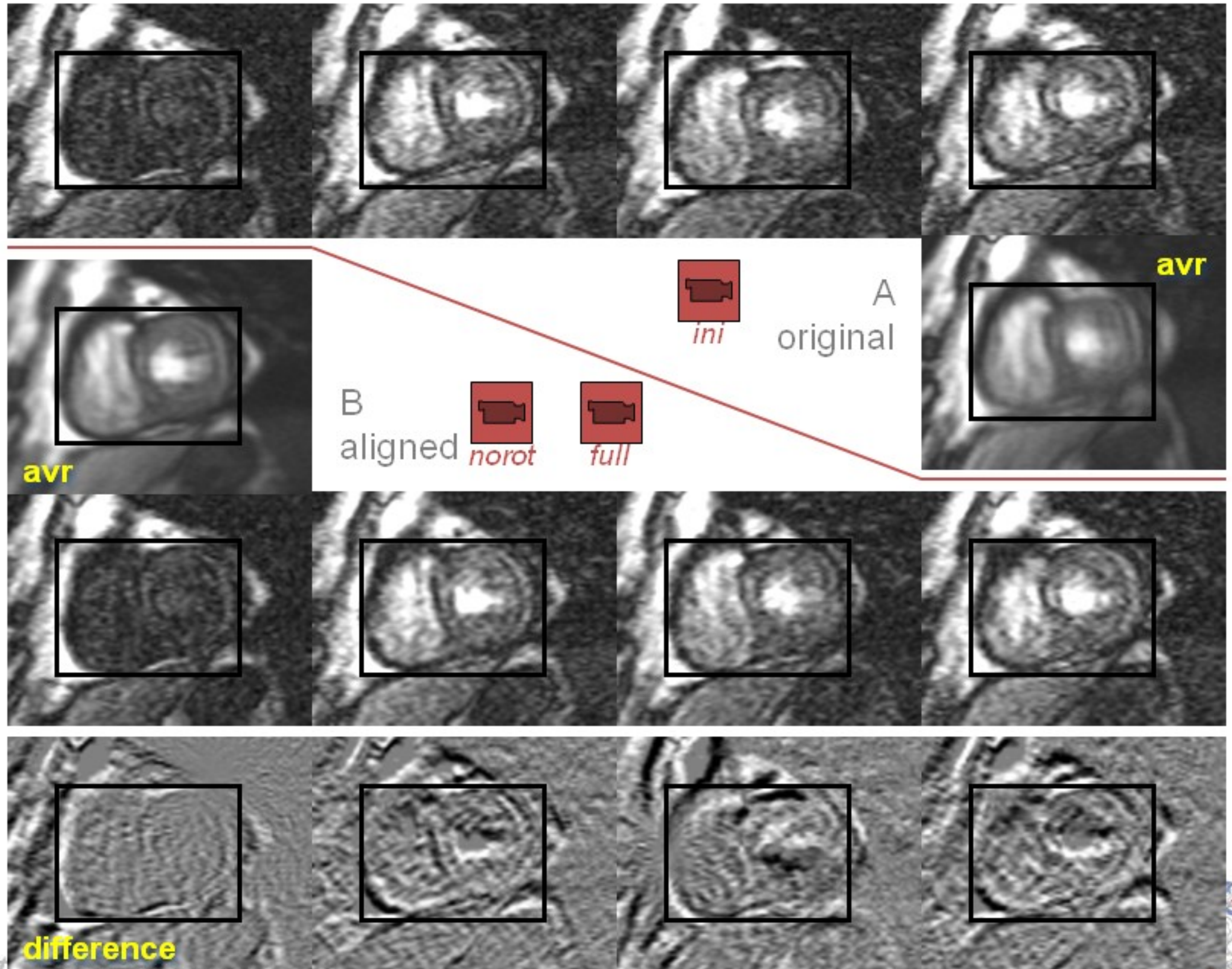


Real need for registration? (motion...)



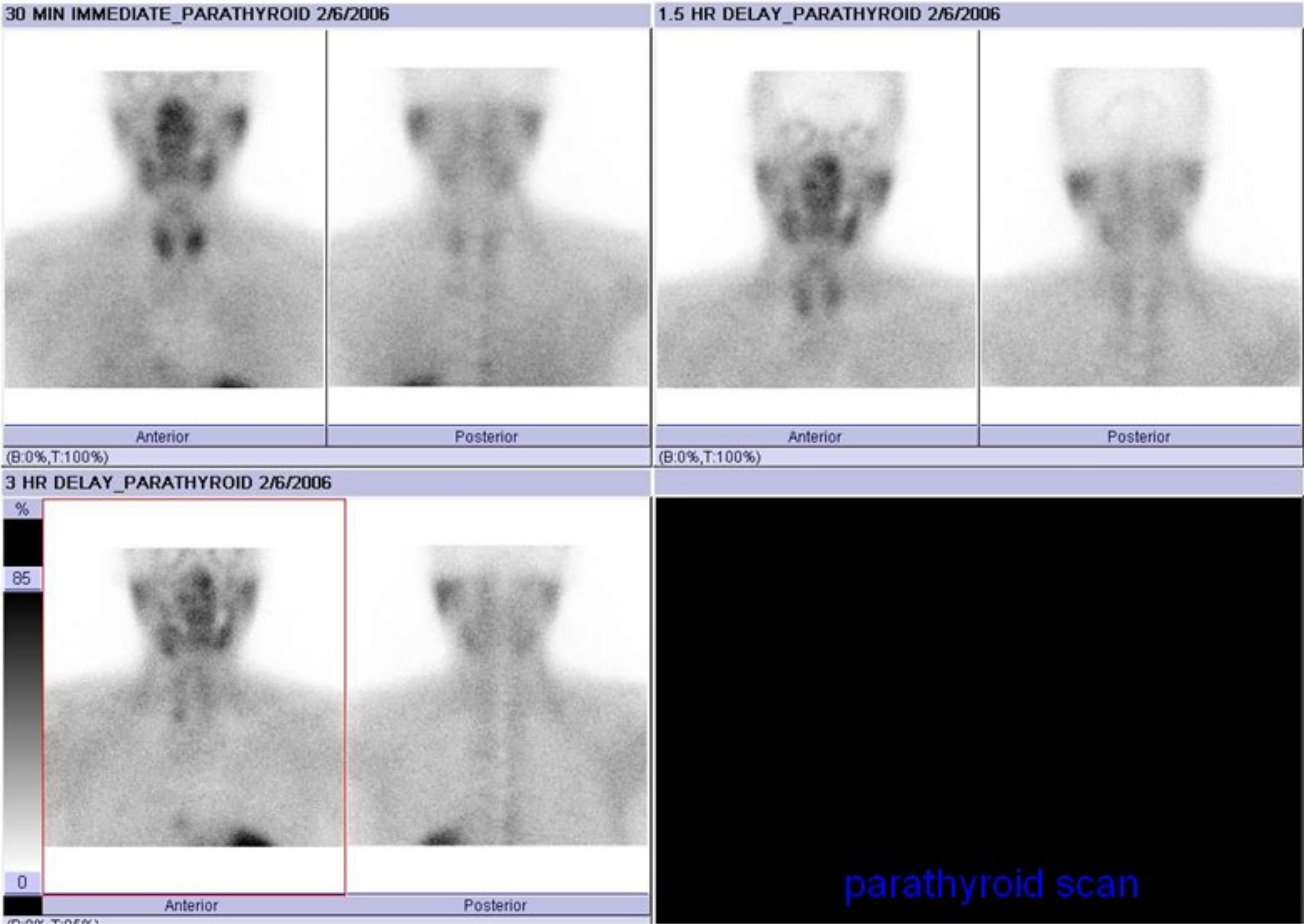
(courtesy University of Chicago)

Time based registration: breathing correction for cardiac MR *(MR courtesy HUG)*



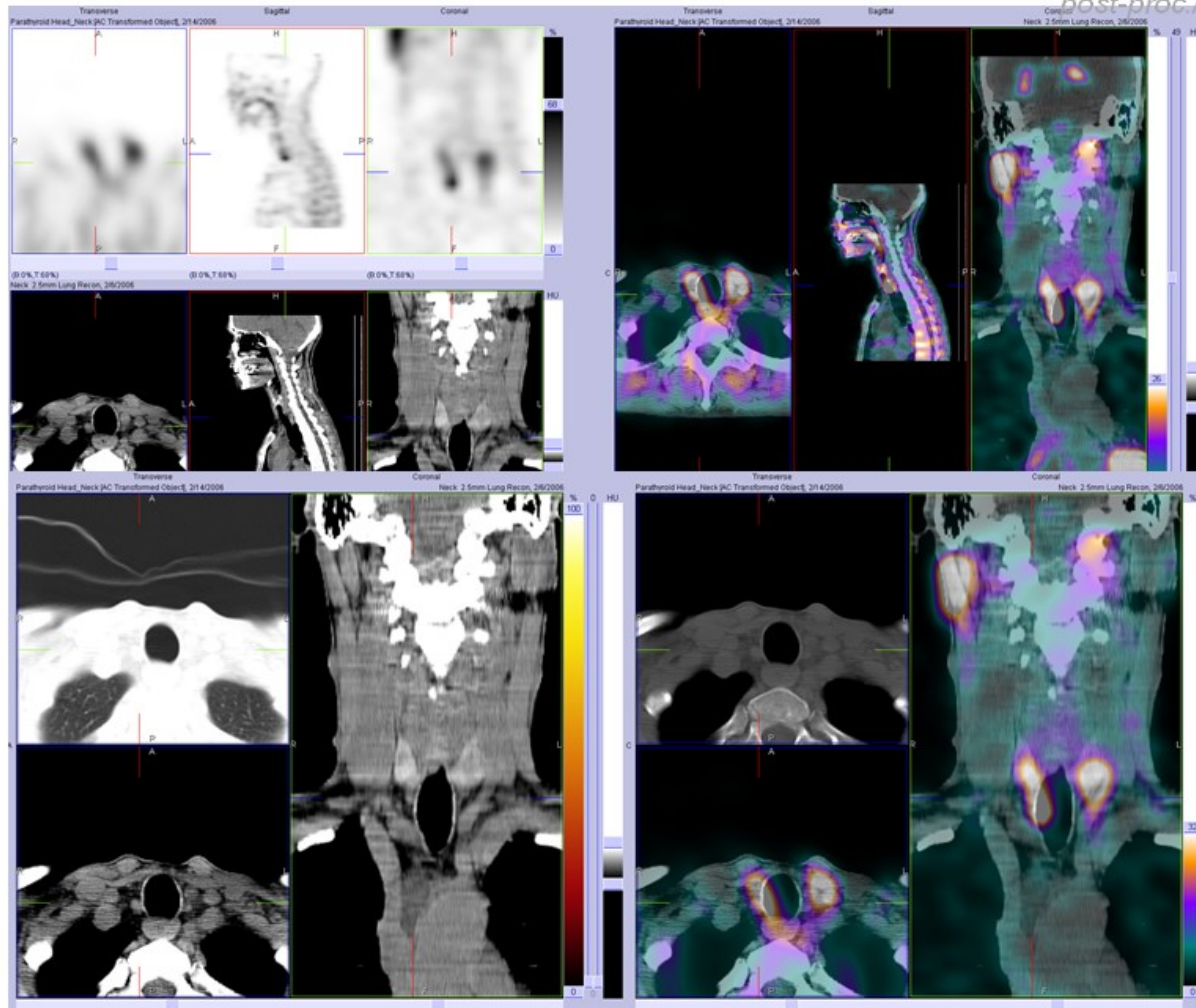
Parathyroid adenoma (KK)

(source data MDACC;
post-proc./visu. LMB)



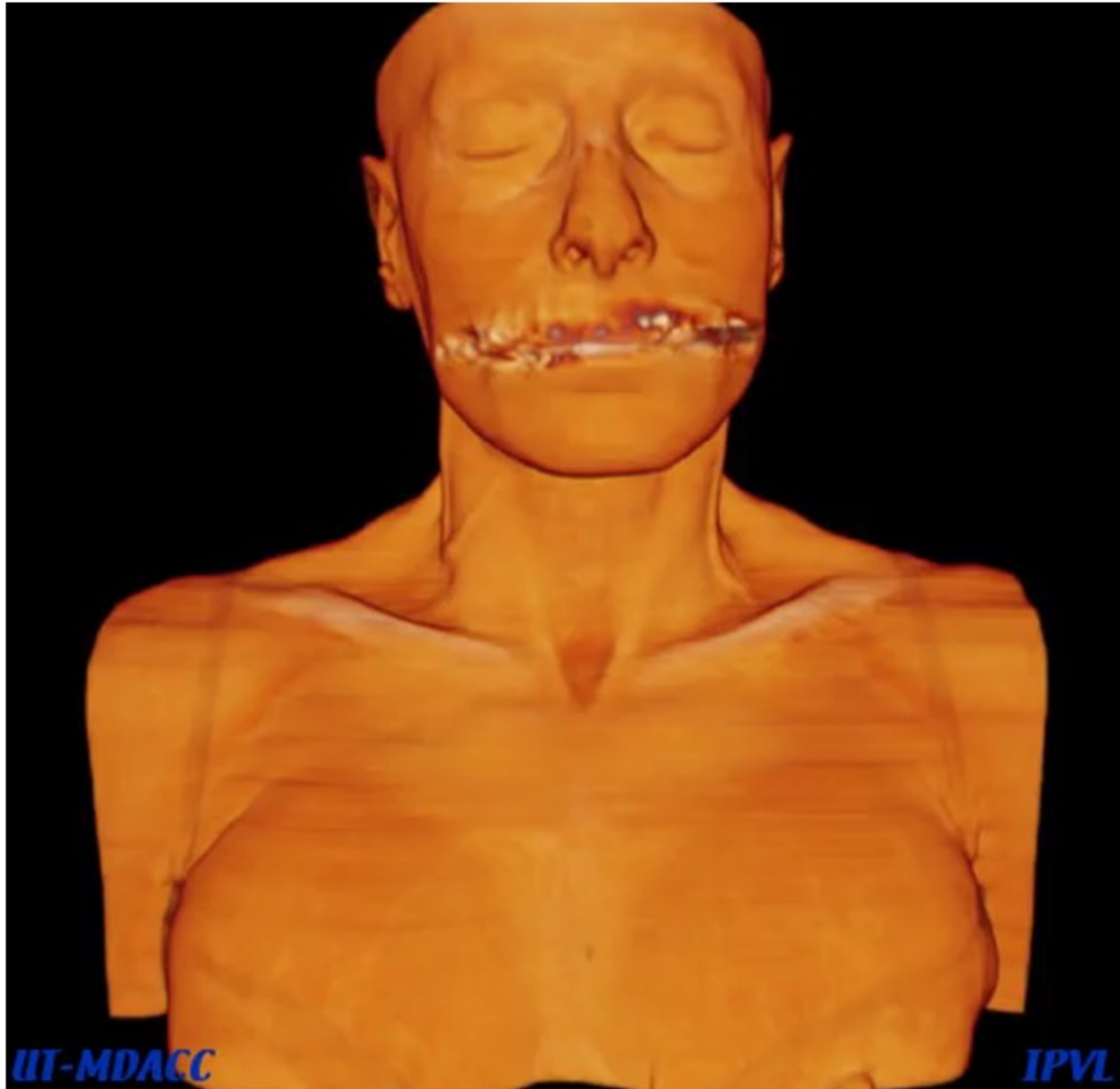
Parathyroid adenoma (KK; Tc-99M-sestamibi SPECT/CT)

(source data MDACC;
post-proc./visu. LMB)

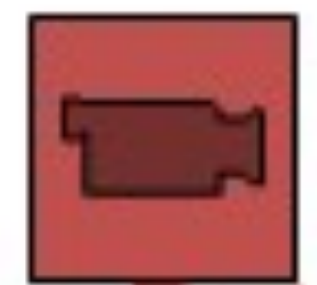


Parathyroid adenoma (KK; Tc-99M-sestamibi SPECT/CT)

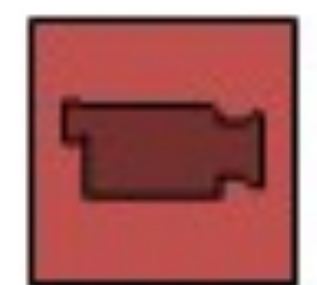
(source data MDACC;
post-proc./visu. LMB)



mip



vrRot



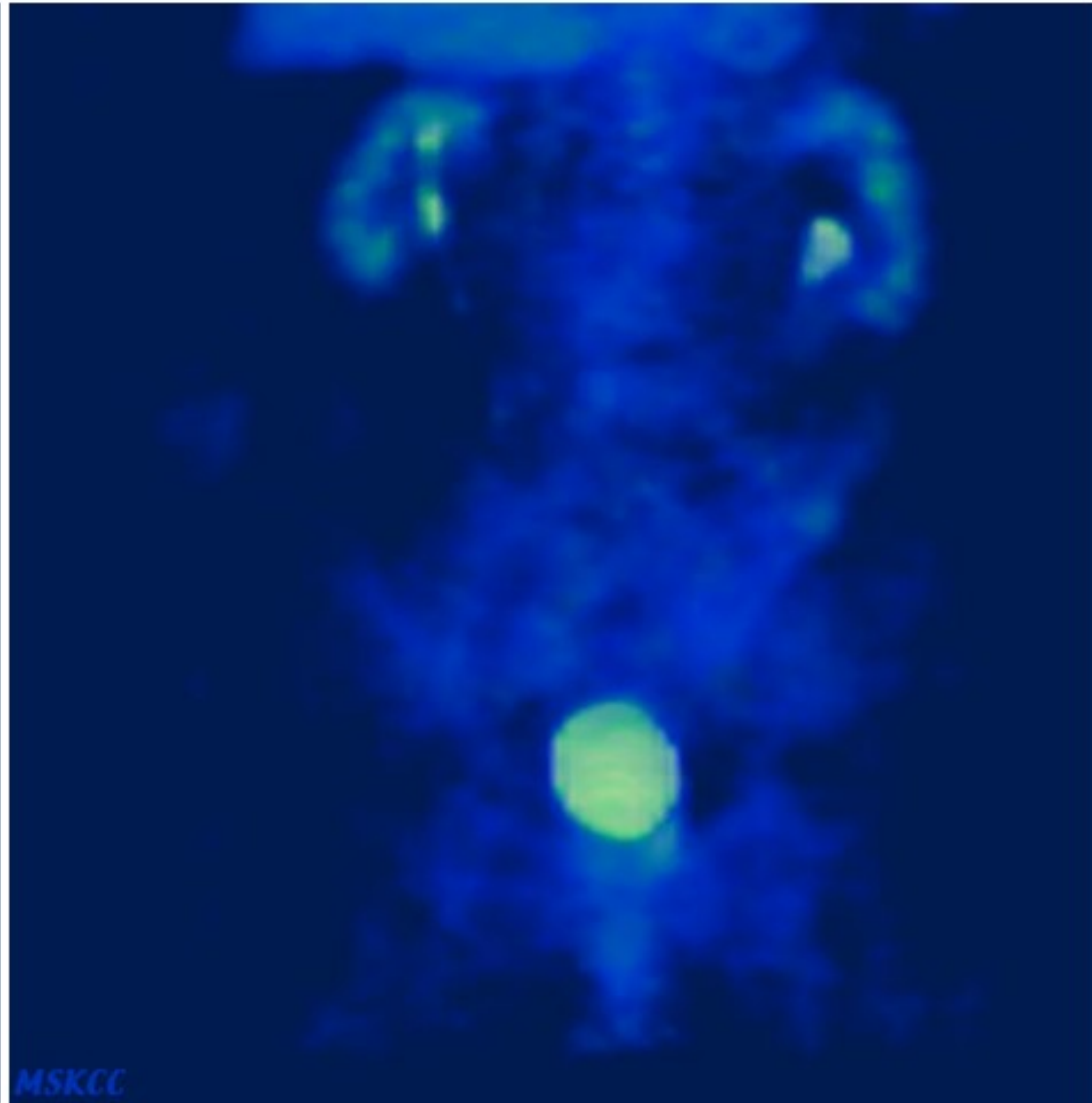
nav



PET/CT: very first combined VR (prostate cancer with FDG) (source data SKCC; post-proc./visu. LMB)



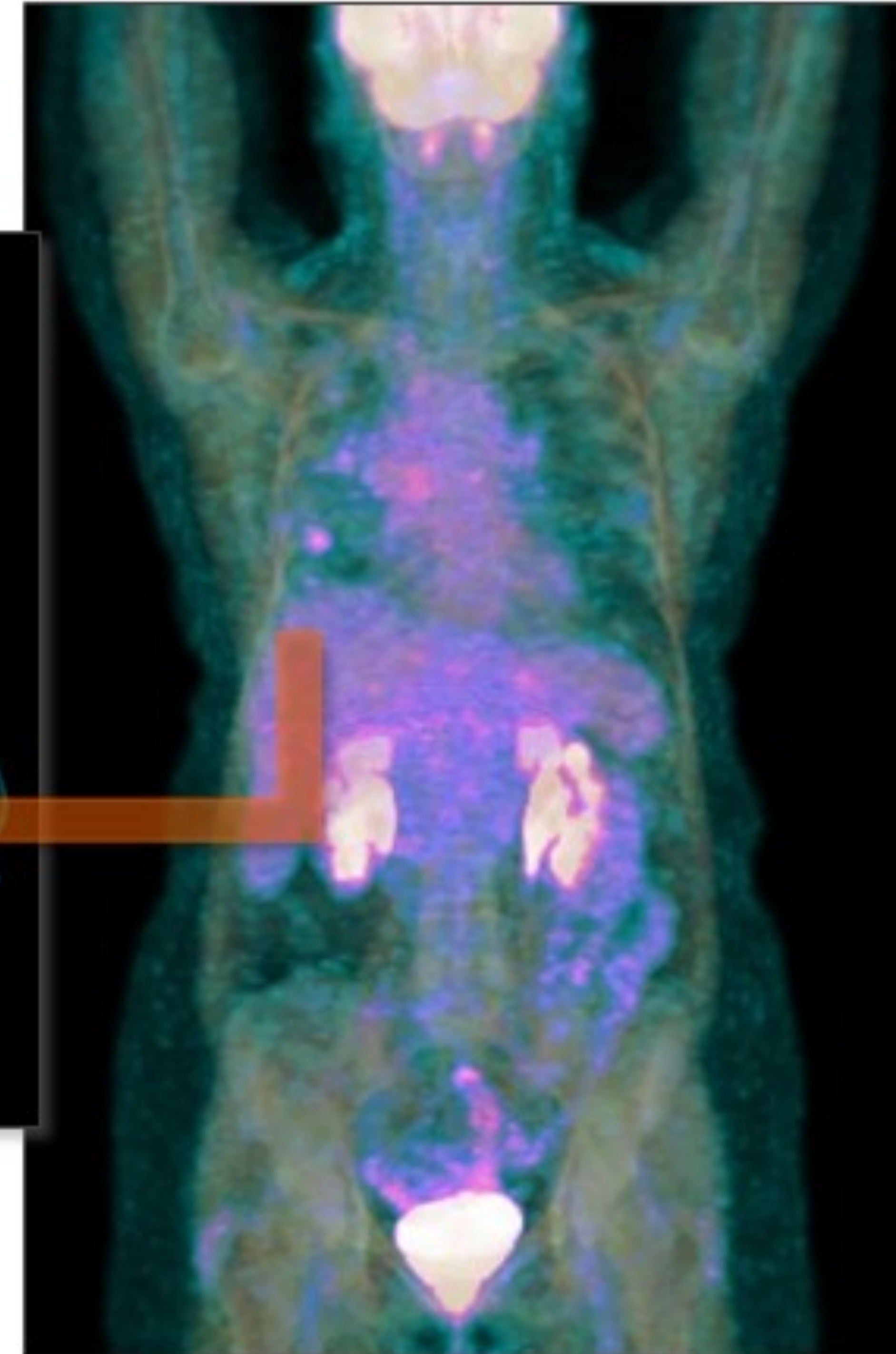
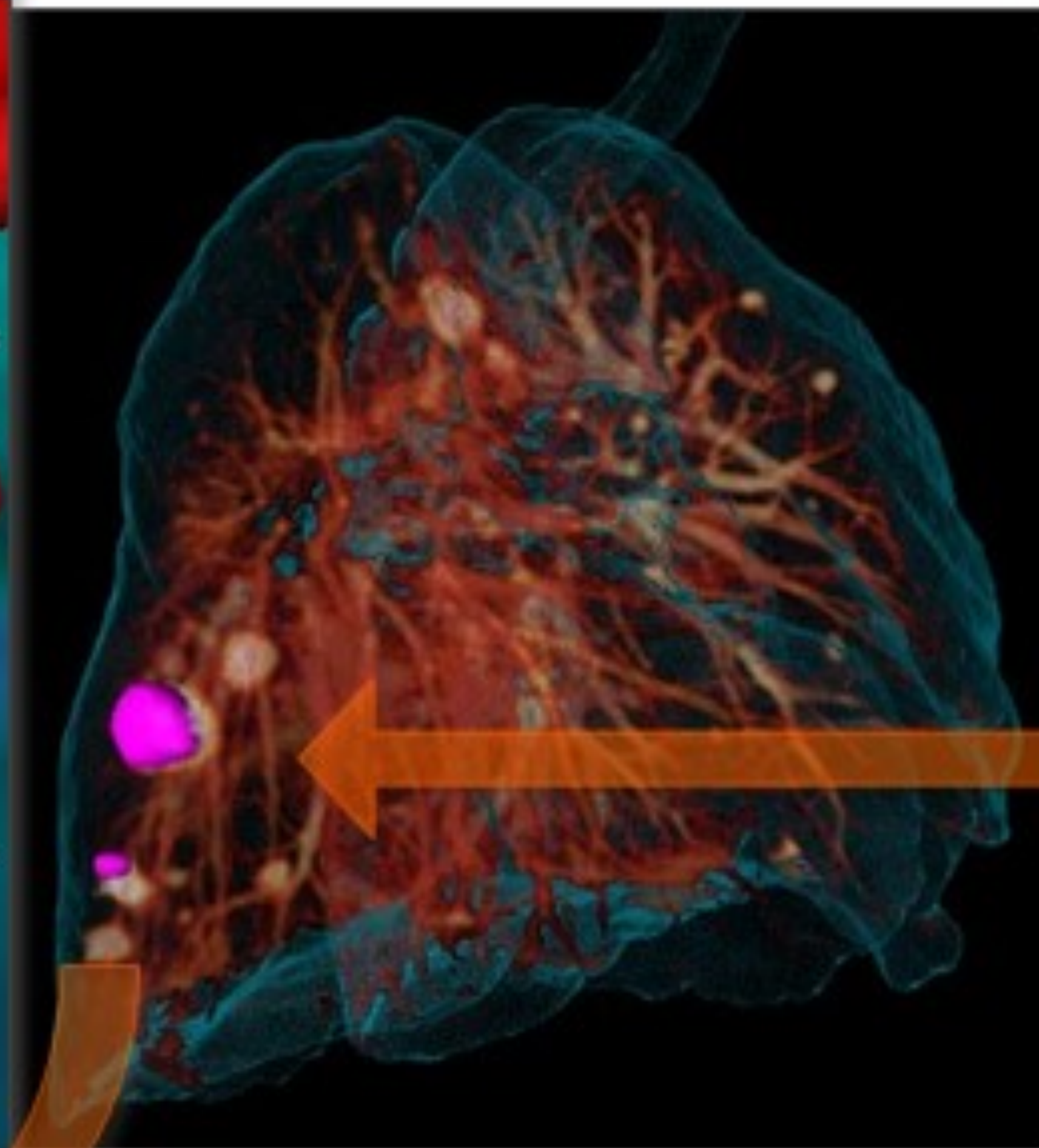
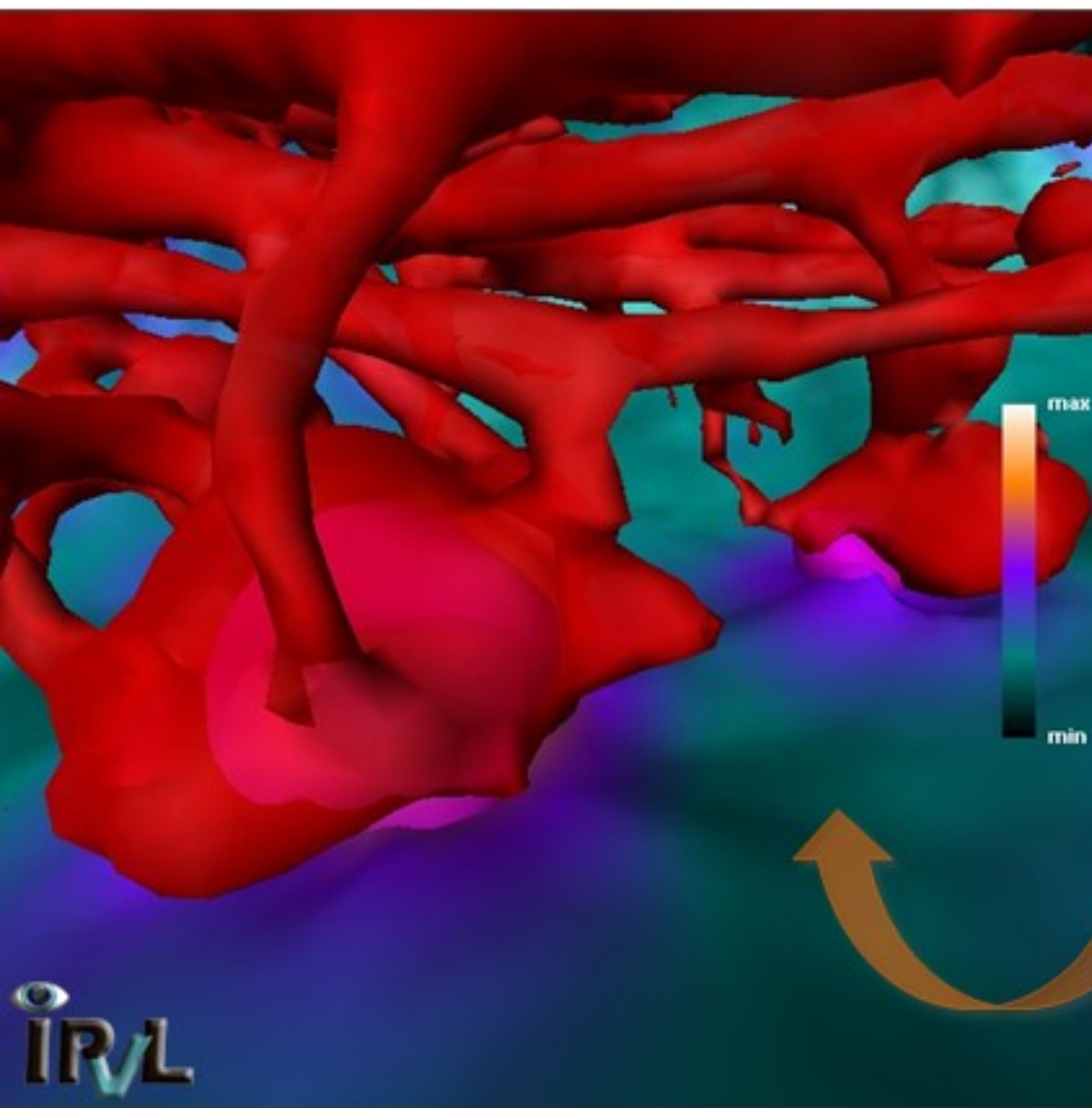
MSKCC



MSKCC

PET/CT in lung tumors

(source data MDACC;
post-proc./visu. LMB)



"...vessel co-option as a mechanism of acquired resistance to anti-angiogenic therapy and could have important implications including the potential therapeutic benefits of targeting vessel co-option in conjunction with vascular endothelial growth factor receptor signaling."

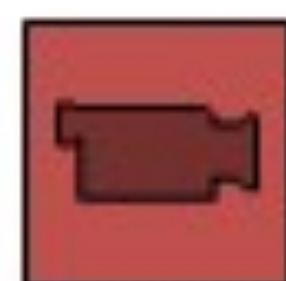
(from E.A. Kuczynski et al, J Natl Cancer Inst 108(8), 2016)



Organ (breast, etc.) specific (hybrid) imaging and fusion:

PET/CT + MR DCE

(source data MDACC;
post-proc./visu. LMB)



Suppress biopsies? (fr; osteosarcoma) - PET ; MR: T1, T1post, T2 ⇔ biopsy
(source data MDACC; post-proc./visu. LMB)

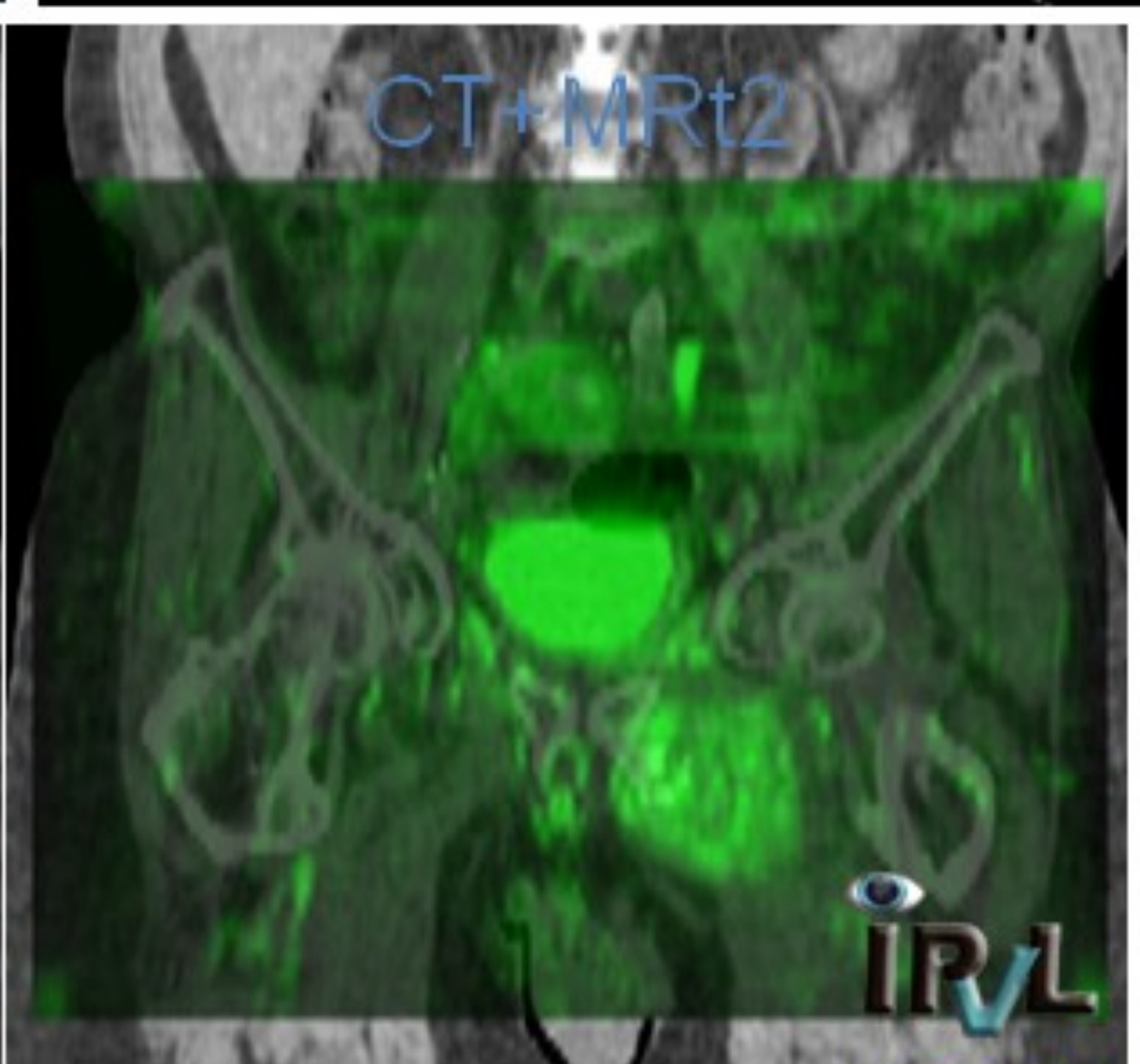
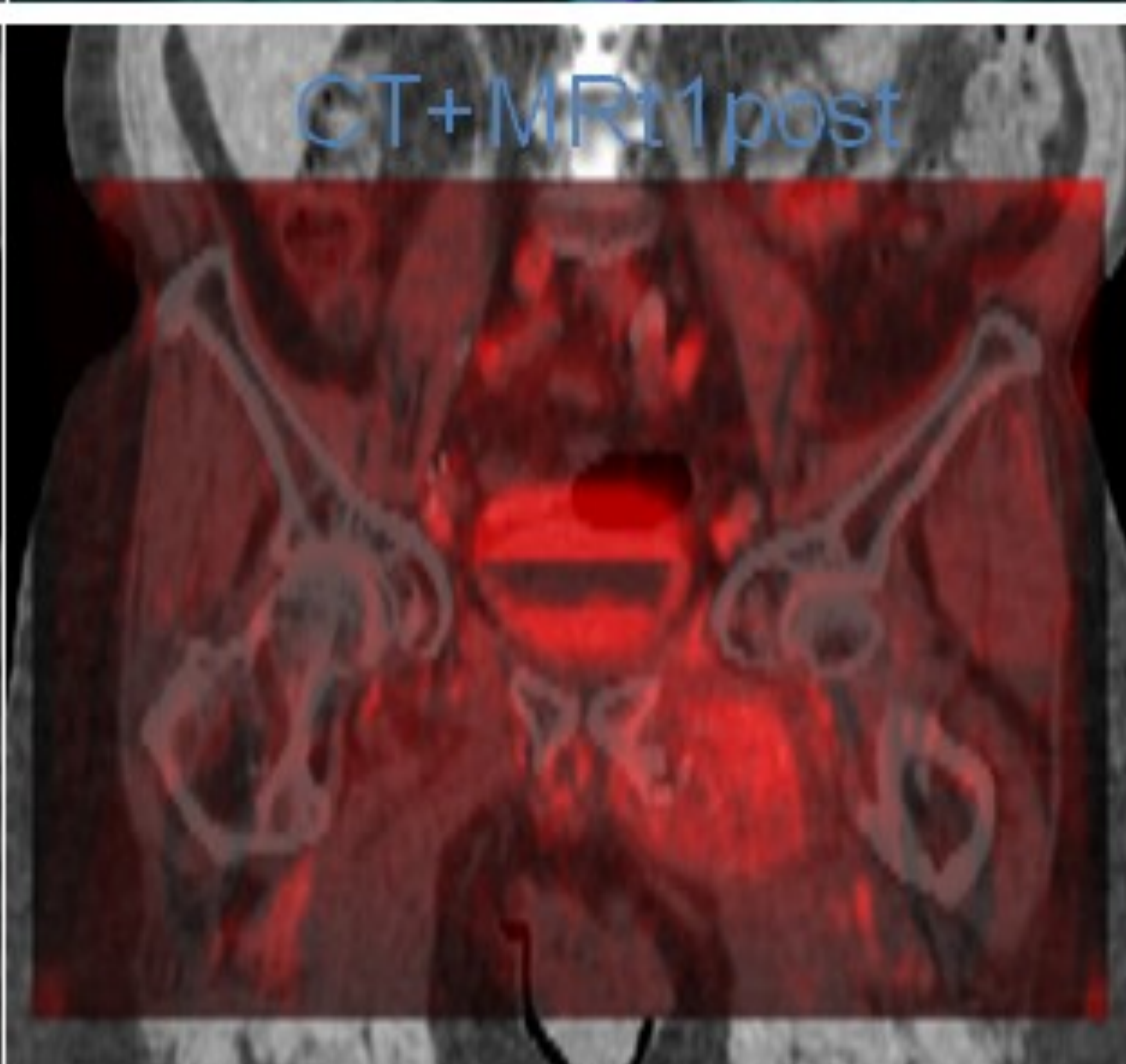
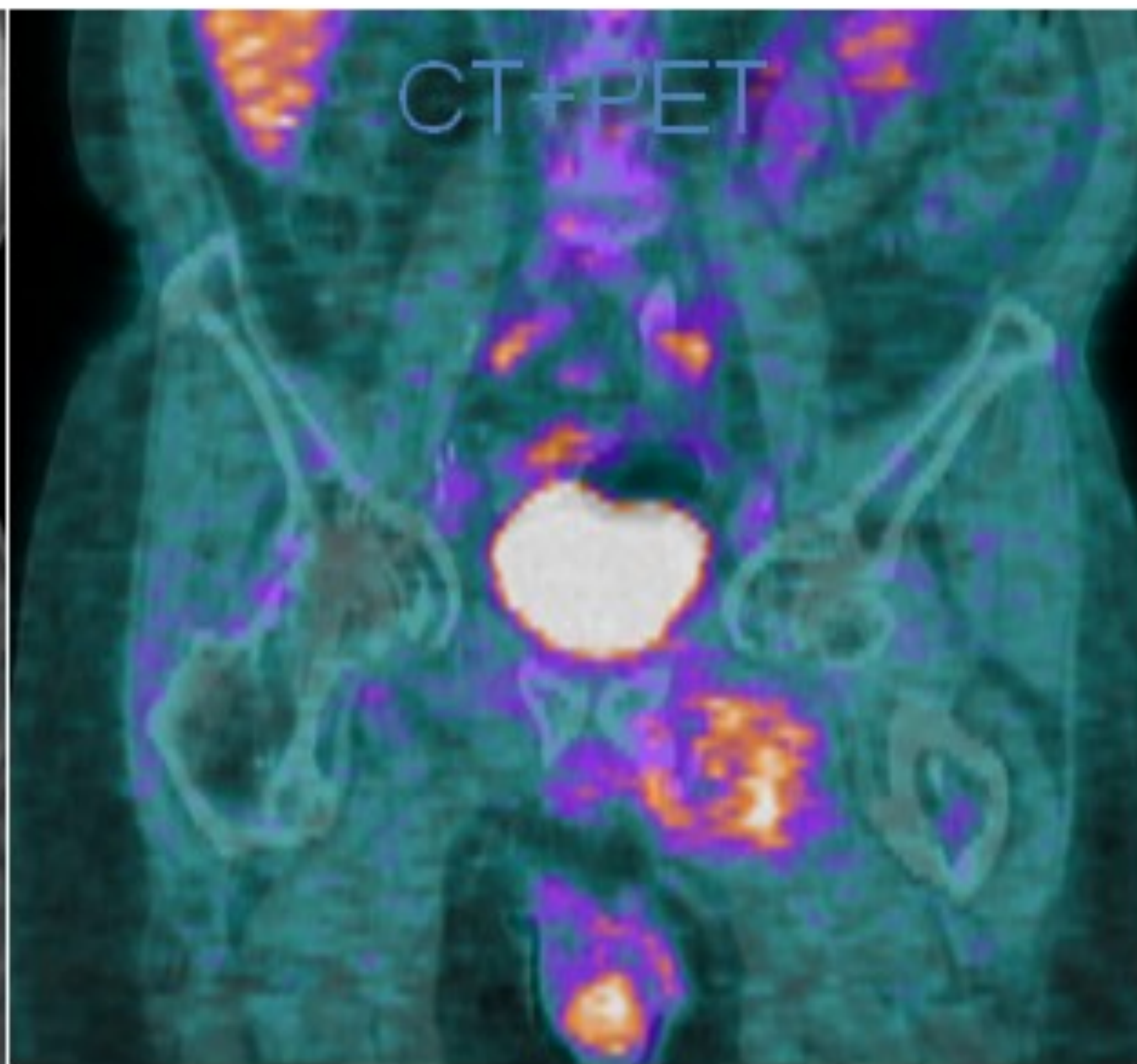
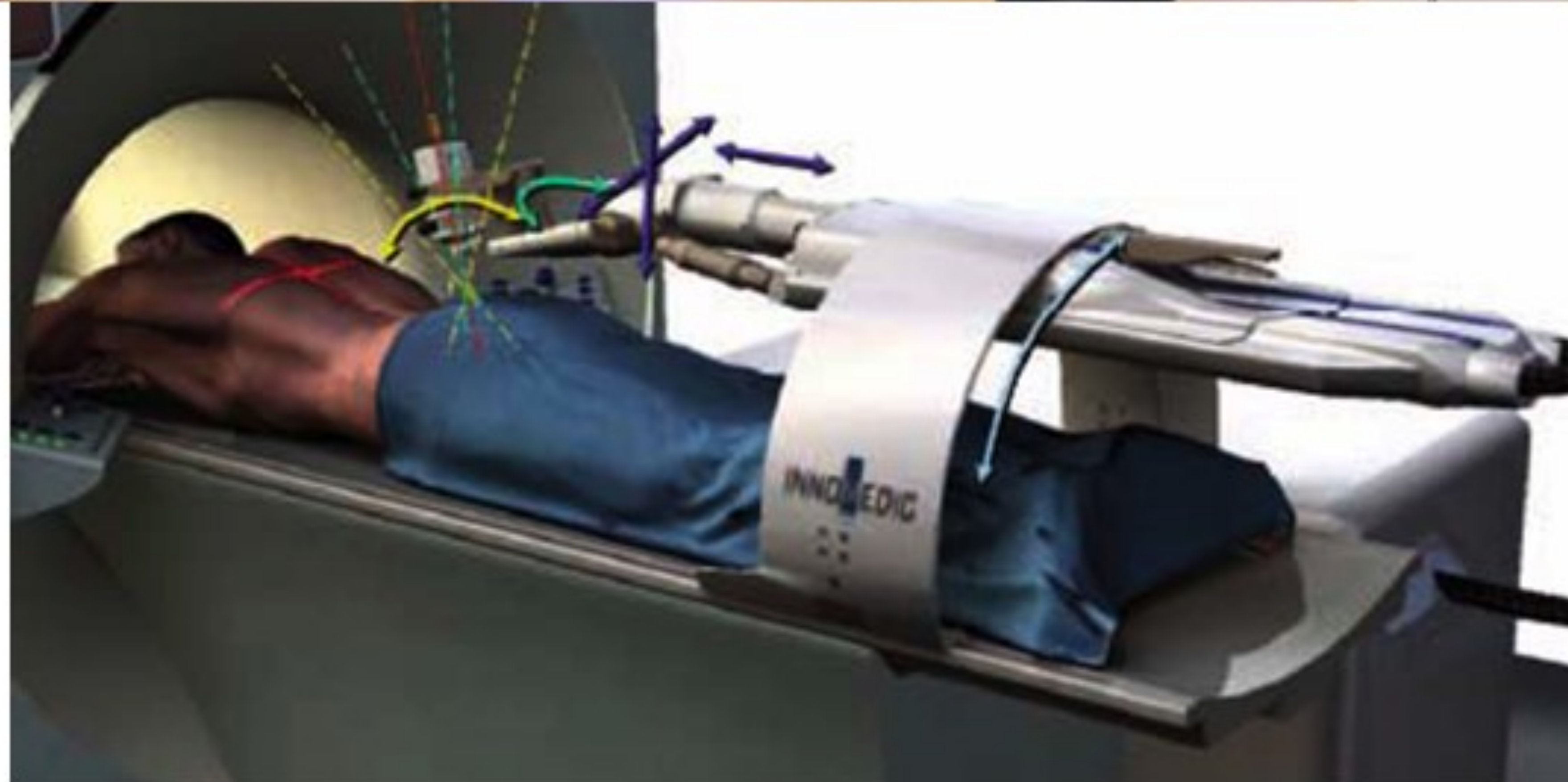


Image guidance: integrated navigation and "robots" + intra-op MM imaging + "real-time" monitoring of therapy (surgery, ablation, RT, etc) \Leftrightarrow one-stop-shops



Subjectivity...



... now ok in politics (aka "AltFacts"), but **to be avoided in clinical settings...**

*=> Imaging as a biomarker!...
can we use imaging to better characterise disease
and/or guide and assess therapy?*



Imaging biomarkers for cancer ... in a nutshell ... V2

(LMB)

Imaging + derived information

Physical, structural measurements: (CT, MR, U/S) *texture*
- "Direct": stereology, BMD, HU; DWI, ADC, DTI; elasto.
- Segmentation: shape, size (RECIST, WHO, volume), content

Physiology/function: (CT, MR, PET, SPECT)
- Motion (respiratory, cardiac, motility, etc.)
- Static uptake: contrast (CT, MR), SUV (PET) *PERCIST*
- Dynamic uptake:
 - multiphase imaging (CT, MR)
 - DCE (MR, CT)
 - PK (PET, MR)

Multimodality/sensor: (incl. time: motion and sequential/follow-up)
- CT, MR, PET, SPECT, U/S, *PA*
multi - Parametric maps

Diagnosis, characterization
Prognosis, management

Treatment planning, delivery, monitoring, follow-up

TRA...

contrast agent, tracer
transport and evaluation,
PK/PD model

*Radiomics
Fingerprinting*

*constant evolution
& innovation...*

other (meta)data,
disciplines (bio, physio, histo, chem, physics, math,...),
species/models (preclinical)

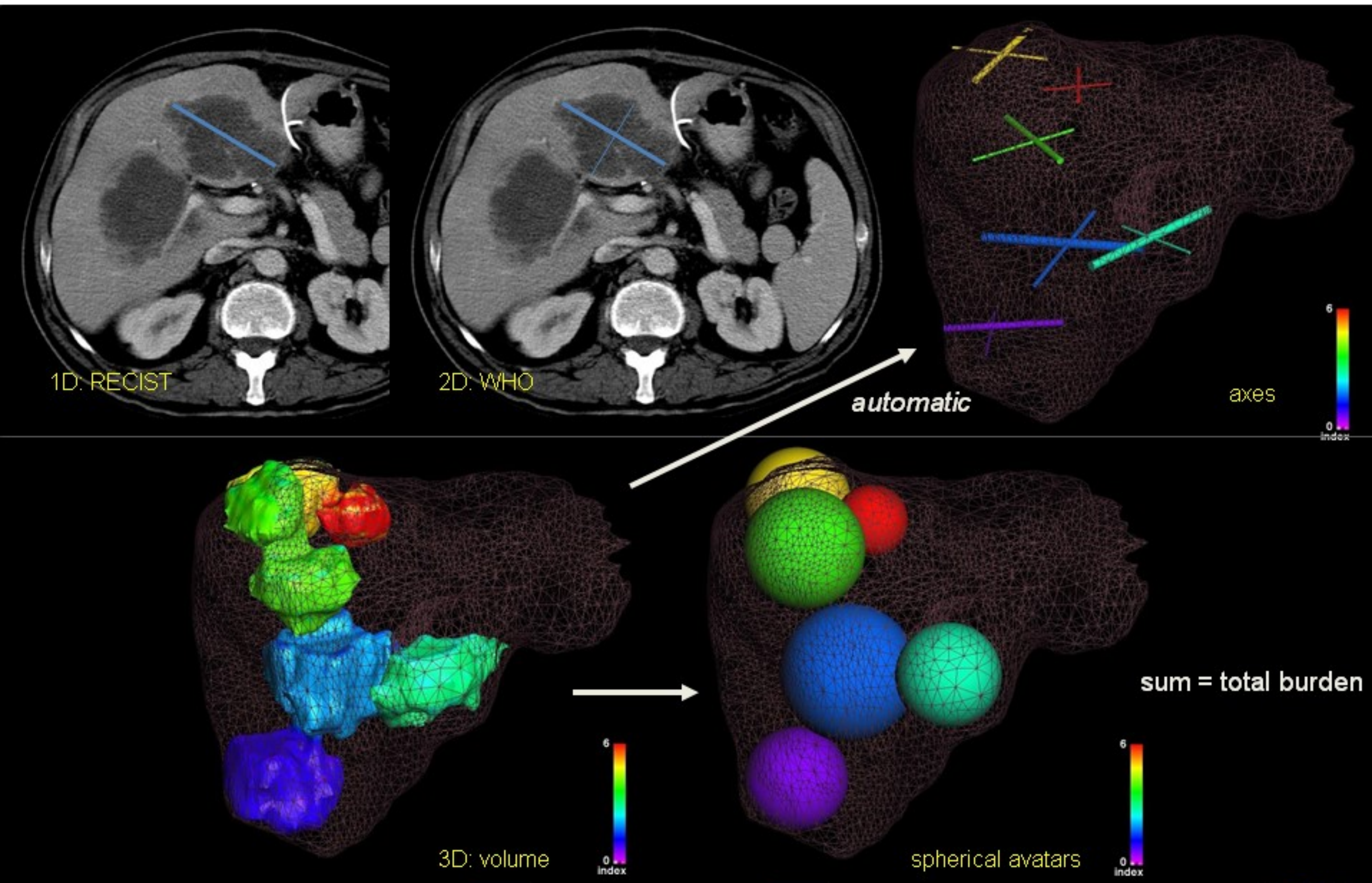
Visualisation

for every new approach and paradigm:



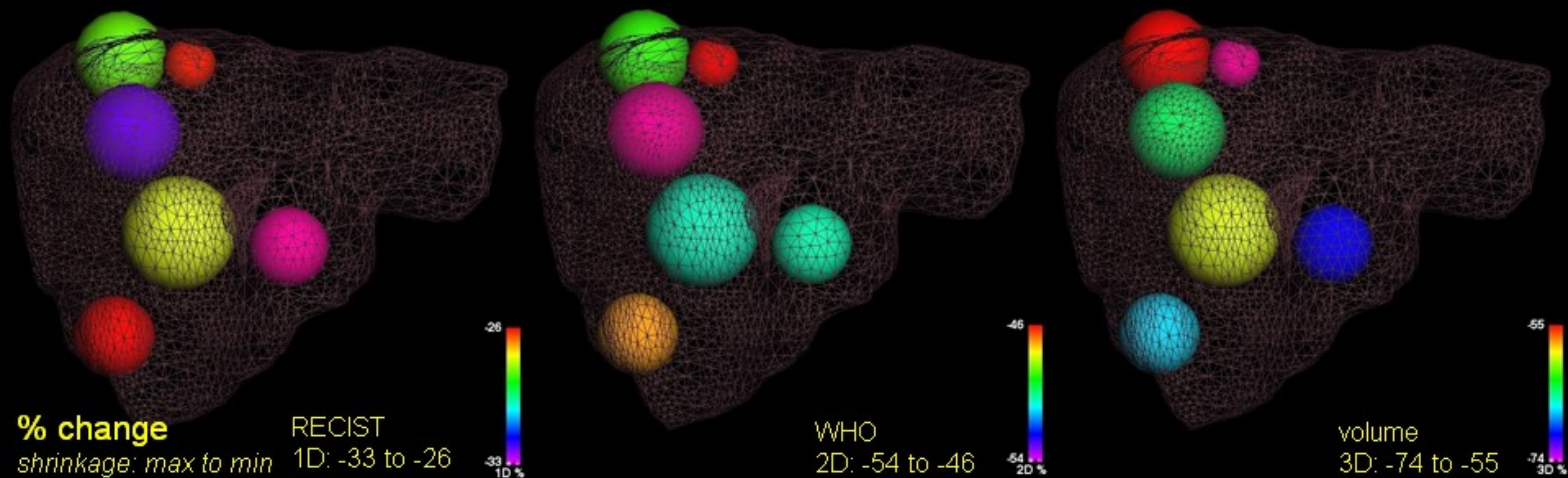
Therapy Response Assessment: morphological (baseline+follow-up) (source data MSKCC; post-proc. Visu. LMB)

(colorectal cancer + liver metastases) (RECIST: Response Evaluation Criteria In Solid Tumors)



Therapy Response Assessment: morphological (baseline+follow-up) (source data MSKCC; post-proc. Visu. LMB)

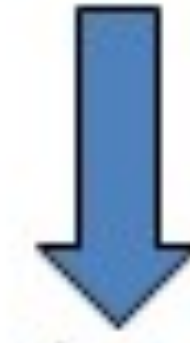
(colorectal cancer + liver metastases) (*RECIST: Response Evaluation Criteria In Solid Tumors*)



Therapy response: functional (breast cancer mets and chemoT)

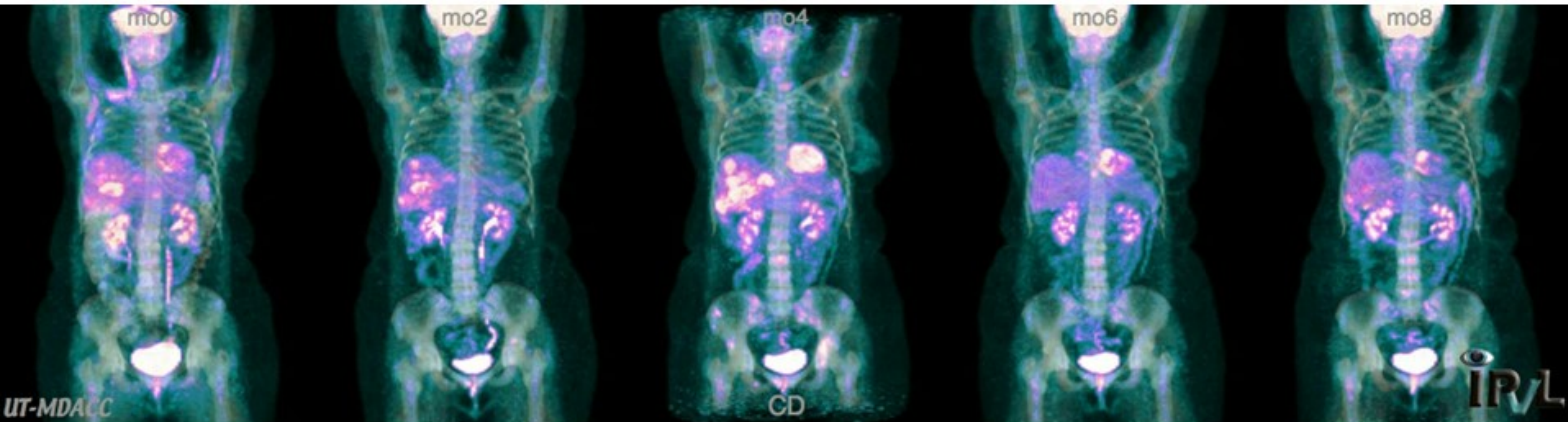
(source data MDACC;
post-proc./visu. LMB)

no response
=> change of regimen



therapy A

therapy B



time + therapy



$$\text{SUV} = \frac{\text{mean activity within a ROI (mCi/ml)}}{\text{injected dose (mCi)/body mass (kg)}}$$

variations: use *body surface area* or *lean body (mass)weight*

SUV is affected by:

- plasma glucose levels
- time between administration and acquisition
- body weight or surface area
- size of ROI
- resolution of scanner and patient motion
(\Leftrightarrow blurring and PVE)
- actual definition

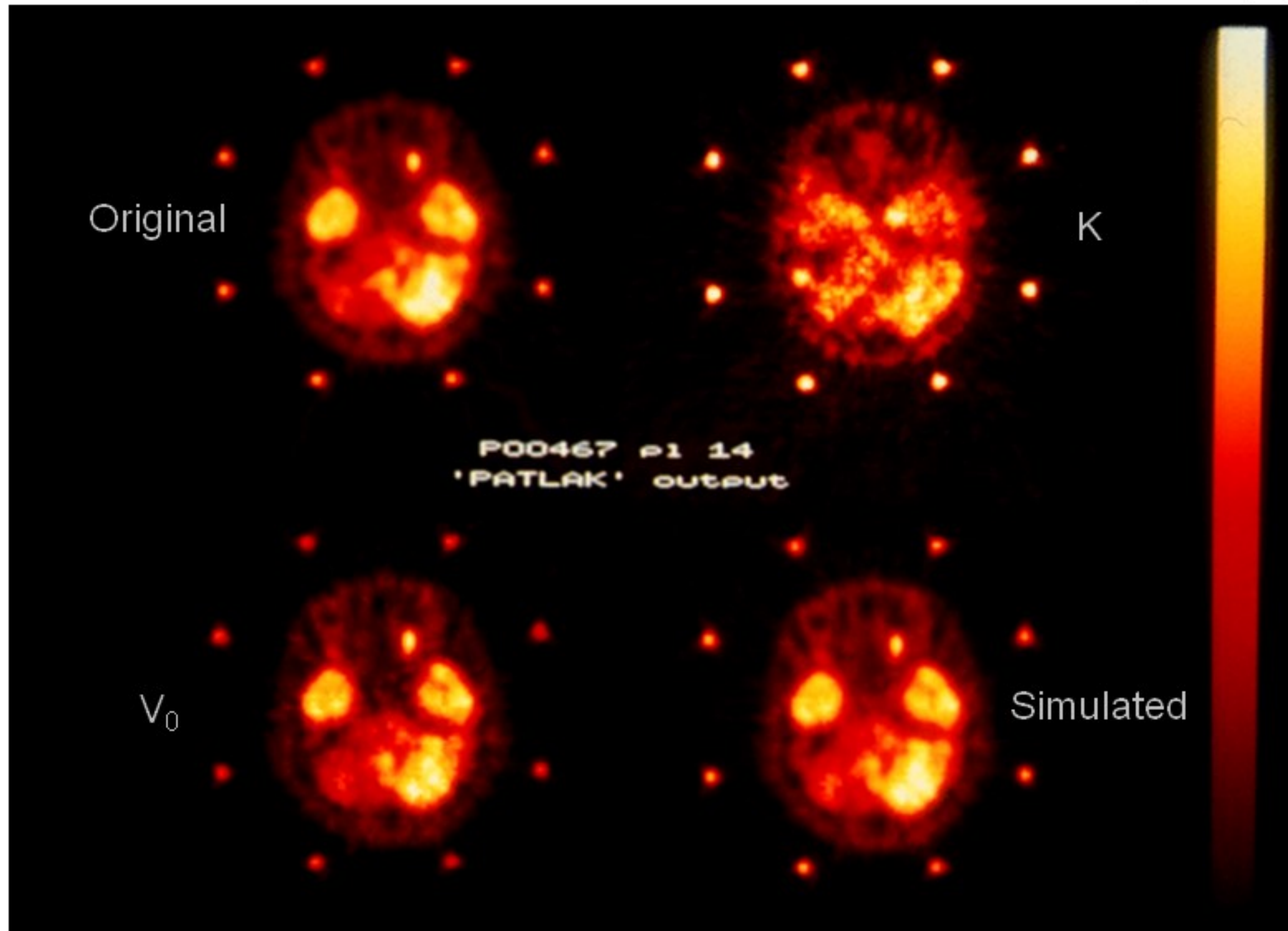
↓
PERCIST

(cf Richard Wahl, 2009)

SUV < 2.0 \approx benign lesion
> 2.0 \approx malignant lesion
(2.0 \approx blood pool activity)



QAI: parametric imaging - FDG PET (Patlak: $R(t) = K \int_0^t C_p(\tau) d\tau + V_0 C_p(t)$)
 $k_4=0 \Rightarrow K \approx k_2 k_3 / (k_2 + k_3)$



but still seldom implemented in the clinic (logistics, throughput, ...)

Radiomics: definition, principle

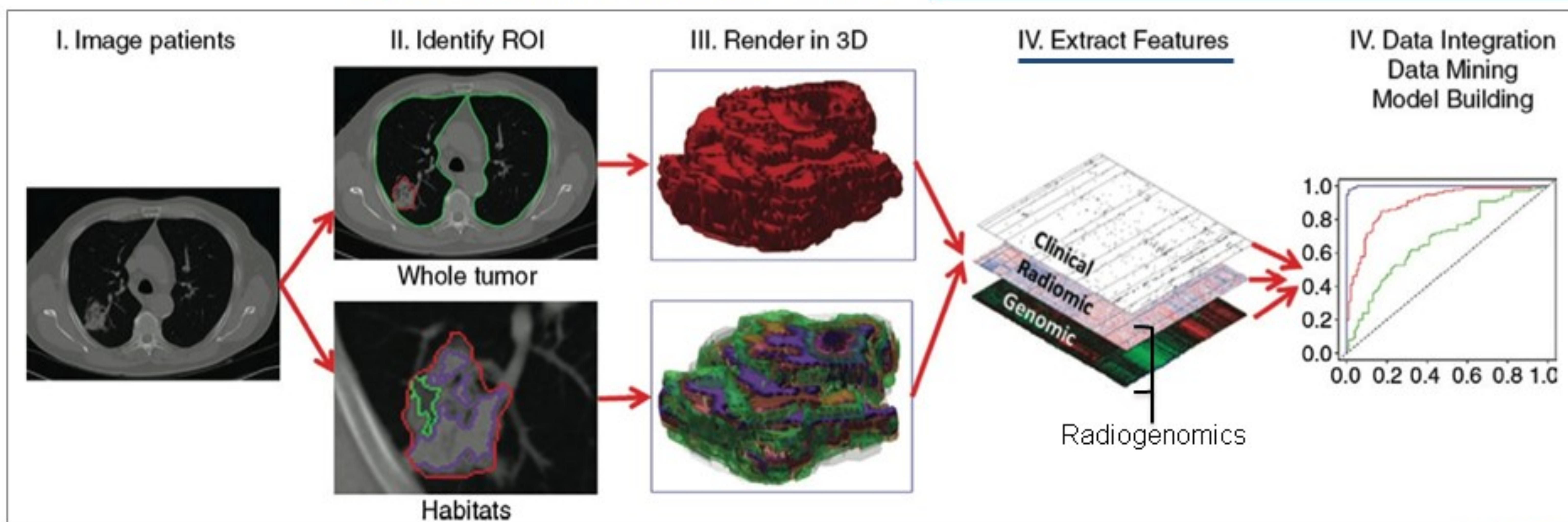
(from R. Gillies et al, Radiol 278(2), 2015)

“Radiomics” = extraction and analysis of large amounts of **advanced quantitative imaging features** from medical images, including standard-of-care images ⇔ very large potential subject pool.

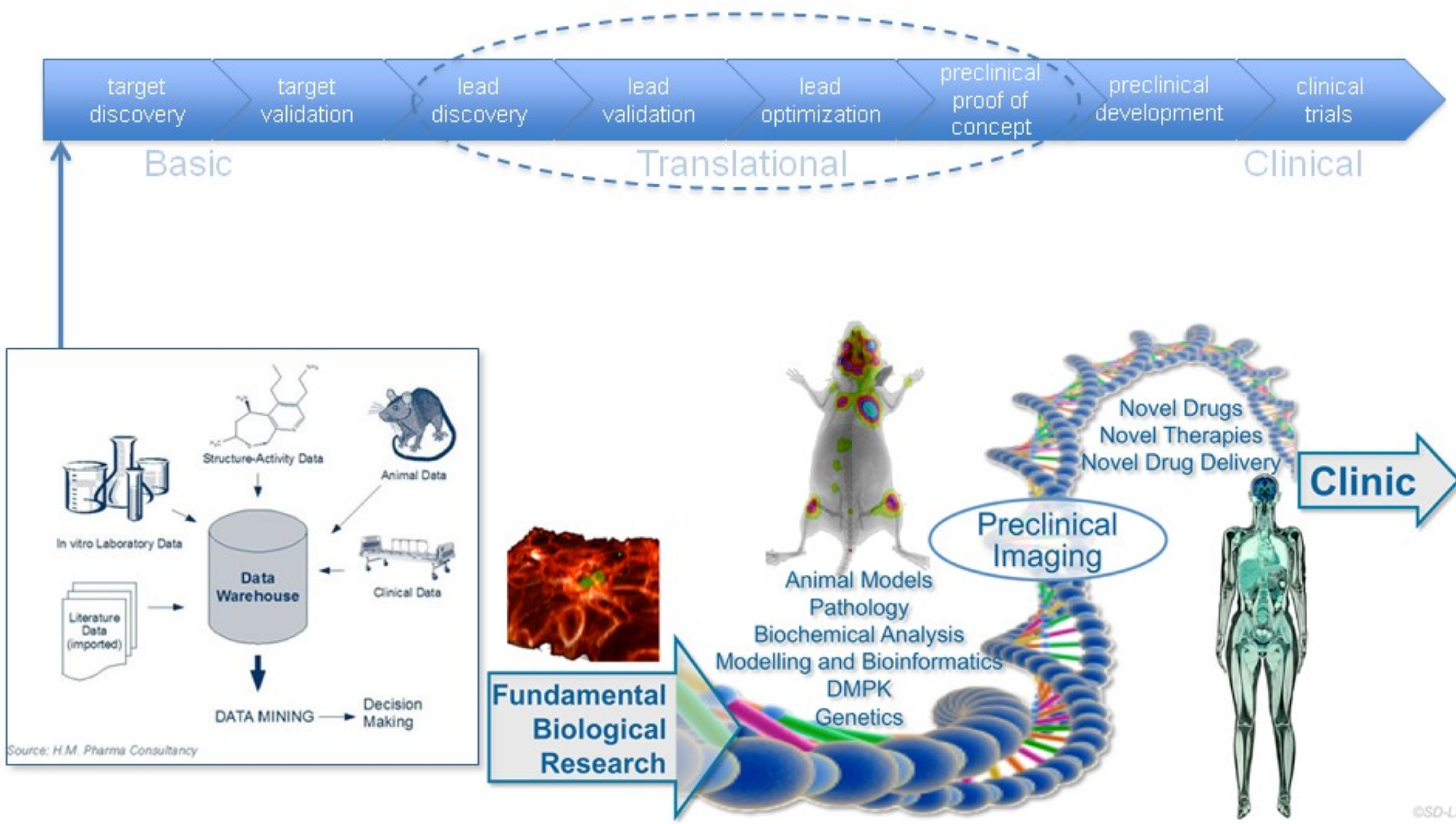
Radiomics data are in a mineable form that can be used to build **descriptive and predictive models** relating image features to **phenotypes** or gene–protein signatures. The core hypothesis of radiomics is that these models, which can include **biological or medical data**, can provide valuable **diagnostic, prognostic or predictive information**. (adapted from V. Kumar et al, 2012)

- **semantic**: through radiologist's visual assessment
- **agnostic**: mathematically derived

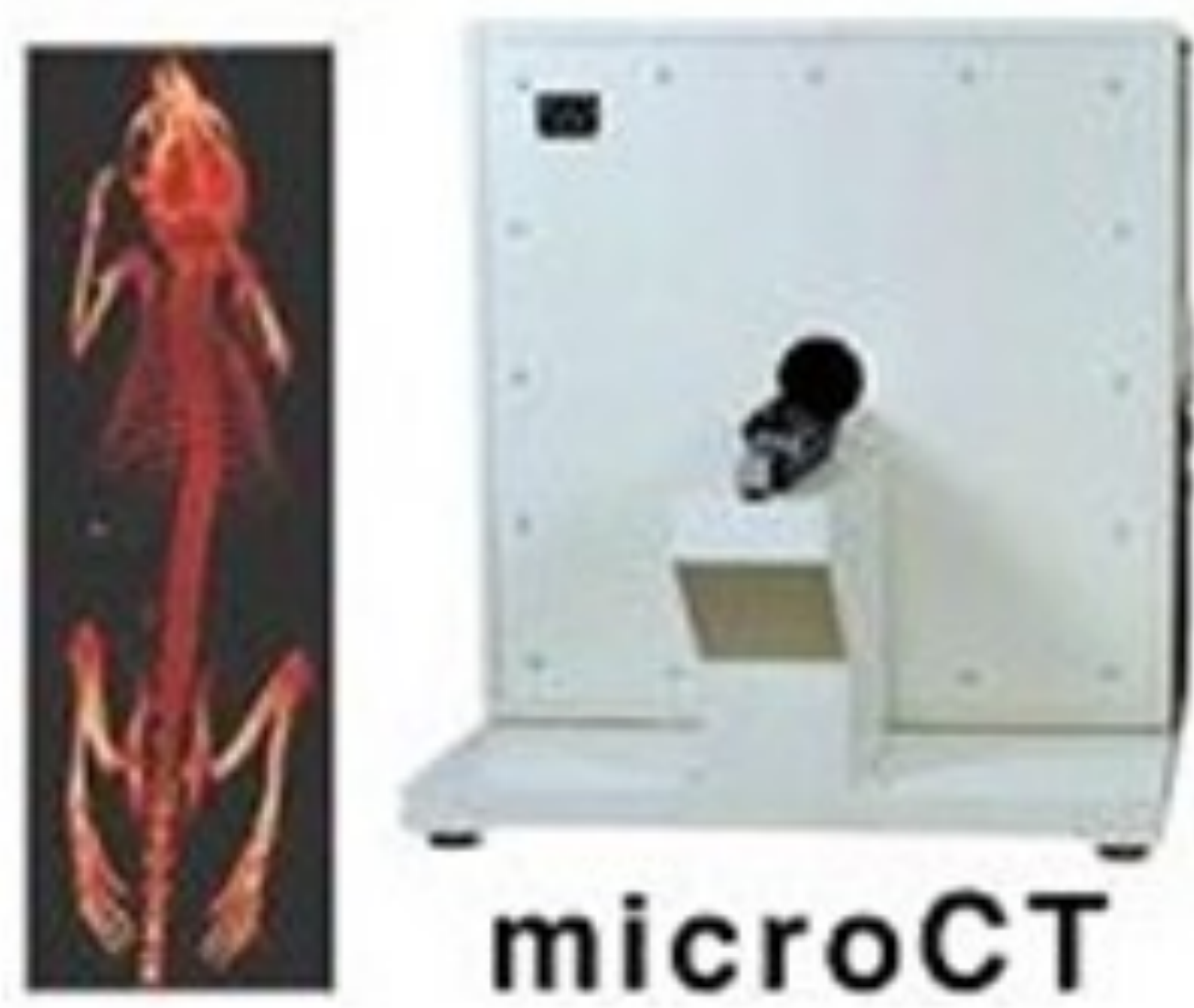
Semantic	Agnostic
Size	Histogram (skewness, kurtosis)
Shape	Haralick textures
Location	Laws textures
Vascularity	Wavelets
Spiculation	Laplacian transforms
Necrosis	Minkowski functionals
Attachments or lepidics	Fractal dimensions



Biomedical translation: the case of drug discovery and development...



MM (structural and molecular) animal imaging

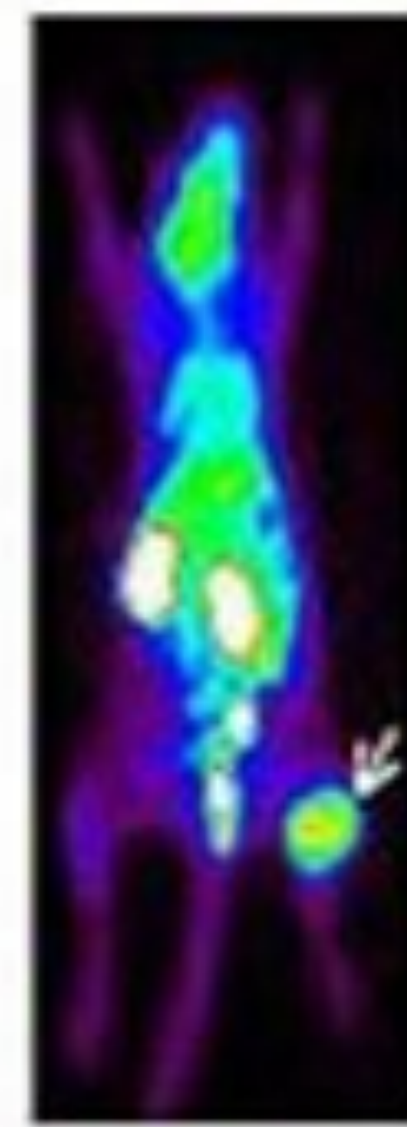


microCT

Animal MRI



microPET



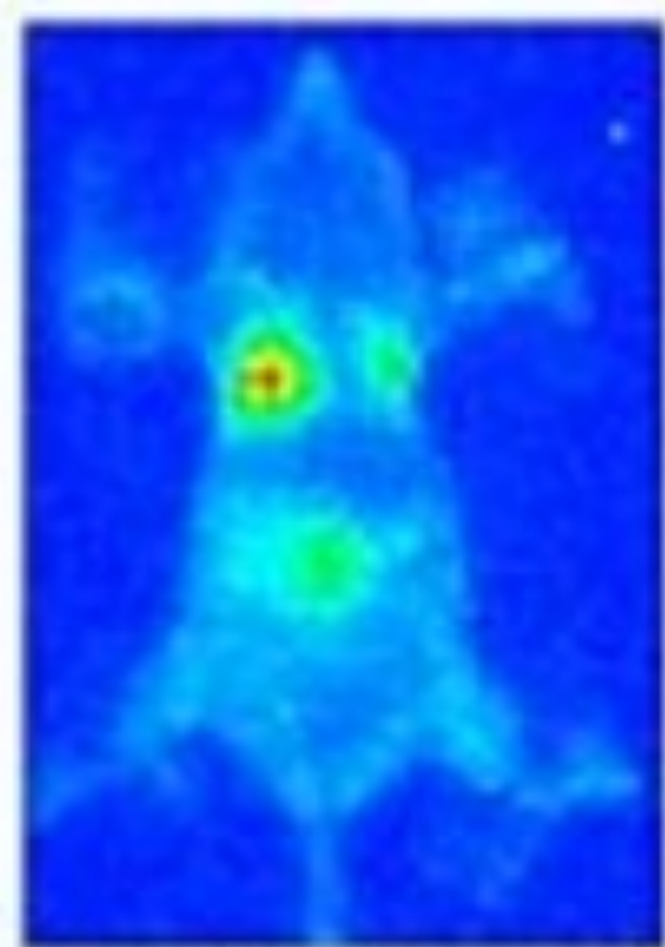
microSPECT



(courtesy CIMI)



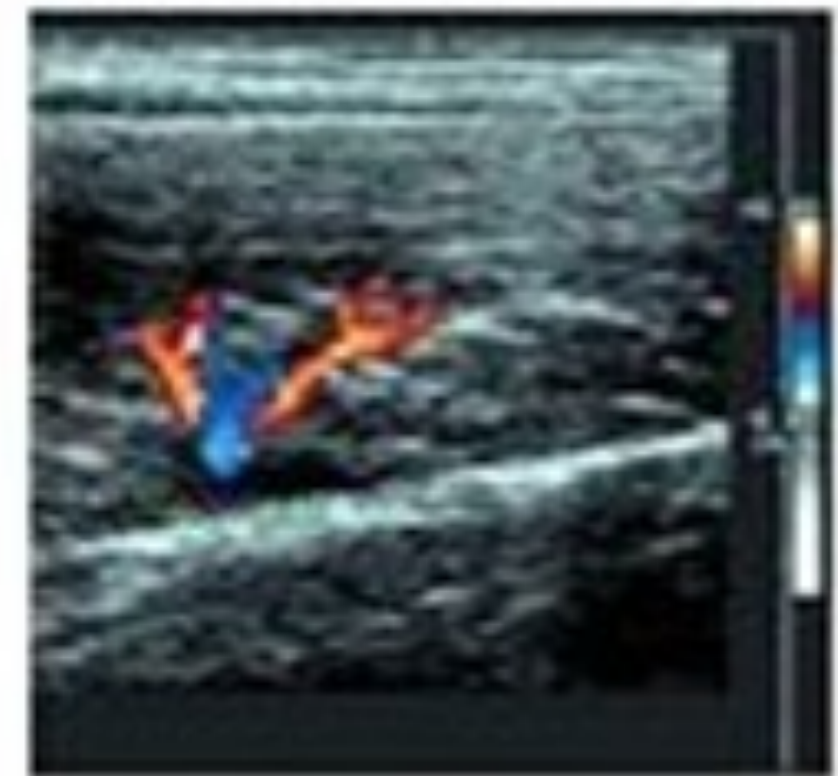
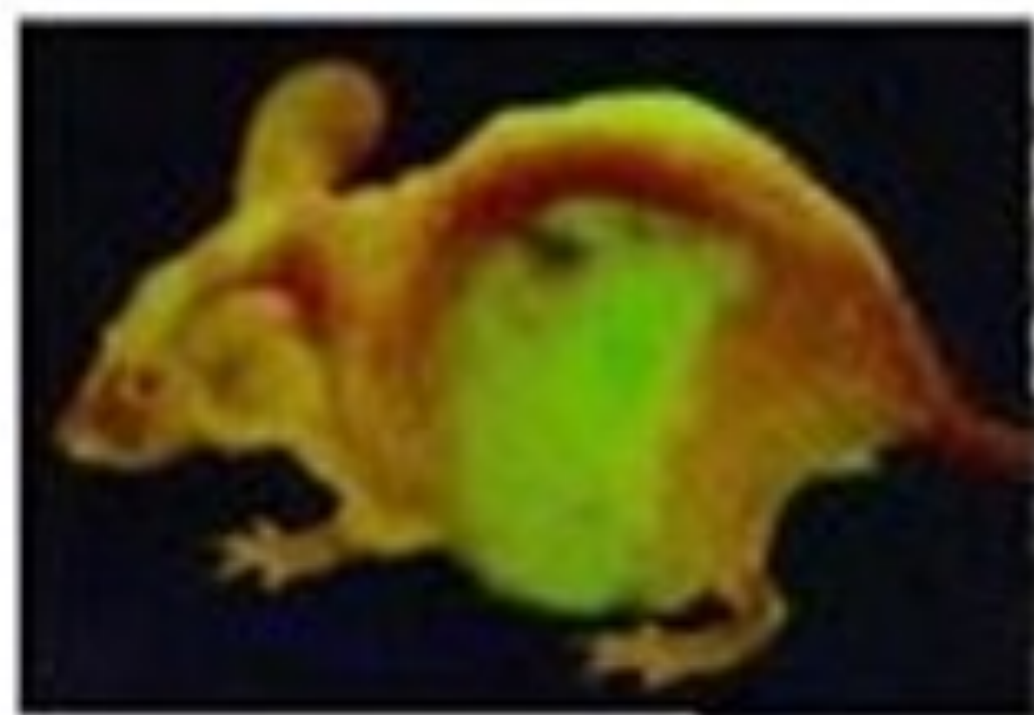
Autoradiography



Bioluminescence



Fluorescence



Ultrasound

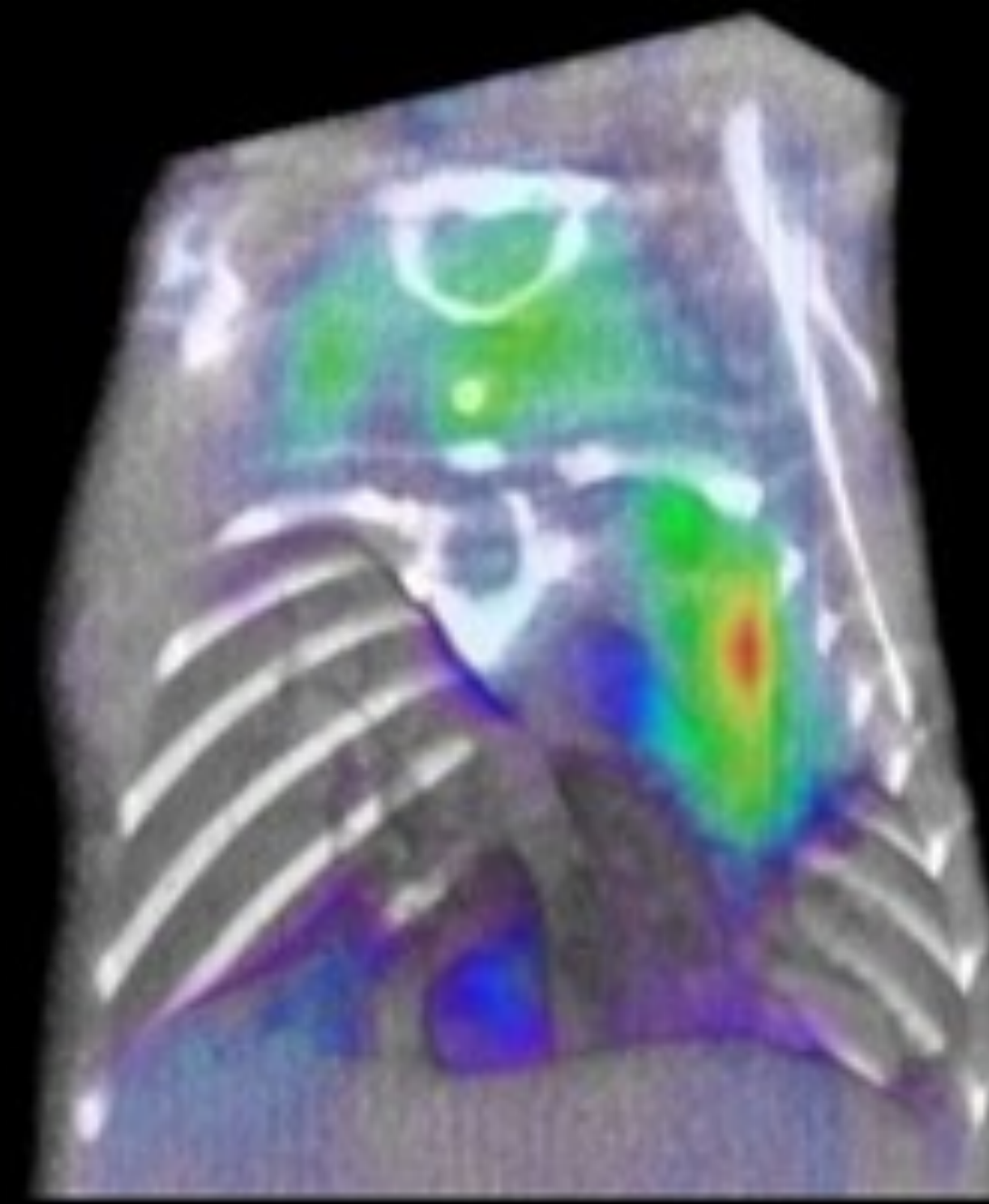
(+ PhotoAcoustic, etc)



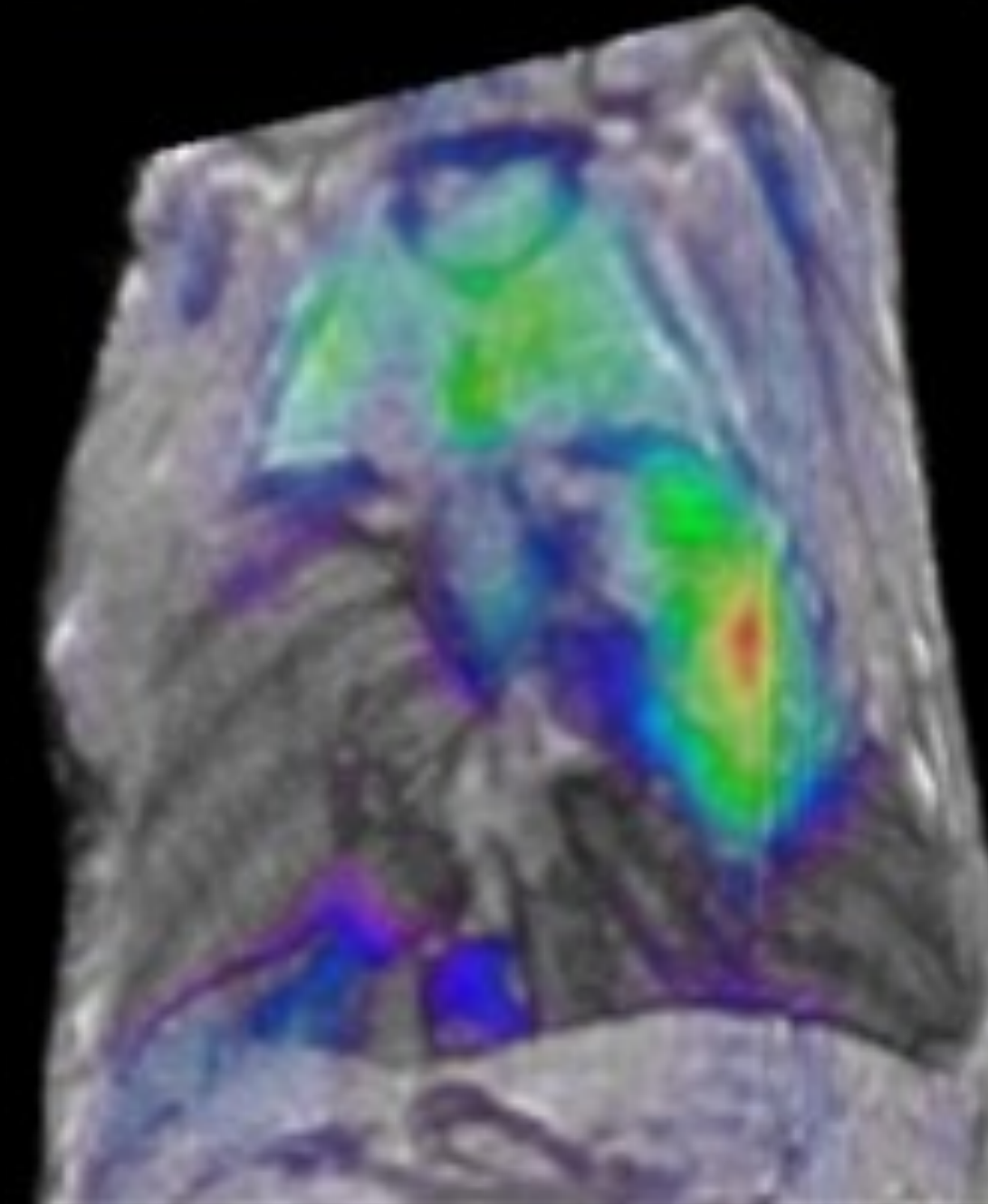
μ MM (CT, MR and PET) of a mouse lung tumor

(source data + collab. J. Kurie, MDACC;
post-proc./visu. LMB)

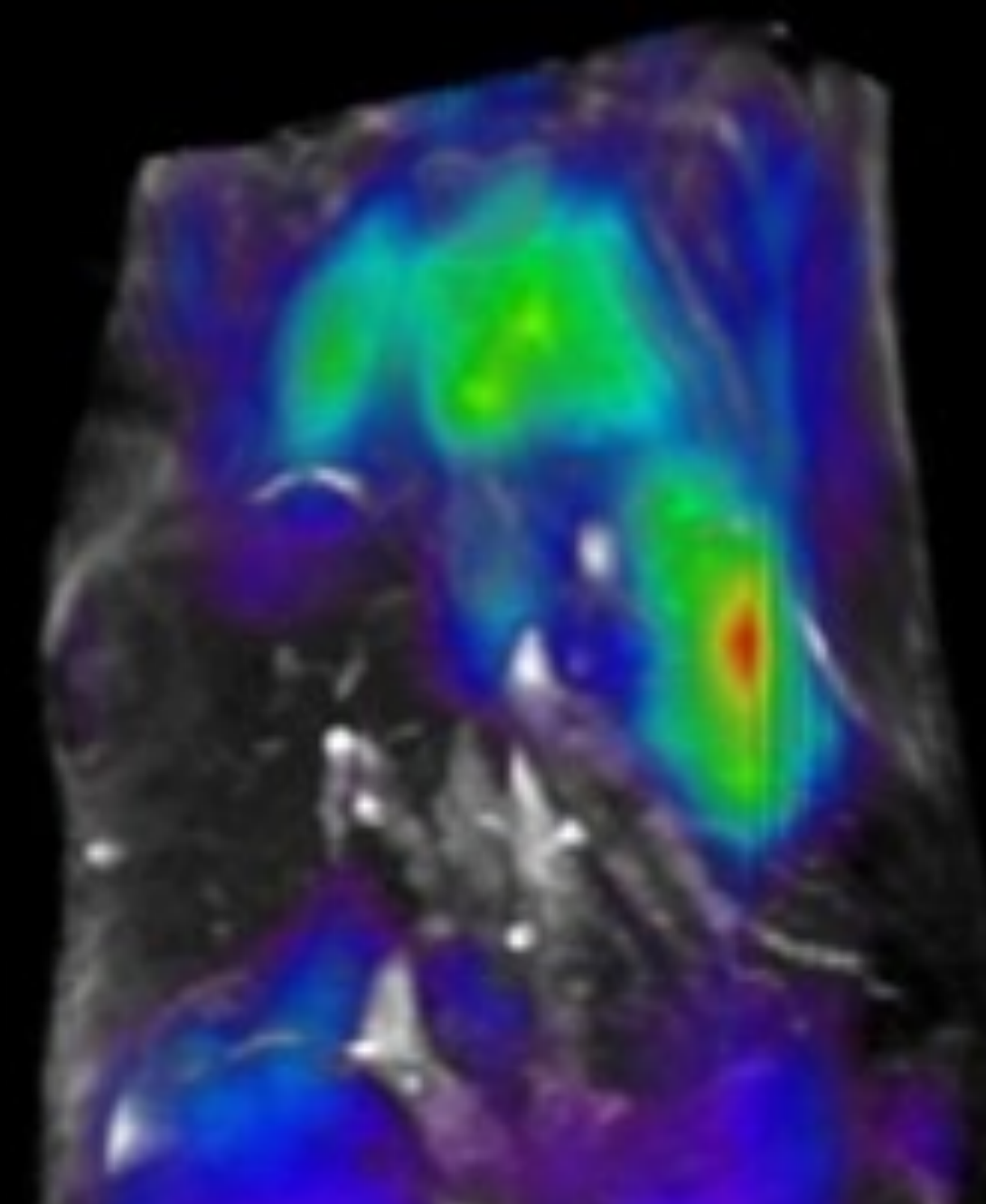
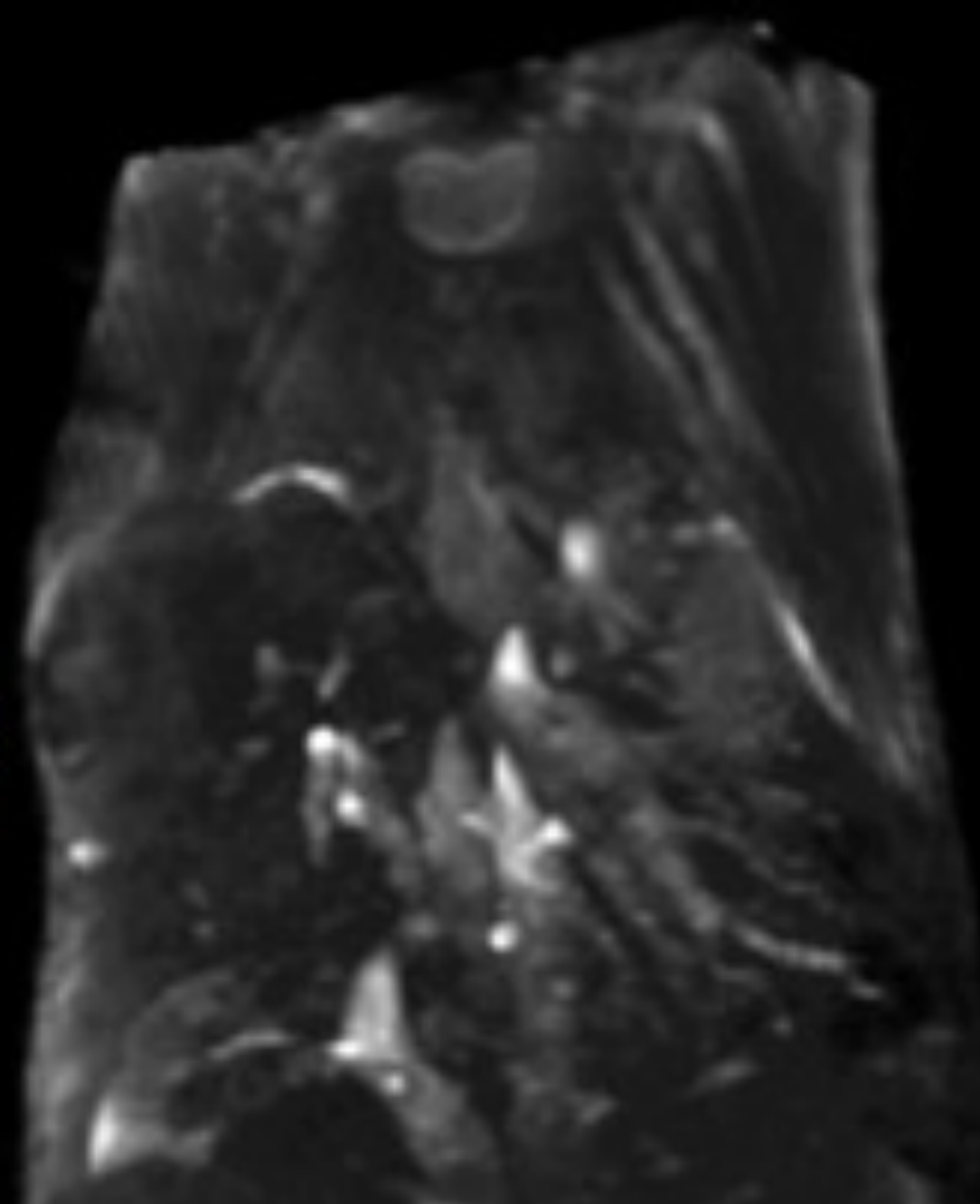
CT



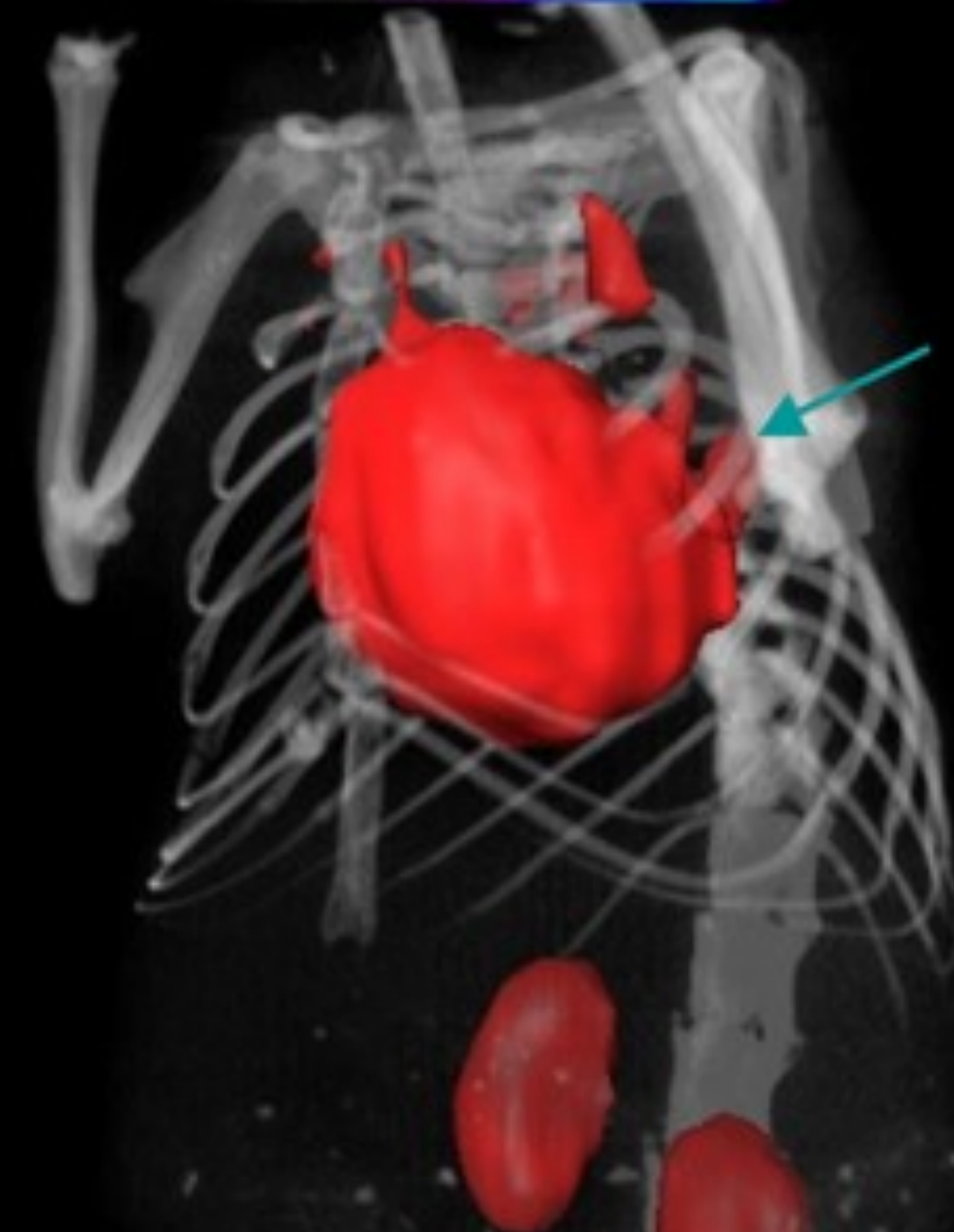
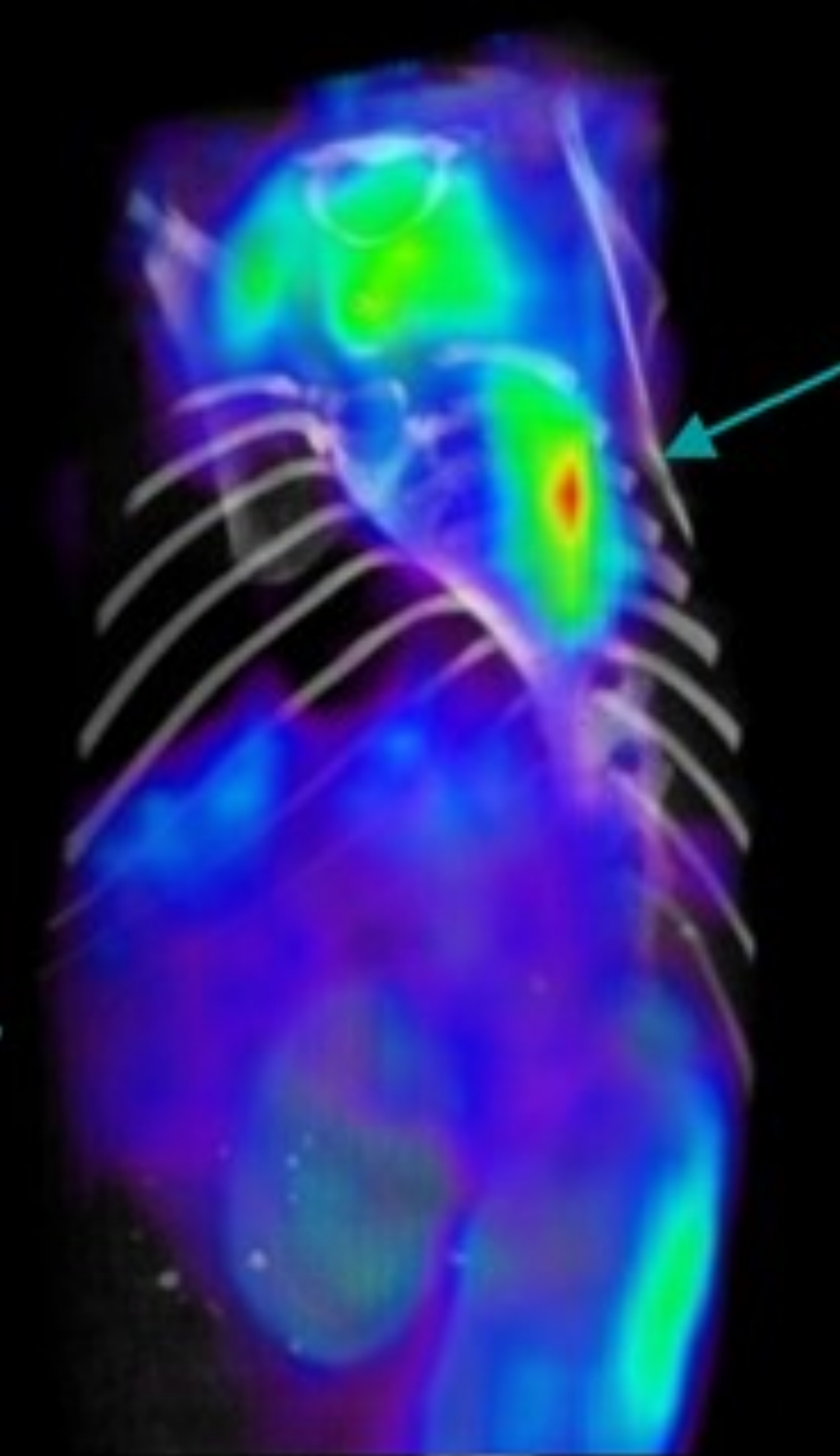
T1



T2



PT



axi



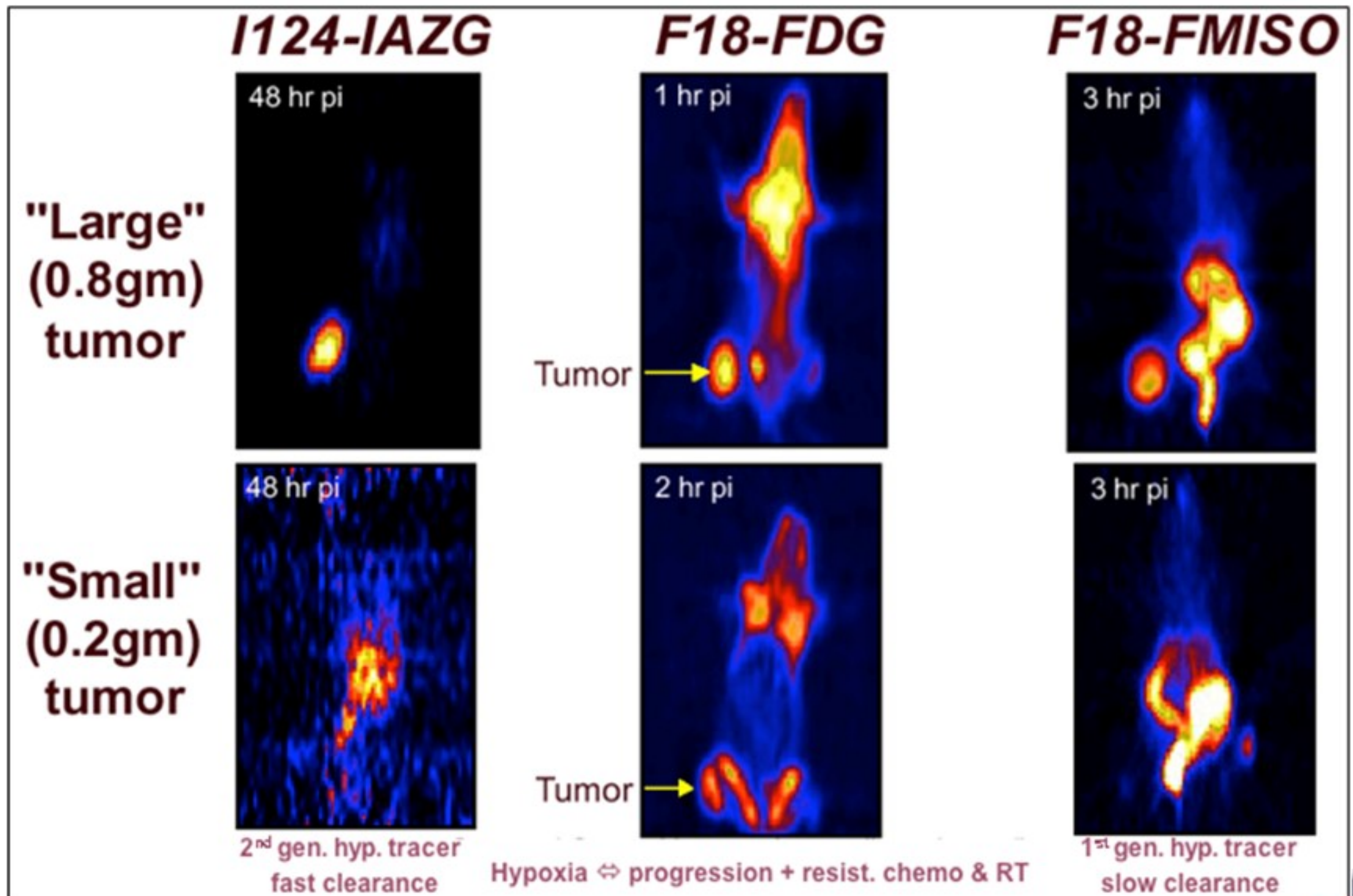
cor



3D

μ PET mouse imaging: multiple hypoxia tracers \Leftrightarrow contenders

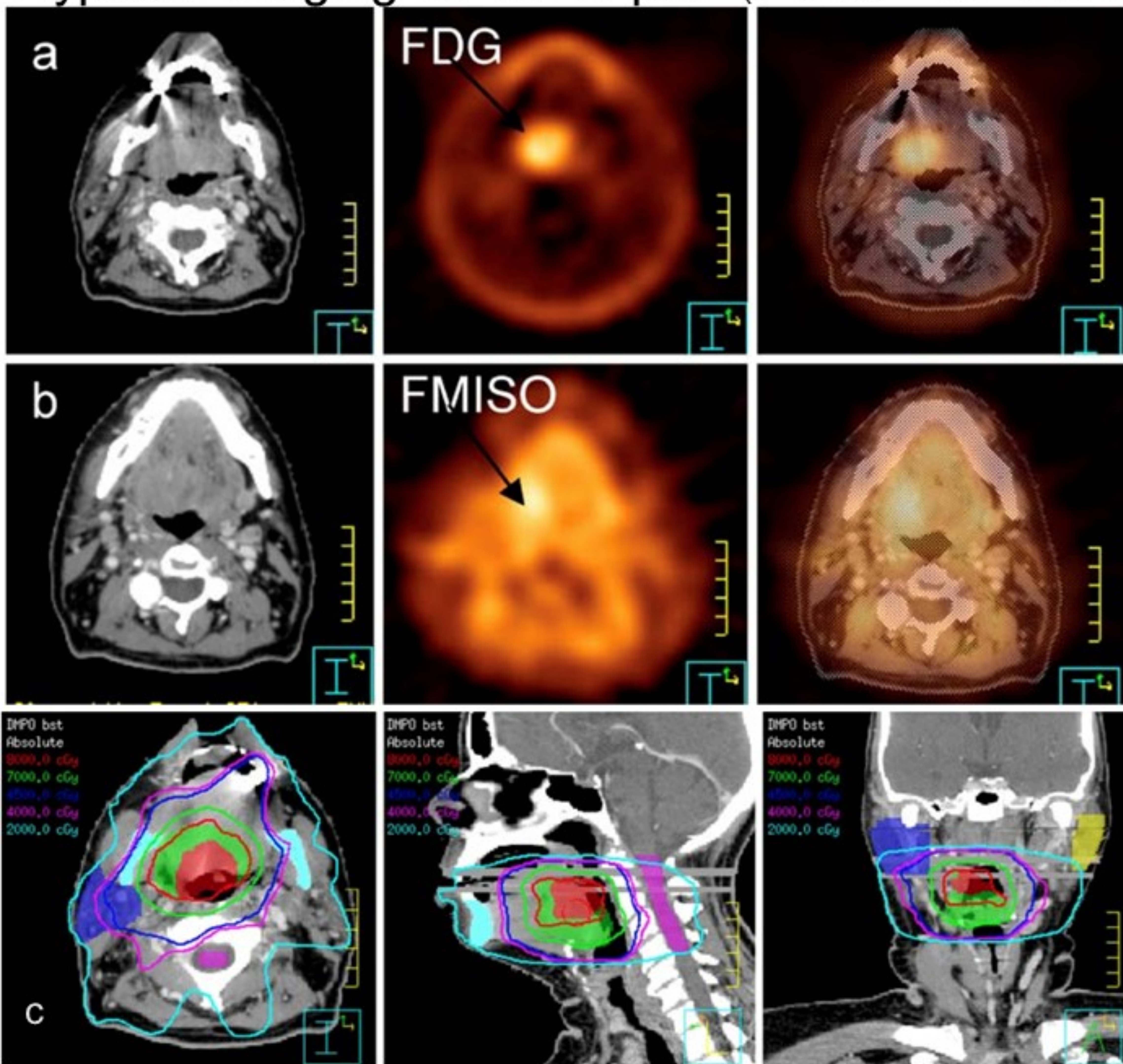
(courtesy SKI-MSKCC)



radioactive half life: ^{18}F : 109.8min; ^{124}I : 4.2d



Hypoxia imaging for IMRT plan (F-MISO = 18F-misonidazole)



From: J.G. Rajendran et al. *Eur J Nucl Med Mol Imaging*, 2006. 33:S44–S53

CT/PET: (a) FDG uptake of primary tumor, (b) FMISO uptake (hypoxic subvolume of primary tumor)

(c) MPR of isodose map (IMRT plan) superimposed to CT.

Primary **FDG PTV (green)** was treated to **70 Gy**.

FMISO PTV (red) was treated with an **additional 10-Gy** boost in a consecutive plan.

Ipsilateral parotid gland (blue) receives **<40 Gy** to 40% of its volume.

Contralateral parotid gland (yellow) receives **<20 Gy** to 20% of its volume.

Mandible (sky blue) constrained to **<70 Gy**.

Cord (pink) constrained to **<45 Gy**



The future?...

"The future is here. It's just not widely distributed yet."

William Gibson, writer

...will likely build on the past and present...

"The best way to predict the future is to create it."

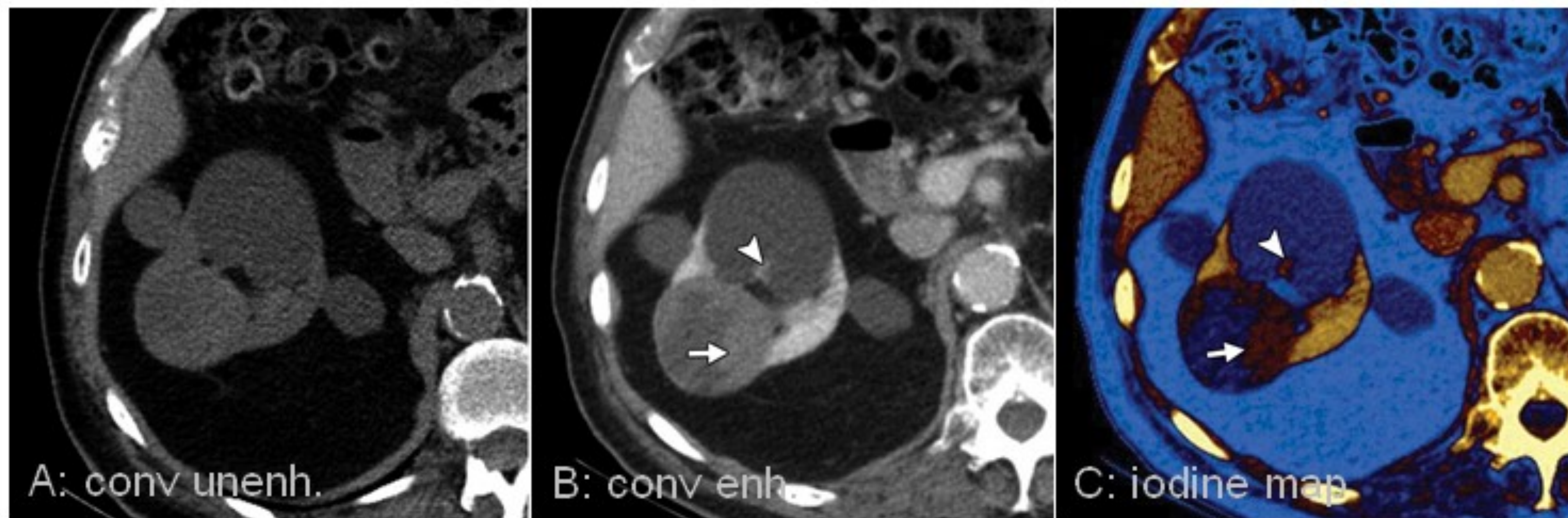
Abraham Lincoln, former US President



Dual(multi)-energy CT: => material decomposition => virtual (un)enhanced, monochromatic (to minimize pseudo-enhancement) and iodine map images



C's 14HU < B's 31HU \Leftrightarrow B's pseudoenhancement minimized => **cyst**



A+B => initial cyst diagnosis; but C's iodine enhancement > **cystic RCC**



More/better probes (PET tracers for cancer as exemple, etc.)

System	Ligand
Glucose metabolism	^{18}F -FDG
Lipid metabolism	^{11}C -acetate
Nucleic acid metabolism	^{18}F UdR (uridine)
Amino acids	^{11}C -Methionine
	^{11}C -Tyrosine
	^{18}F -FMT
	^{18}F -FET (fluoro ethyl tyrosine)
Hypoxia	^{18}F -FMISO (fluoro misonidazole)
	^{64}Cu -ATSM
	^{124}I -IAZG
	^{18}F -FAZA
Proliferation	^{18}F -FLT (fluoro deoxy thymidine)
	^{11}C -Thymidine
	^{18}F -FMAU
Bone mets	^{18}F -NaF

also: ^{11}C -Leucine, Monoclonal AntiBodies (mAb's), etc.

*"The future of PET
and NM is in
probes, not
instrumentation"*

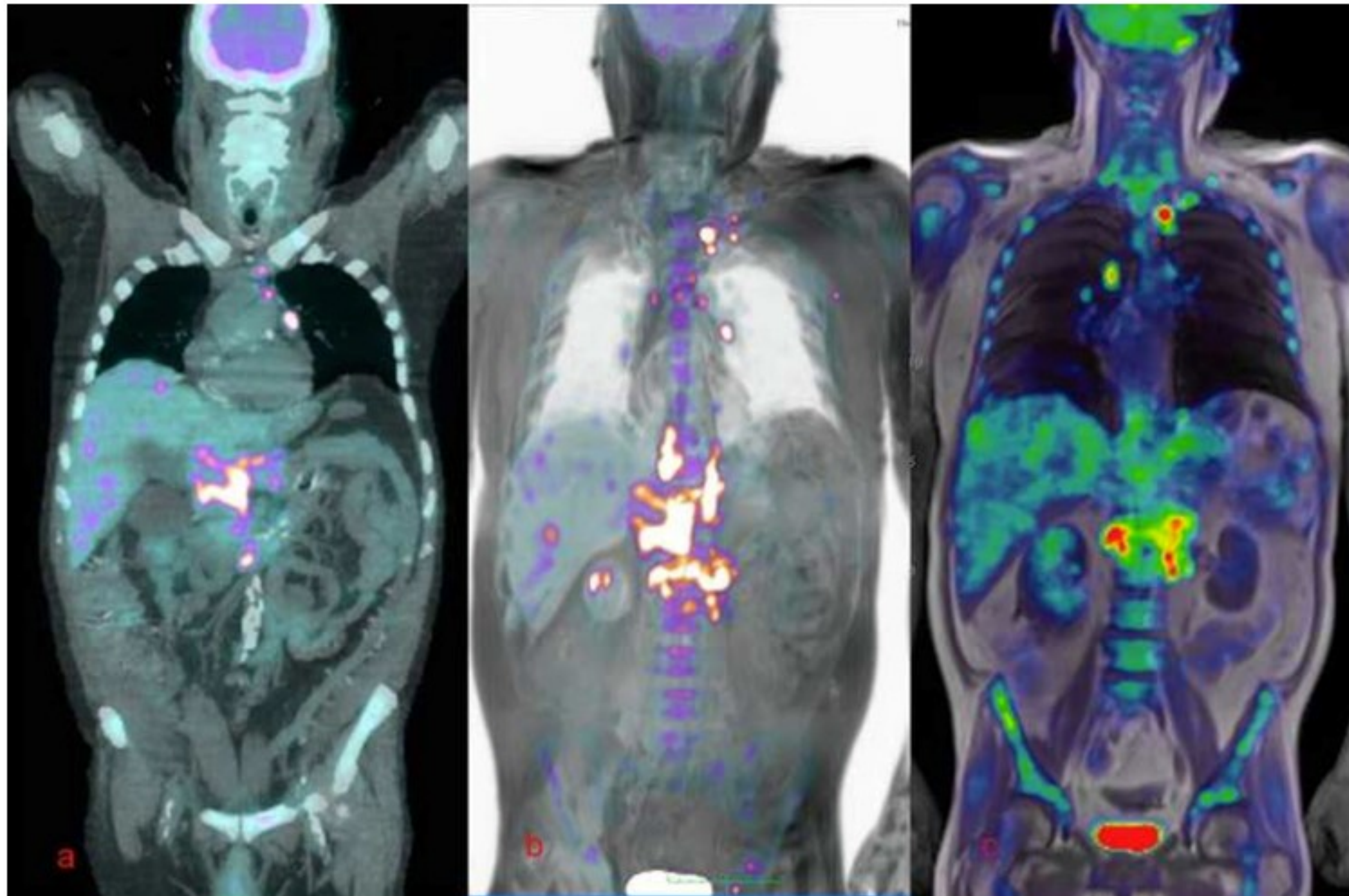
*Mike Phelps,
famous PET pioneer*

radioactive half life: ^{11}C : 20.4min; ^{18}F : 109.8min; ^{64}Cu : 12.7h; ^{124}I : 4.2d



More/better integration via hybrids and exploitation thereof

PET/CT & PET/MR (also SPECT/CT, etc.)

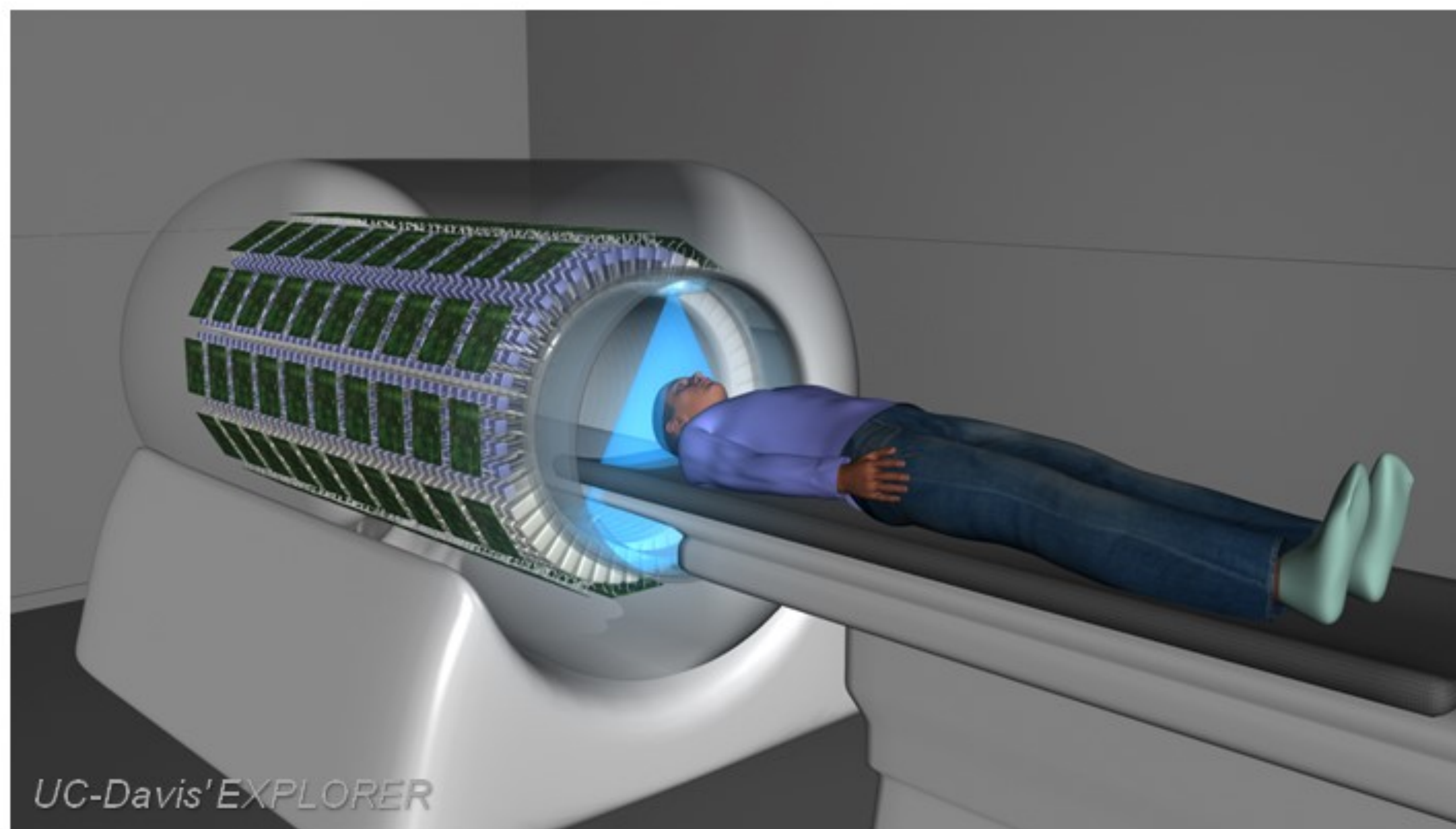


¹⁸F-FDG PET CT scan (thick MPR) (a) shows multiple confluent nodes above and below the diaphragm. Multiple liver lesions are also seen. Cross sectional PET-MRI (b) MIP and (c) volume rendered reconstruction show similar appearances.

(from F. Fraioli et al. *The British Institute of Radiology*, Nov 2013)

Truly whole body (and beyond...)

=> opportunity/challenge of new metrics & paradigms



also: industrial,
backscatter,
muons, ...

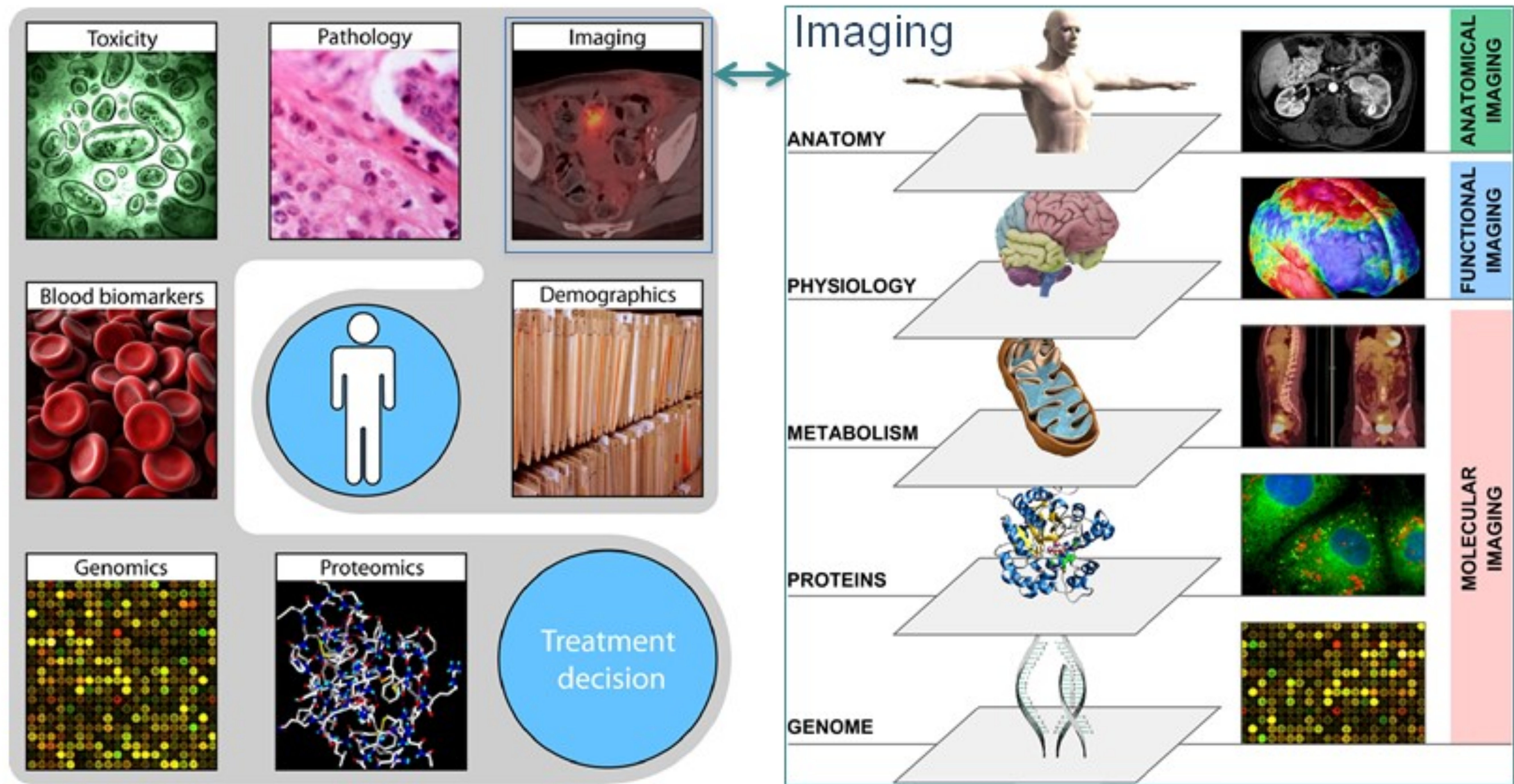


Whole-body (PET/CT melanoma): of the forest and the trees... (source data CRIF; post-proc./visu. LMB)



More/better metrics and integration

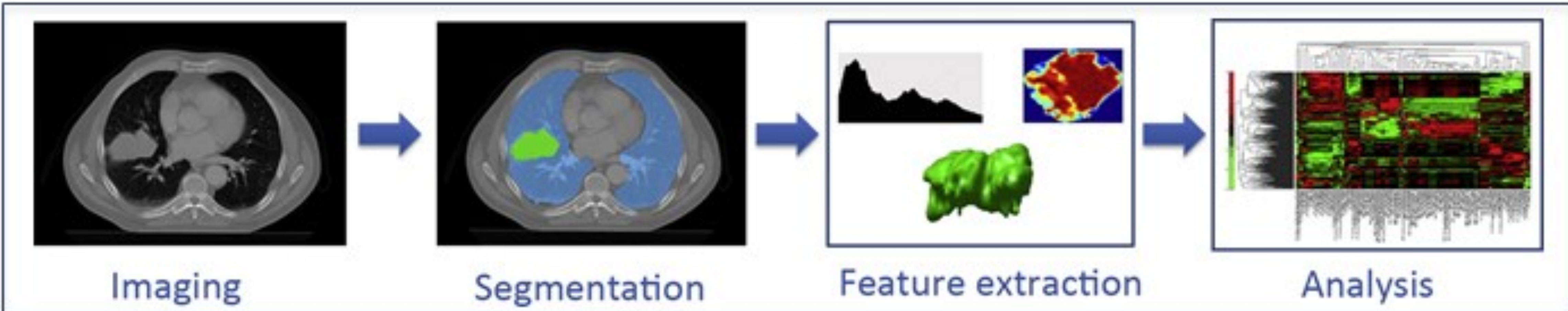
(in part from P. Lambin et al.,
Eur J of Cancer, 48(4), 2012)



appearance, size, RECIST, WHO, volume, motion, texture, DWI, shape, etc.

4D, DCE, PK, fMRI, MRS, mpMR, etc.

SUV, PERCIST, 4D, PK/PD, probe pathway, mpMR, etc.



radiomics pipeline

Extracting more information from medical images using advanced feature analysis ⇔ "advanced CAD"



"Imaging" as part of a bigger whole... "Big Data" (+ *M & D Learning...*)

Transforming cancer patient care with IBM Watson* for Oncology.

In the analyst call, Michael Karasick, IBM VP of Innovations, described a new type of healthcare hierarchy of needs when he spoke of Watson's evolution.

- **assistance** that the computer would be able to give medical practitioners in their diagnosis and treatment of patients.
- **understanding**, where Watson mapped certain patterns and conditions pertaining to the patient, based upon what the practitioner input.
- **decisions** on potential diagnoses, which were based upon the earlier inputs from practitioners and the mapping of conditions and patterns.
- **discovery**, is where the analytics peruse a vast body of papers and other subject matter expert inputs related to research on the condition being studied.

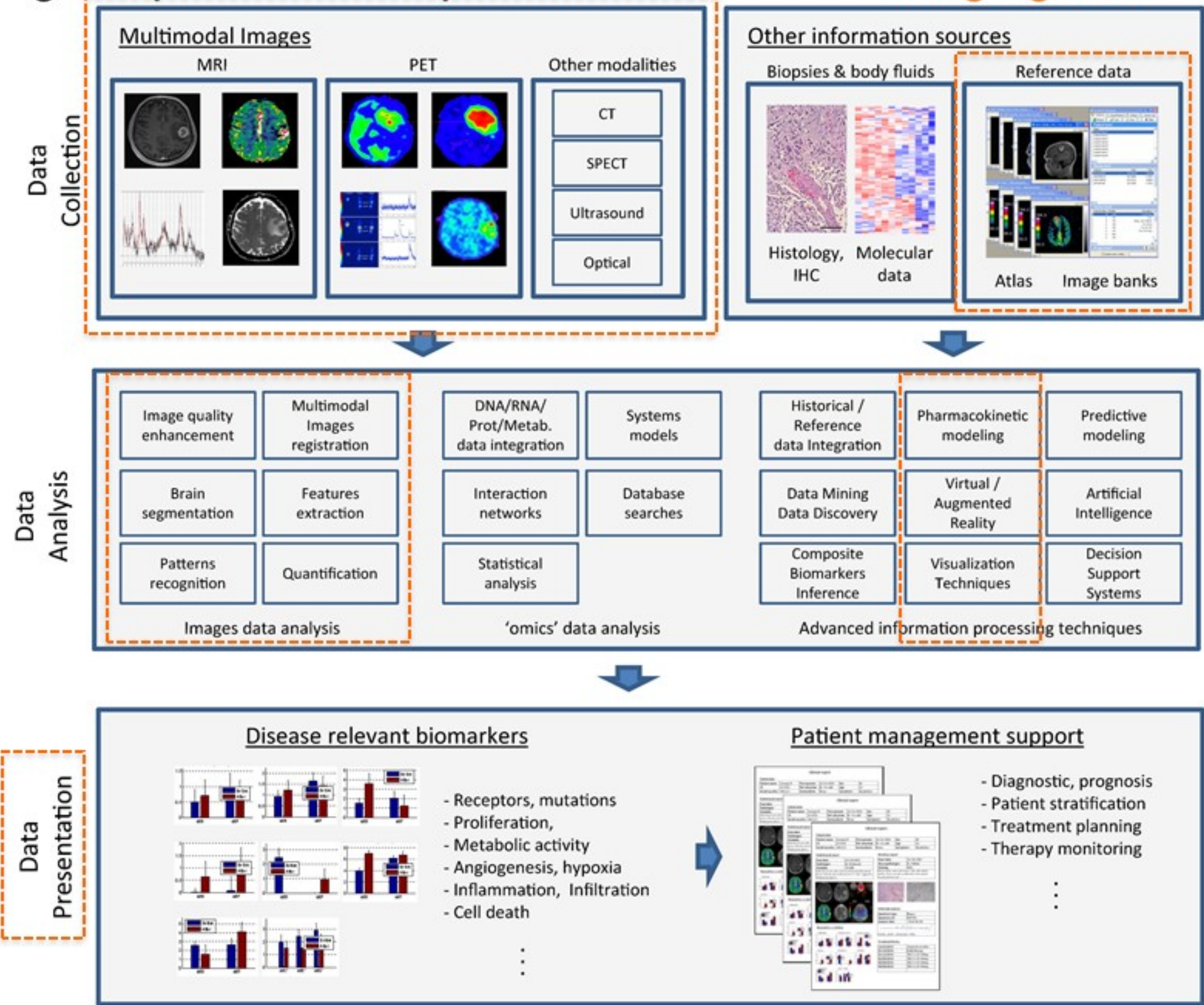
Meet the computer diagnosing cancer



By @loufaglia @CNNTech

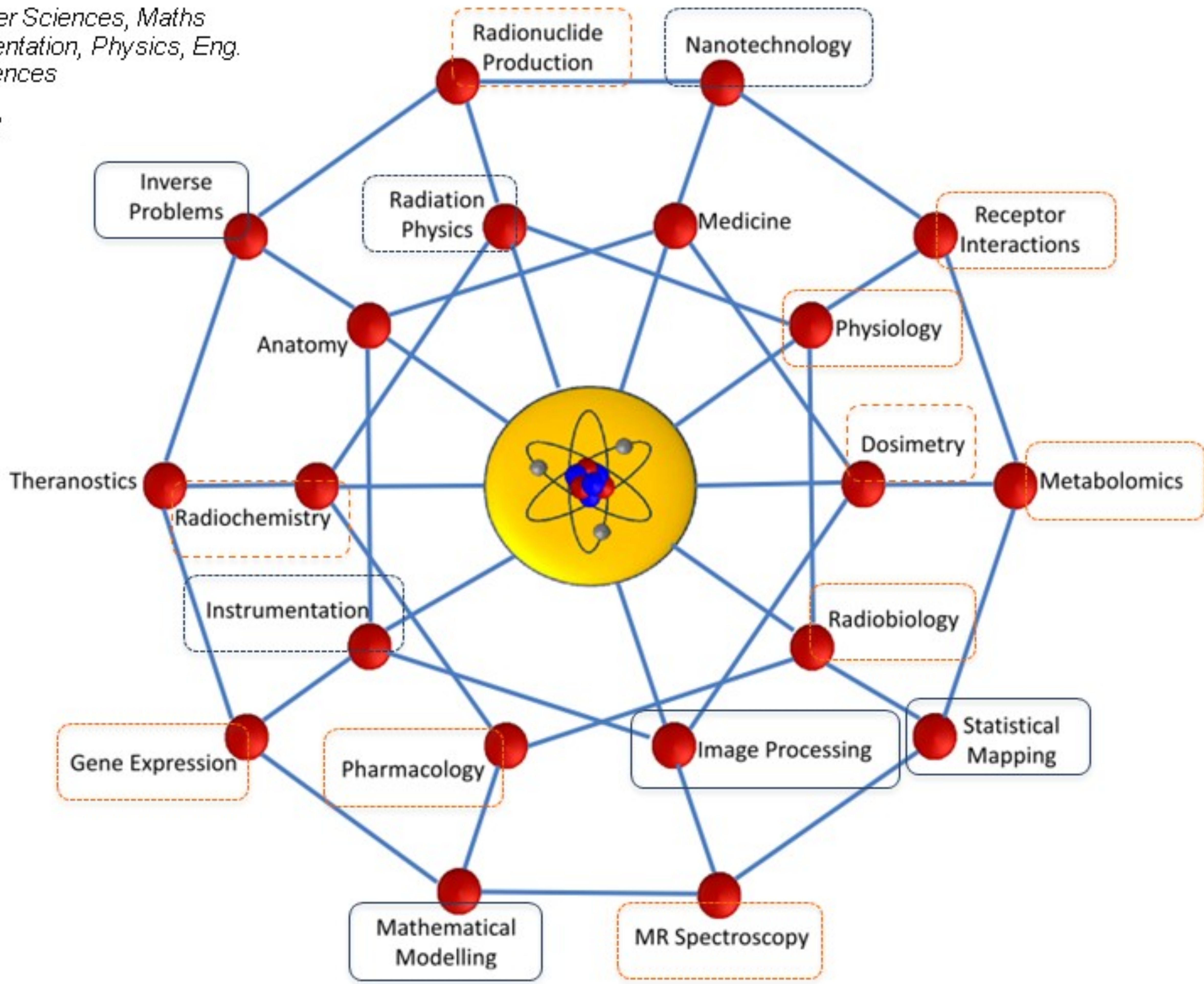


End goal: personalised/precision medicine: **imaging bits**



Biomedical imaging: inherently multidisciplinary

(derived from D. Bailey,
EJNMMI Physics 2014, 1:4)



Conclusions:

**Software registration isn't dead
and remains a key component of modern*
imaging and protocols (*incl. hybrids*)**

**Much is happening on the advanced*
imaging and translational* sides of things**

Personalized/precision medicine is imminent*...

**moving targets...*



Acknowledgments:

➤ Most examples and thoughts from work and time spent at:

- Service Hospitalier Frédéric Joliot (SHFJ) (*Instrumentation & Computing Head a.i.*)
CEA, Orsay, F
- Biophysics & Nuclear Medicine (*Research & Computer Scientist*)
UCLA, Los Angeles, USA
- PET & Biomedical Cyclotron Center (*founding co-Director, Instrum. & Comp. Head*)
ULB, Brussels, B
- Laboratory of Functional and Multidimensional Imaging (LFMI) (*founding Director*)
HUG, Geneva, CH
- Departments of Radiology & Medical Physics (*AIL founding Technical Director*)
MSKCC, New York, USA
- Image Processing & Visualization Laboratory (IPVL) &
Scientific Computing Resource (SCR) (*founding Director*)
Dept of Imaging Physics; MDACC, Houston, USA
- Centre for Oncology and Molecular Medicine - DMS (*Imaging Head*)
Clinical Research Imaging Facility (CRIF) (*founding Head*)
Division Imaging & Technology (*founding co-Head*)
University & Ninewells Hospital, Dundee, UK
- School/Dept. of Computer Science (*Prof. of Medical Imaging, Head*)
University of Lincoln, UK

➤ Other contributions as credited; past, present and **future** collaborations and partners

