

# Developments in AI model: 2D and 3D

14/03/25

[lorenzo.marini@phd.unipi.it](mailto:lorenzo.marini@phd.unipi.it)

# Sommario

## Fase 2D

- Risultati finali
  - Modello, addestramento, gamma-analisi
- DEMO InTrEPID-2D system
- Paper per Radiotherapy & Oncology

## Fase 3D

- Data exploration
- Coregistrazione CT-RTDOSE
- Qualche idea...

Fase 2D

# Dati, preprocessing, rete, training, gamma-analisi

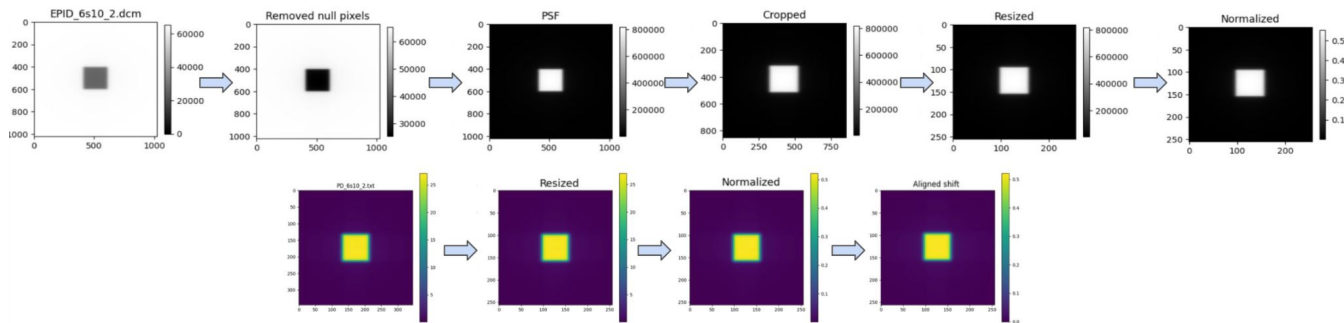
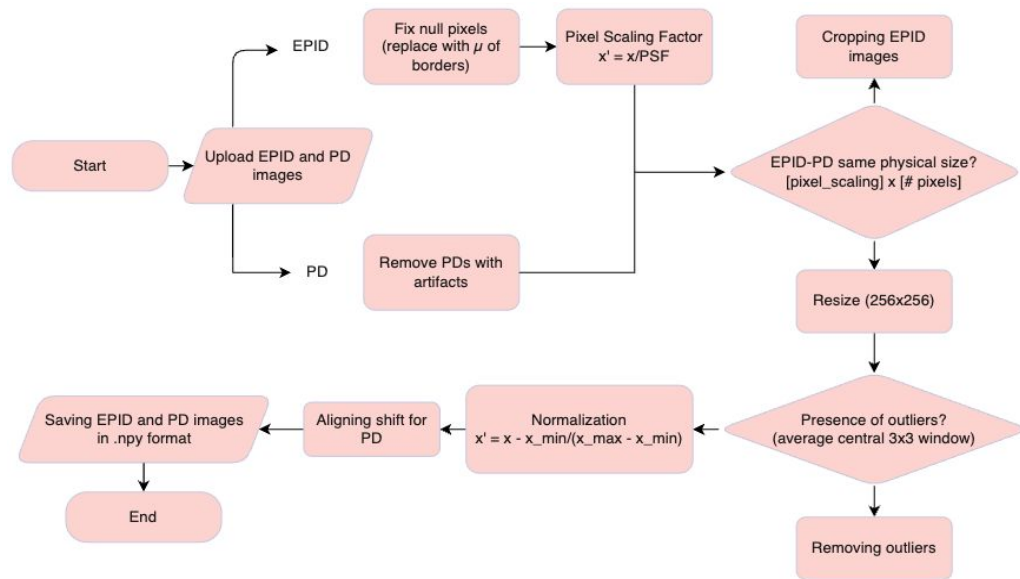
Table 1: Composizione del dataset suddiviso per forma e dimensione dei fantocci. OMG= Omogeneous, MP=Multiplug

Area [cm <sup>2</sup> ]	OMG	CIRS	SLAB	MP	Tot
2x2	14	8	7	-	29
5x5	12	9	7	12	40
10x10	16	7	7	8	38
20x20	14	-	7	-	21
2x10	13	8	7	8	36
10x2	13	8	2	-	10
15x1	4	8	-	8	16
1x15	-	8	-	-	8
4x20	-	-	7	-	7
Ellipses	-	-	-	8	8

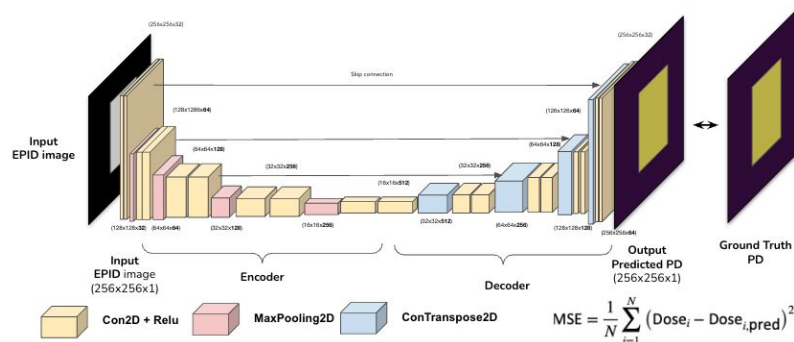
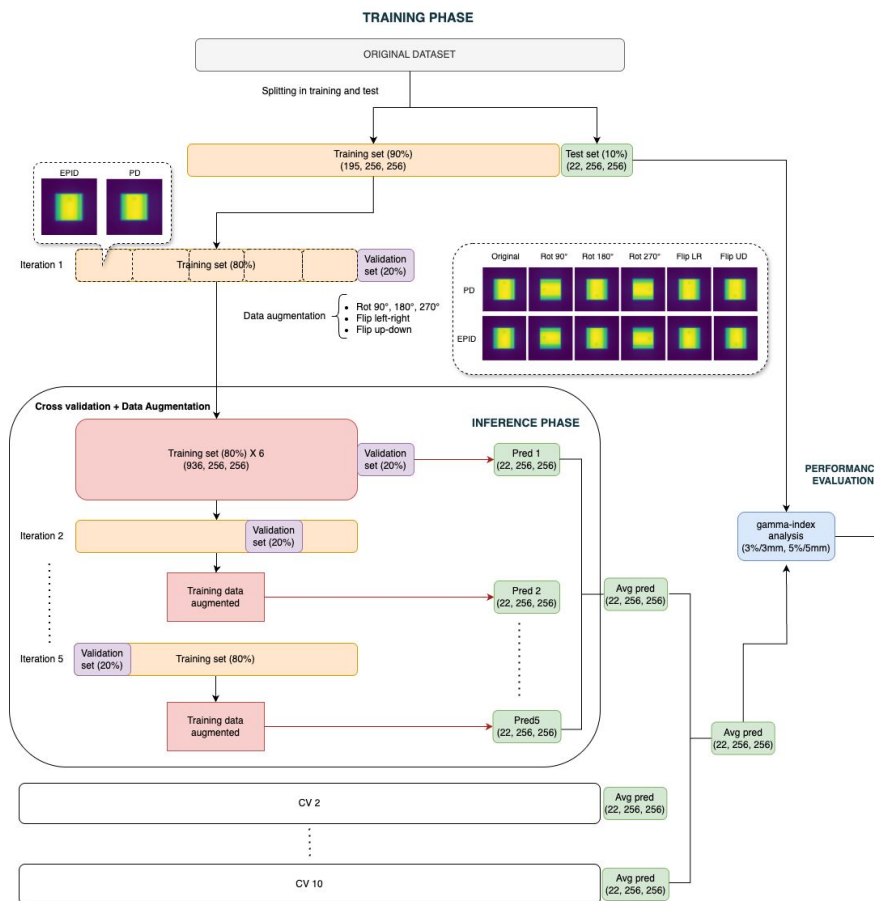
	Phantoms	QUADRATI				RETTANGOLI					CERCHI/ELLISSI
		2x2 cm^2	5x5 cm^2	10x10 cm^2	20x20 cm^2	2x10 cm^2	10x2 cm^2	15x1 cm^2	1x15 cm^2	4x20 cm^2	
TEST SET	Omogeneous	6s10b3_1	6s30b1_2	6s10b2_3	6s20_4	6s10b2_6				SLABBone_5	
		CIRS_Ogra	CIRS_270	CIRS_Ogradi		CIRS_270gra	CIRS_90gr	CIRS_Ogra	CIRS_Ogra		
		CIRS di_ISOlesio	gradi_ISO	_ISOlesione		di_ISOcentro	adi_ISOlesi	di_ISOcent	di_ISOlesio		
		ne_1	lesione_2	_3		_6	one_7	ro_8	ne_9		
	SLAB	SLABinbo	SLAB_Bo	SLABBone_3	SLAB_Bone_2slab_4	SLABinhomA					
		mAir2_1	neb2_2slab_2			ir2_6					
			Multiplug_90g_5c	Multiplug_P		Multiplug_P					
	Multi_Plug		m_ARIA_2	MMA_90g_3		MMA_180g_6		Multiplug_Ti_90g_8		Multiplug_Ti_tondo_1E0g	
	# images	3	4	4	2	4	1	2	1	1	1



# Dati, preprocessing, rete, training, gamma-analisi



# Dati, preprocessing, rete, training, gamma-analisi



## Hyperparameter Optimization

A GS approach was implemented to find out the best hyperparameter configurations: # filters: [8, 16], learning rate: [0.01, 0.001, 0.0001], batch size: [4, 8, 16, 32] → 24 different configuration!

## Cross-Validation Strategy

To improve generalization and reduce overfitting, **10 CV** schemes were applied. Each training: **150 epochs per model, 5-fold CV per training**

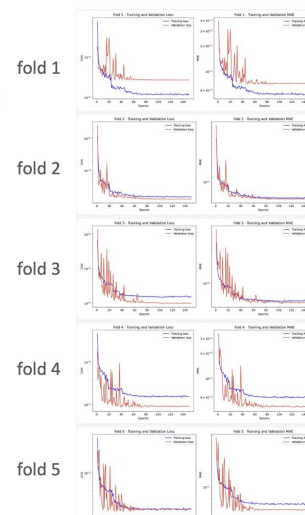
## Ensemble Learning

An EL approach was adopted by averaging the predictions of the best models from each training run.

## Hardware

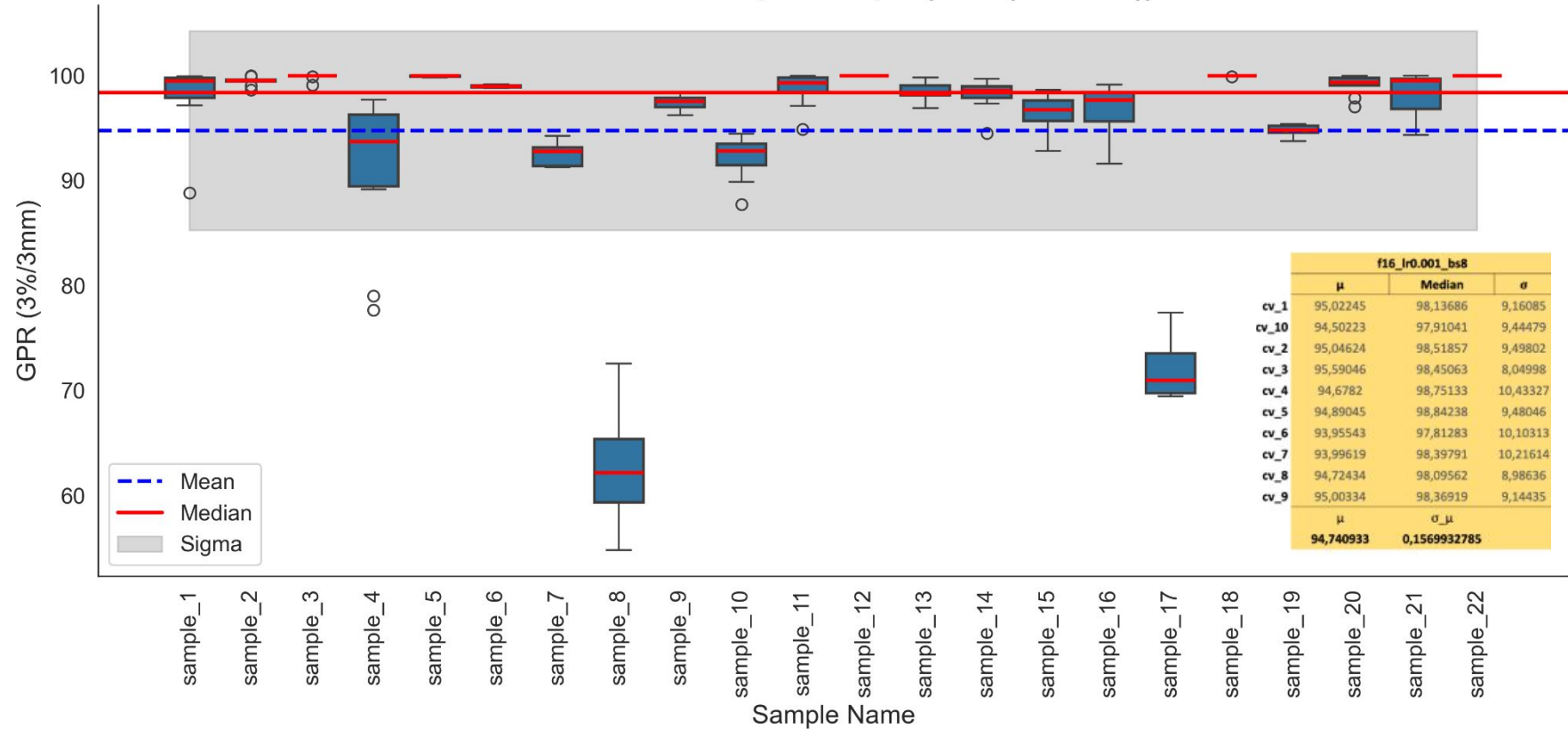
The training process was conducted using the **NVIDIA V100 Tensor Core GPU** of the **Computing Center of the INFN - Pisa division**.

**f16, lr0.001, bs8**

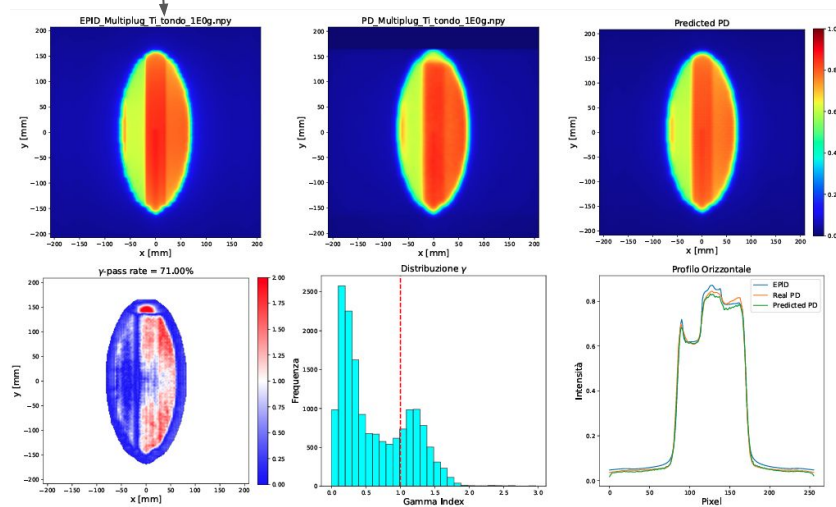
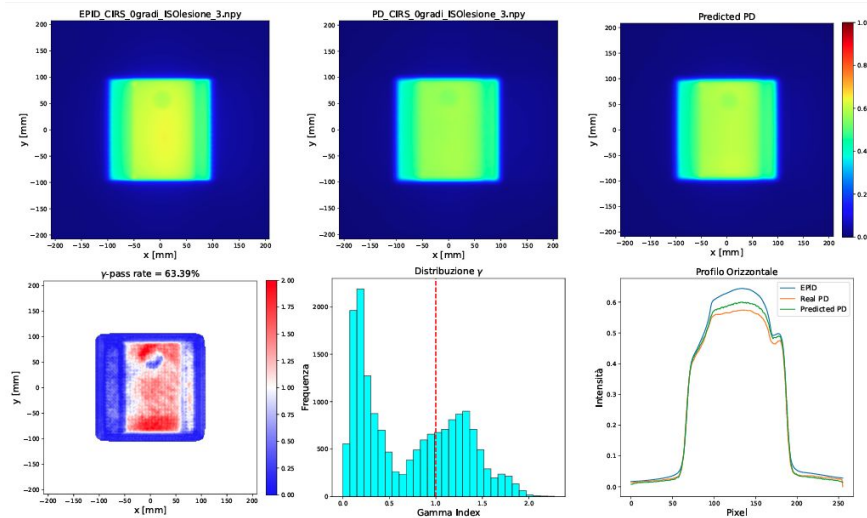
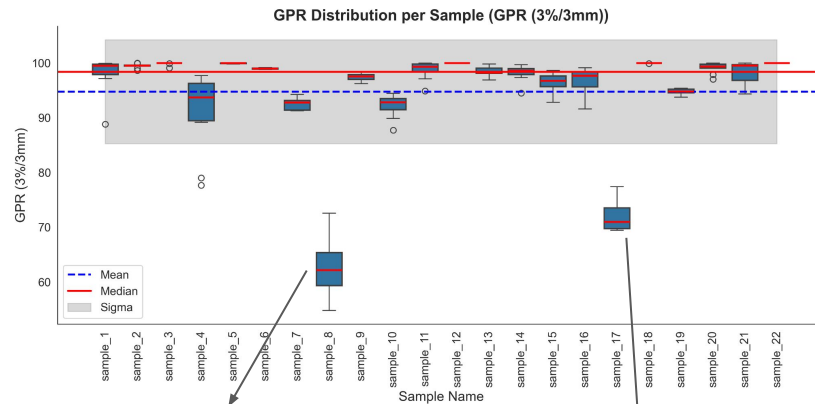


# Dati, preprocessing, rete, training, gamma-analisi

GPR Distribution per Sample (GPR (3%/3mm))



# Dati, preprocessing, rete, training, gamma-analisi

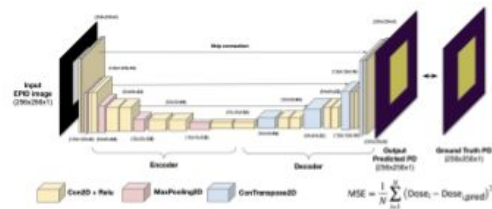




189 pairs of EPID and PD images were allocated to the training set, 22 pairs to the test set.

TRAINING SET	QUADRANT					RETANGULO					CERCHETTO				
	Phantoms	2x2 cm <sup>2</sup>	3x3 cm <sup>2</sup>	5x5 cm <sup>2</sup>	10x10 cm <sup>2</sup>	20x20 cm <sup>2</sup>	2x2 cm <sup>2</sup>	3x3 cm <sup>2</sup>	5x5 cm <sup>2</sup>	10x10 cm <sup>2</sup>	2x2 cm <sup>2</sup>	3x3 cm <sup>2</sup>	5x5 cm <sup>2</sup>	10x10 cm <sup>2</sup>	20x20 cm <sup>2</sup>
Grayscale	14	12	16	14	13	13	8	8	8	8	13	4			
CPS	8	8	7			8	8	8	8		7				
SLAR	7	7	7	7		7	7	7	7		7				
Multi_Plan	12	8				8					8				
Somma	29	40	38	21	26	38	30	38	8	20	31				
Percentage	13.00%	17.00%	17.00%	9.00%	18.00%	4.00%	7.00%	5.00%	5.00%	9.00%	5.00%				

TEST SET	QUADRANT					RETANGULO					CERCHETTO				
	Phantoms	2x2 cm <sup>2</sup>	3x3 cm <sup>2</sup>	5x5 cm <sup>2</sup>	10x10 cm <sup>2</sup>	20x20 cm <sup>2</sup>	2x2 cm <sup>2</sup>	3x3 cm <sup>2</sup>	5x5 cm <sup>2</sup>	10x10 cm <sup>2</sup>	2x2 cm <sup>2</sup>	3x3 cm <sup>2</sup>	5x5 cm <sup>2</sup>	10x10 cm <sup>2</sup>	20x20 cm <sup>2</sup>
Grayscale	1	1	1	1	1	1	1	1	1	1	1				
CPS	1	1	1	1	1	1	1	1	1	1	1				
SLAR	1	1	1	1	1	1	1	1	1	1	1				
Multi_Plan	1	1	1	1	1	1	1	1	1	1	1				
Somma	5	5	5	5	5	5	5	5	5	5	5				



DL Architecture

5 depth levels with skip connections between encoder and decoder. Each level applies **2 convolutions (3x3)** with filters doubling up to **128** at the bottleneck, using **MaxPooling** for downsampling and **Transposed Convolutions** for upsampling, with **Batch Normalization** in deeper layers and **ReLU** as activation.

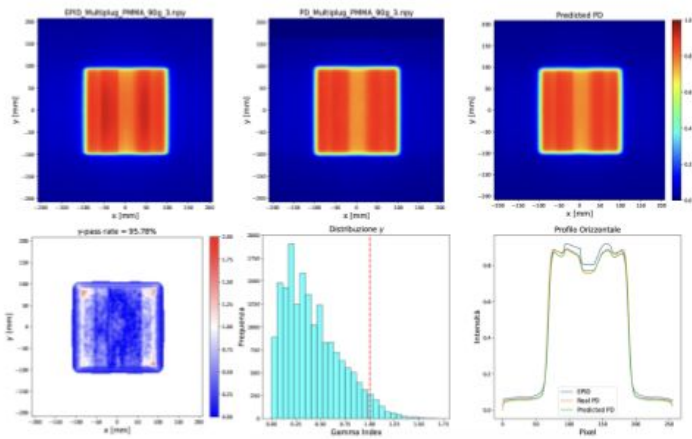
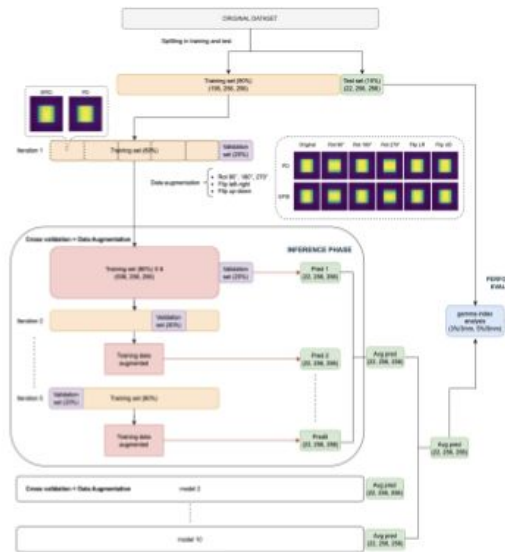
a)	3%/3mm	$\mu$	$\sigma_\mu$	Med
		94,53%	2,08%	98,00%
	5%/5mm	99,55%	0,25%	100,00%
b)	3%/3mm	$\mu$	$\sigma_\mu$	Med
		92,45%	2,42%	98,00%
	5%/5mm	99,27%	0,48%	100,00%
c)	3%/3mm	$\mu$	$\sigma_\mu$	Med
		92,21%	2,59%	98,00%
	5%/5mm	98,44%	1,33%	100,00%

## Conclusions

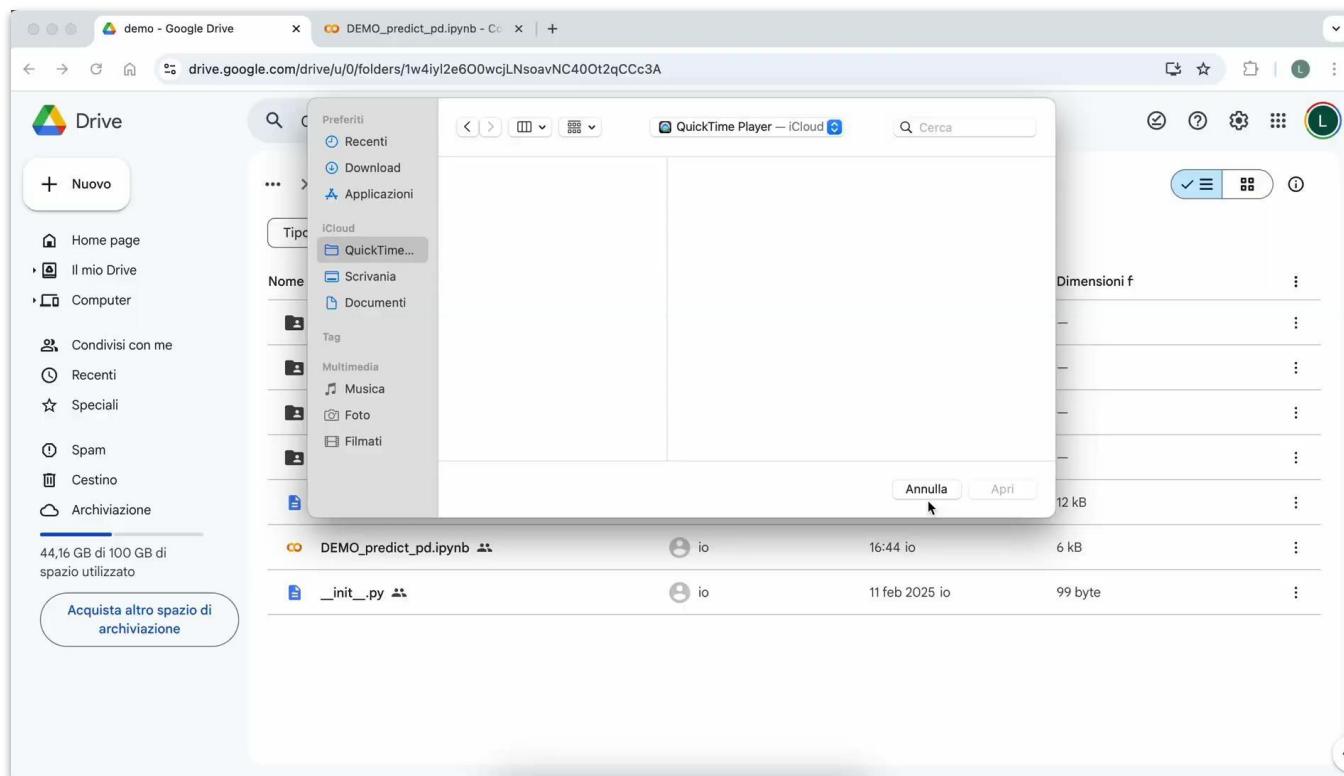
### Training

Grid search: # initial filters (8, 16), learning rate (0.01, **0.001**, 0.0001), batch size (4, 8, 16, 32)

Training phase: 10 cv with k=5, epochs 150, data augmentation



# DEMO 2D-InTrEPID system



<https://drive.google.com/drive/folders/1w4iy12e6O0wcjLNsoavNC40Ot2qCCc3A?ths=true>



# Paper

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Per essere aggiunti, chiedere ad @Alessandra

6 pagine, 6 figure. Manca qualche autore??

## Deep learning methods for transit dosimetry with EPID images

L. Marini<sup>1,2</sup>, C. Mozza<sup>4</sup>, A. C. Kraan<sup>5</sup>, M. Avanzo<sup>4</sup>, L. Lizza<sup>2</sup>, A. Retico<sup>2</sup>, I. Meattini<sup>4,6</sup>, L. Marazzo<sup>4,6</sup>, C. Talamonti<sup>1,4,6</sup>

<sup>1</sup>University of Pisa, Department of Computer Science, Pisa, Italy

<sup>2</sup>Istituto Nazionale di Fisica Nucleare (INFN), Pisa Division, Pisa, Italy

<sup>3</sup>Centro di Riferimento Oncologico (CRO) di Aviano, Florence, Italy

<sup>4</sup>Department of Experimental and Clinical Biomedical Sciences Mario Serio, University of Florence, Florence, Italy

<sup>5</sup>Istituto Nazionale di Fisica Nucleare (INFN), Section of Florence, Florence, Italy

<sup>6</sup>Azienda Ospedaliero-Universitaria Careggi, Florence, Italy

### Abstract

**Background and Purpose:** The complexity of external beam photon radiotherapy has significantly increased, leading to a higher risk of errors that require advanced systems for dose verification. Electronic Portal Imaging Devices (EPIDs), originally designed for patient positioning, have become essential tools for in-vivo dosimetry (IVD) and pre-treatment dose verification. Traditional dose reconstruction methods, such as physical models and Monte Carlo simulations, face limitations in complexity, computational power, and time-efficiency, hindering routine clinical use.

**Materials and methods:** This study introduces a deep learning (DL) framework for 2D dose reconstruction using real EPID images acquired during treatment. A 2D U-Net model maps EPID images to water-equivalent portal dose images (PDIs), leveraging a database of 211 EPID-PD pairs obtained from phantoms with diverse material densities. The performance was evaluated through mean absolute error (MAE) metrics and 2D -analysis with a 3mm/3% and 5mm/5% criterion. **Results:** 2D framework achieved a predictive mean -passing rate of  $(94.53 \pm 2.08)\%$  and  $(99.55 \pm 0.25)\%$  for the 3mm/3% and 5mm/5% criteria, respectively. The median -passing rates across all test cases were 98.03% and 100.00%, demonstrating the model's robustness across different scenarios. The dose predictions completed within 1 second, significantly faster than TPS calculations ( $\approx 20$ –30 minutes). **Conclusions:** Deep learning enables real-time, accurate dose reconstruction from EPID images, demonstrating its potential for in-vivo treatment verification. This work highlights the performance of the 2D model, emphasizing their clinical potential in radiotherapy.

**Keywords:** EPID, Transit dosimetry, Radiotherapy, Deep Learning, U-Net

### 1. Introduction

In the last decades, the complexity of external beam photon radiotherapy treatments has increased significantly [1]. However, due to the advancement in technical complexity, the risk of errors has grown too requiring sophisticated systems to ensure correct dose administration and verification [2]. Errors caused by equipment malfunctioning, patient positioning or anatomical changes must be promptly detected to prevent accidents and comply with the stringent EU dose verification guidelines [3].

Electronic Portal Imaging Devices (EPIDs), initially designed for real-time monitoring patient positioning, have become the preferred tools for both in-vivo dosimetry (IVD) and for pre-treatment dose verification

[4, 5, 6]. EPIDs measure X-ray fluence on an amorphous silicon flat-panel detector, producing 2D digital images containing dose information (the so-called Portal Dose, PD). Therefore, a dose reconstruction is then possible, and it is generally achieved by means of physical models and Monte Carlo simulations. Nonetheless, these methods have several limitations, due to the complexity of the dose reconstruction algorithms, the computing power and the significant time-consuming, limiting their routine clinical use.

In this scenario, Artificial Intelligence, in particular Deep Learning (DL), offers a promising alternative for IVD in radiotherapy [7, 8, 9]. By bypassing the traditional physical models, DL can achieve Monte Carlo-level accuracy while maintaining the speed of a trained model. Most existing studies are mainly based on simu-

# Struttura del paper

## 1. Introduction

## 2. Materials and methods

- Experimental setup and data collection*
- Preprocessing*
- Deep learning model and training phase*
- Grid search, repeated k-fold cross-validation and ensemble learning*
- Evaluation of the performance*

## 3. Results

- Dataset quality*
- Model performance*

## 4. Discussion

## 5. Credit authorship contribution statement

## 6. Declaration of competing interest

## 7. Acknowledgements

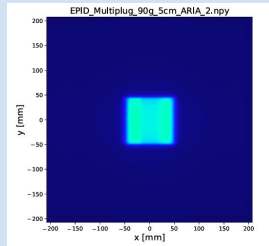
## 8. References

- ing phantoms with higher density gradients, where the  $\gamma$ -passing rate dropped below 77%. This suggests that the model may struggle with more complex dose distributions, where important scatter effects or significant heterogeneities are present.
- 5. Credit authorship contribution statement**
- Lorenzo Marini; Carlotta Mozzi; Aafke C. Kraan; Michele. Avanzo; Francesca Lizzi; Alessandra Retico; Icro Meattini; Livia Marrazzo; Cinzia Talamonti;
- 6. Declaration of competing interest**
- The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.
- 7. Acknowledgements**
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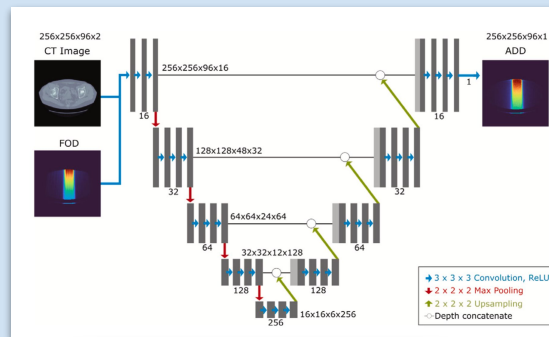


Fase 3D

EPID (.dcm)



2D InTrEPID  
System



### 2.3. First order dose approximation

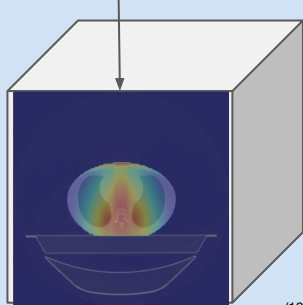
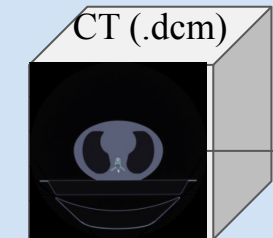
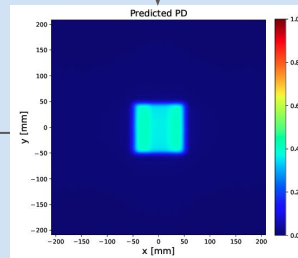
With every patient 3D dose simulation, the corresponding EPID transmitted signal was simultaneously scored as a 2D dose-to-water distribution, as explained in 2.2. This 2D EPID signal was used to produce a 3D FOD approximation inside the patient geometry. The FOD is corrected for geometric magnification, inverse square effects and attenuation inside the patient geometry.

The corrections were performed separately: first, the 2D dose-to-water distribution was backprojected into several parallel planes  $d$  towards the radiation source, creating a 3D matrix of dose-to-water distribution ( $D_{w,d}$  in Eq. (3)). This step was performed using a modified 3D ray tracing algorithm [55] based on the Siddon method [56], which accounted for geometric magnification. Next, the intensity of the signal was corrected for the inverse square law effect (second term in Eq. (3)), where  $r_d$  and  $r_{EPID}$  are the distances from the radiation source to the projection plane  $d$  and the EPID plane, respectively.

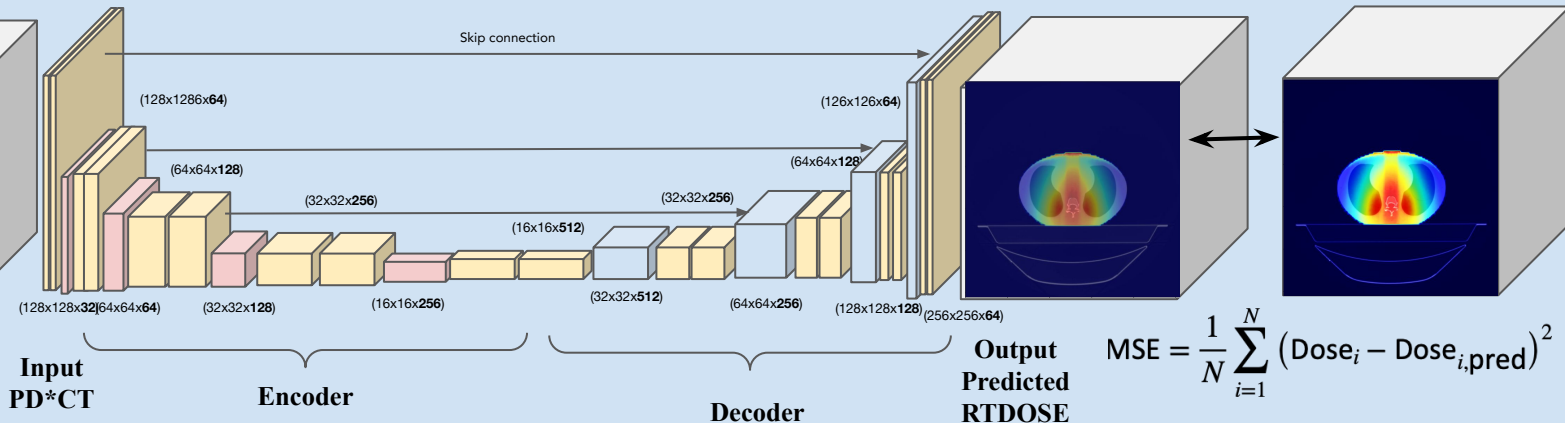
$$FOD_d = D_{w,d} \cdot \left( \frac{r_d}{r_{EPID}} \right)^{-2} \cdot e^{(\mu_{w,E} \cdot L)} \quad (3)$$

Predicted PD (.npy)

*Martins et al. Physica Medica, 2023*



$\approx$  Dose grezza



$$MSE = \frac{1}{N} \sum_{i=1}^N (Dose_i - Dose_{i,pred})^2$$

## data

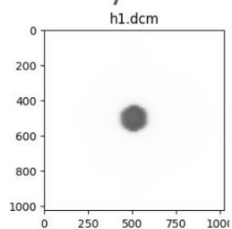
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CIRS Invivo statico^1cm sotto SW_CIRSin vivo RTDOSE_2022-11-23_130930_..._n1_ (0000, 0050)
CIRS Invivo statico^1cm sotto SW_CIRSin vivo RTDOSE_2022-11-23_130930_..._n1_ (0000, 0060)
CIRS Invivo statico^1cm sotto SW_CIRSin vivo RTDOSE_2022-11-23_130930_..._n1_ (0000, 0070)
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CIRS_Invivo_statico_1cm_sotto_SW_CIRSin vivo RTDOSE_2022_11_23_130930_1cm_sotto_SW
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portali 2023 02 16
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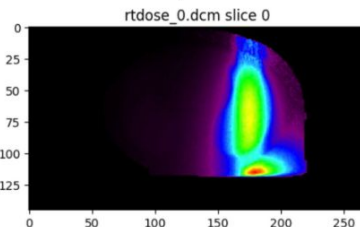
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(0008, 0013) Instance Creation Time         TM: '181333'
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(0008, 0016) SOP Class UID                  UI: 'RT Plan ST'
(0008, 0018) SOP Instance UID                UI: '1.16.840.1.114337.1.1.168209365.0'
(0008, 0020) Study Date                     DA: '20221213'
(0008, 0030) Study Time                     TM: '130930'
(0008, 0050) Accession Number                SH: ''
(0008, 0060) Modality                       CS: 'RTPLAN'
(0008, 0070) Manufacturer                   LO: 'CMS, Inc.'

(0008, 0005) Specific Character Set          CS: 'ISO_IR 100'
(0008, 0012) Instance Creation Date         DA: '20230421'
(0008, 0013) Instance Creation Time         TM: '181335'
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(0008, 0016) SOP Class UID                  UI: 'RT Structure Set Storage'
(0008, 0018) SOP Instance UID                UI: '1.16.840.1.114337.1.1.168209365.0'
(0008, 0020) Study Date                     DA: '20221213'
(0008, 0030) Study Time                     TM: '130930'
(0008, 0050) Accession Number                SH: '1'
(0008, 0060) Modality                       CS: 'RTSTRUCT'
(0008, 0070) Manufacturer                   LO: 'Computerized Medical Systems'

```

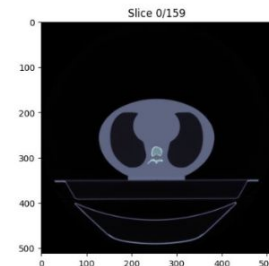


**EPID**  
4 images in dicom  
format  
(1024x1024)  
16 bit



**RTDOSE**  
5 images, 79 frames for  
image in dicom format  
(145x268), 16 bit

**CT**  
**159** images in dicom  
format  
(512x512), 16 bit



**DATASET:**  
[https://drive.google.com/drive/folders/1sfwZSsa1BGuNRctZqY\\_K5z6v/rdQZAug?usp=drive\\_link](https://drive.google.com/drive/folders/1sfwZSsa1BGuNRctZqY_K5z6v/rdQZAug?usp=drive_link)

# Coregistrazione

## How does 3D Slicer overlay the RTDOSE and Image data in the correct position?

Support python dicom extensions-manager registration slicerrt



Iman\_Shokatian

1 Oct 2019

Hello,

I'm currently an MS student in Medical Physics and I have a great need to be able to overlay an isodose distribution from an RTDOSE file onto a CT image from a .dcm file set.

I've managed to extract the image and the dose pixel arrays myself using pydicom and dicom\_numpy, but the two arrays are not the same size! So, if I overlay the two together, the dose will not be in the correct position based on what the Elekta Gamma Plan software exported it as.

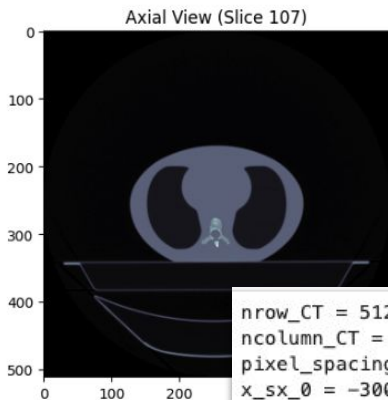
I've played around with slicerrt and it obviously is able to do this even though the arrays are not the same size. However, I think I cannot export the numerical data when using slicerrt...I can only scroll through and view it as an image. What section of the code has the algorithm for overlaying the RTDOSE to an image?

Oct 2019

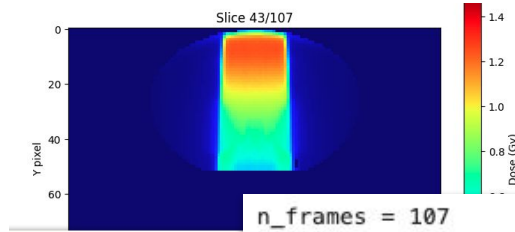
1 / 23

Oct 2019

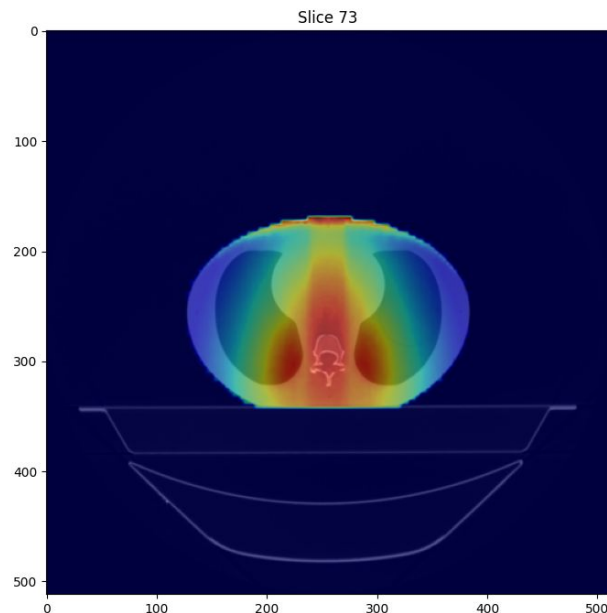
Sep 2023



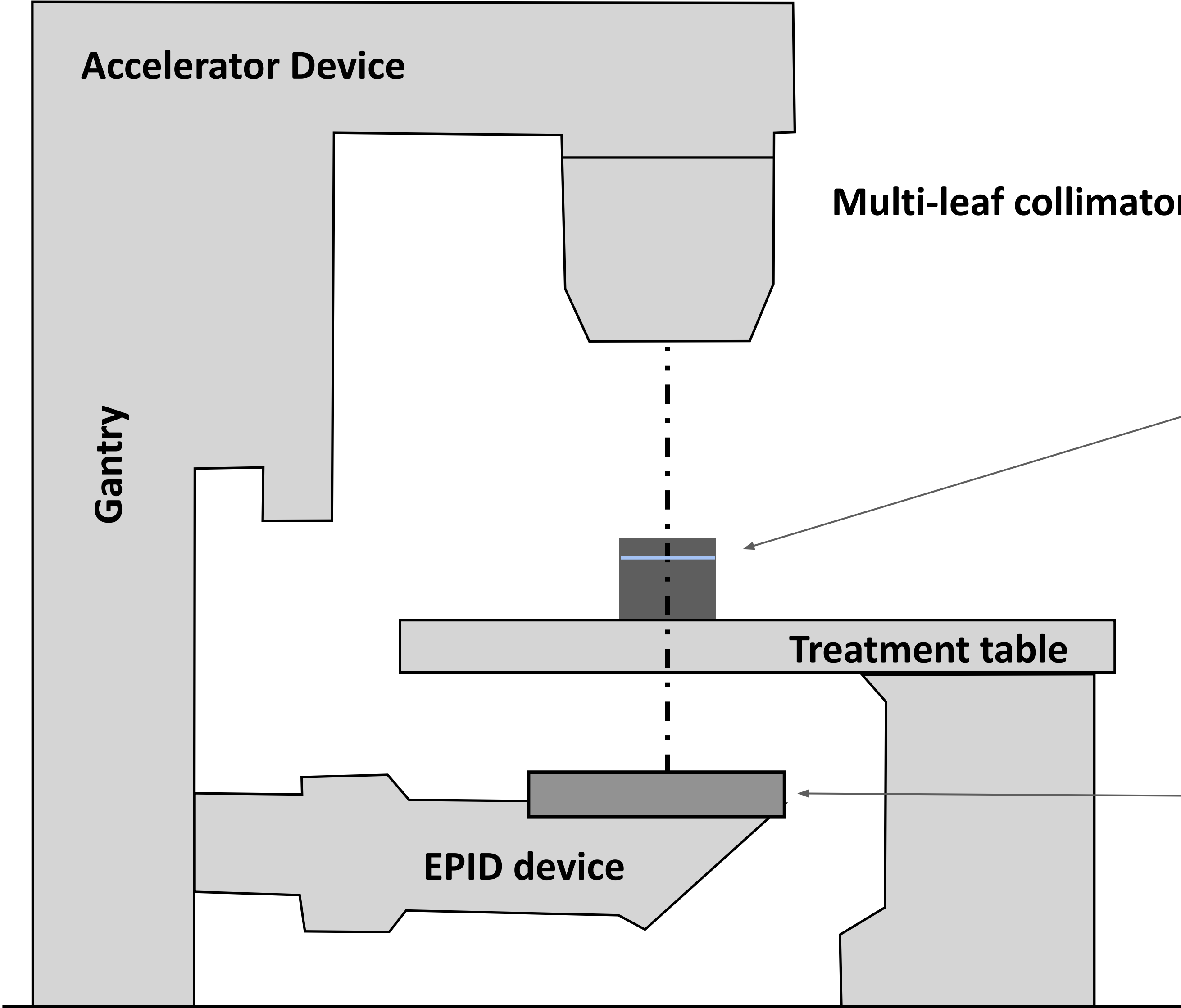
```
nrow_CT = 512
ncolumn_CT = 512
pixel_spacing_CT = [1.171875, 1.171875] mm
x_sx_0 = -300 mm
y_sx_0 = -207 mm
z_0 = -220 mm
```



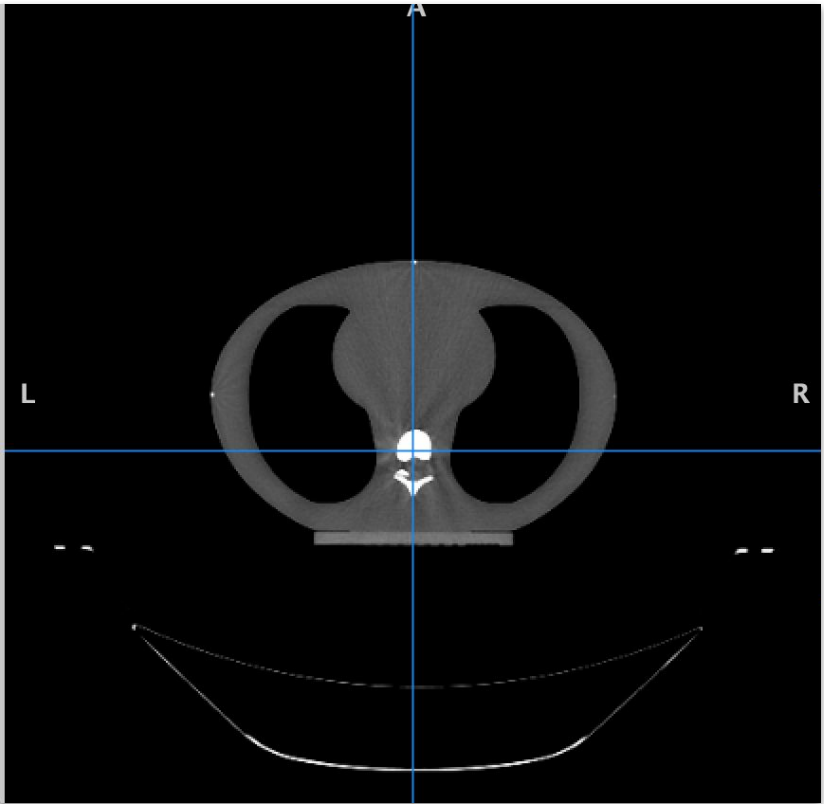
```
n_frames = 107
rows = 95
columns = 135
pixel_spacing = [4, 4] mm
x_sx_0 = -268.6 mm
y_sx_0 = -11.6 mm
z_sx_0 = -219 mm
```



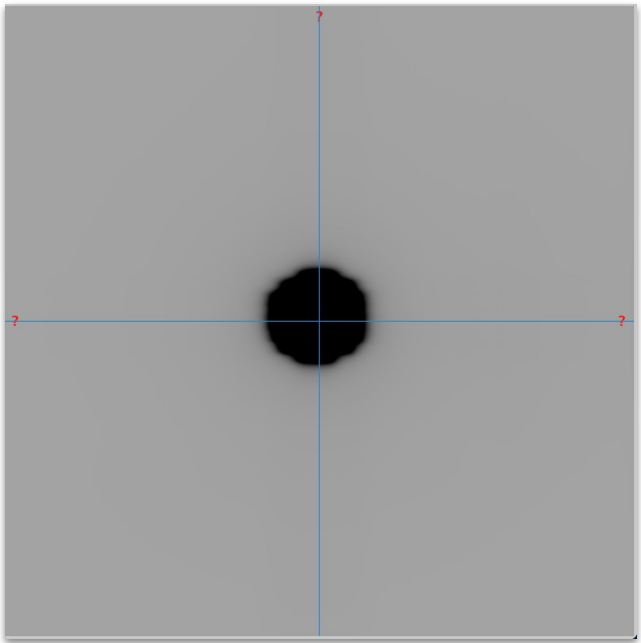
# Experimental setup



Phantom CT

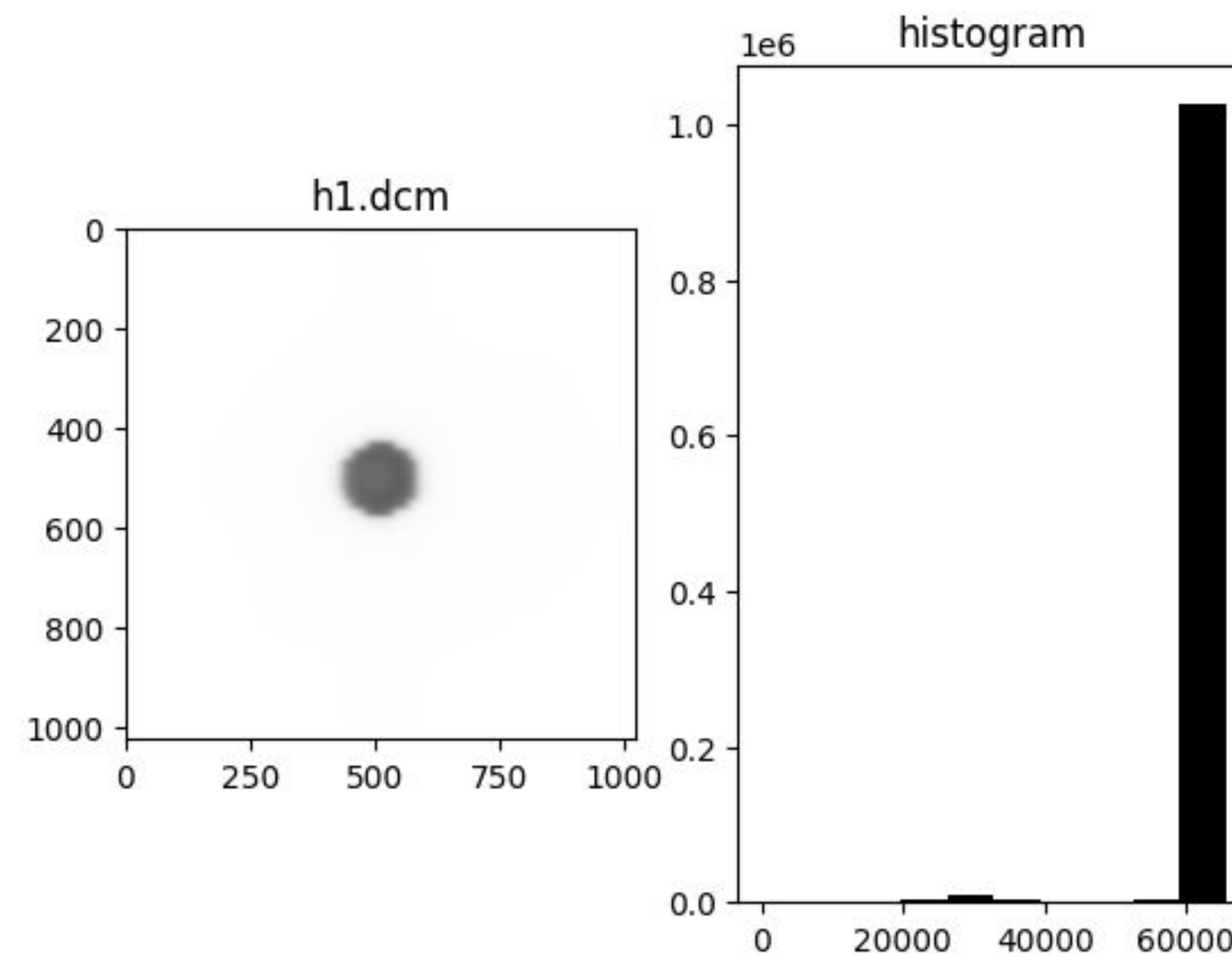


EPID image





(0020, 0011) Series Number	IS: '10517'
(0020, 0013) Instance Number	IS: '10521'
(0020, 0020) Patient Orientation	CS: ''
(0020, 1002) Images in Acquisition	IS: '1'
(0021, 0010) Private Creator	LO: 'Elekta: Portal image'
(0021, 1002) Private tag data	DS: '0.0937319'
(0028, 0002) Samples per Pixel	US: 1
(0028, 0004) Photometric Interpretation	CS: 'MONOCHROME2'
(0028, 0010) Rows	US: 1024
(0028, 0011) Columns	US: 1024
(0028, 0100) Bits Allocated	US: 16
(0028, 0101) Bits Stored	US: 16
(0028, 0102) High Bit	US: 15
(0028, 0103) Pixel Representation	US: 0
(0028, 1050) Window Center	DS: '62234.3'
(0028, 1051) Window Width	DS: '22555.1'
(3002, 0002) RT Image Label	SH: 'iViewPortalImage'
(3002, 000a) Reported Values Origin	CS: 'OPERATOR'
(3002, 000c) RT Image Plane	CS: 'NORMAL'
(3002, 000e) X-Ray Image Receptor Angle	DS: '0.0'
(3002, 0011) Image Plane Pixel Spacing	DS: [0.405, 0.405]
(3002, 0012) RT Image Position	DS: None
(3002, 0020) Radiation Machine Name	SH: ''
(3002, 0022) Radiation Machine SAD	DS: '1000.0'
(3002, 0024) Radiation Machine SSD	DS: '0.0'
(3002, 0026) RT Image SID	DS: '1600.0'
(300a, 00b3) Primary Dosimeter Unit	CS: ''
(300a, 011e) Gantry Angle	DS: '-0.1'
(7fe0, 0010) Pixel Data	OW: Array of 2097152 elements

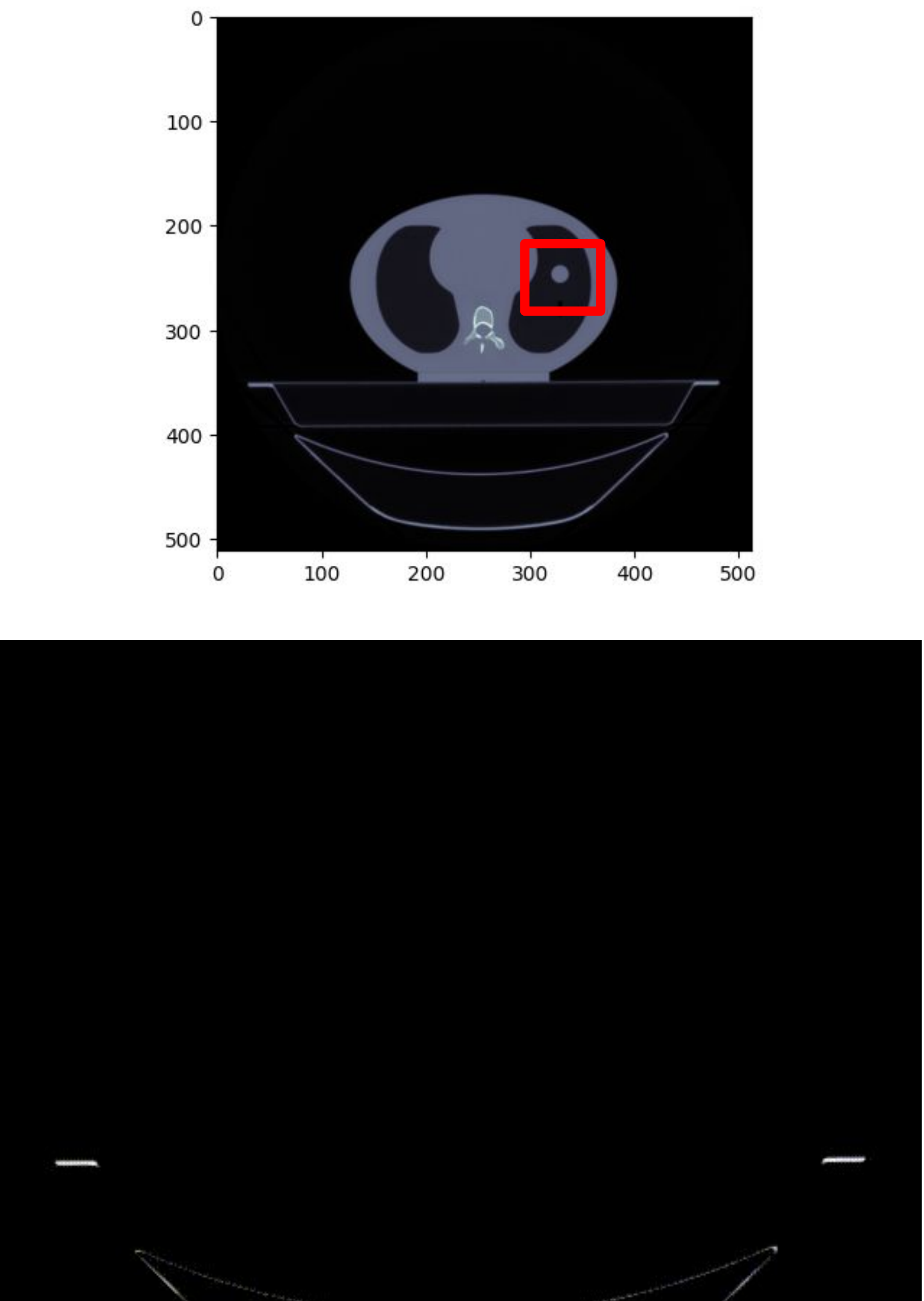


See: [DICOM Standard Browser](#)



# CT images

```
Dataset.file_meta -----
(0002, 0000) File Meta Information Group Length  UL: 174
(0002, 0001) File Meta Information Version      OB: b'\x00\x01'
(0002, 0002) Media Storage SOP Class UID        UI: CT Image Storage
(0002, 0003) Media Storage SOP Instance UID     UI: 1.2.840.113704.1.111.2344.1669205565.21376
(0002, 0010) Transfer Syntax UID               UI: Explicit VR Little Endian
(0002, 0012) Implementation Class UID          UI: 2.16.840.1.114362.1
(0002, 0013) Implementation Version Name       SH: 'MIM723M31402'
-----
(0008, 0005) Specific Character Set            CS: 'ISO IR 100'
(0008, 0008) Image Type                       CS: ['ORIGINAL', 'PRIMARY', 'AXIAL', 'HELIX']
(0008, 0012) Instance Creation Date           DA: '20221123'
(0008, 0013) Instance Creation Time           TM: '131245'
(0008, 0016) SOP Class UID                    UI: CT Image Storage
(0008, 0018) SOP Instance UID                 UI: 1.2.840.113704.1.111.2344.1669205565.21376
(0008, 0020) Study Date                      DA: '20221123'
(0008, 0022) Acquisition Date                DA: '20221123'
(0008, 0023) Content Date                   DA: '20221123'
(0008, 002a) Acquisition DateTime            DT: '20221123131232+0100'
(0008, 0030) Study Time                      TM: '130930'
(0008, 0032) Acquisition Time                TM: '131232'
(0008, 0033) Content Time                    TM: '131233.985001'
(0008, 0050) Accession Number                SH: ''
(0008, 0060) Modality                       CS: 'CT'
(0008, 0070) Manufacturer                   LO: 'Philips'
(0008, 0080) Institution Name                LO: 'A.O.U. CAREGGI      FI'
(0008, 0081) Institution Address             ST: 'FIRENZE'
(0008, 0090) Referring Physician's Name     PN: ''
(0008, 1010) Station Name                   SH: 'HOST-7424'
(0008, 1030) Study Description               LO: '1cm sotto SW'
(0008, 103e) Series Description              LO: 'TORACE 5 MM'
(0008, 1040) Institutional Department Name   LO: 'Radioterapia DEAS_1'
(0008, 1070) Operators' Name                PN: ''
(0008, 1090) Manufacturer's Model Name      LO: 'Brilliance Big Bore'
(0008, 1140) Referenced Image Sequence     1 item(s) ----
```



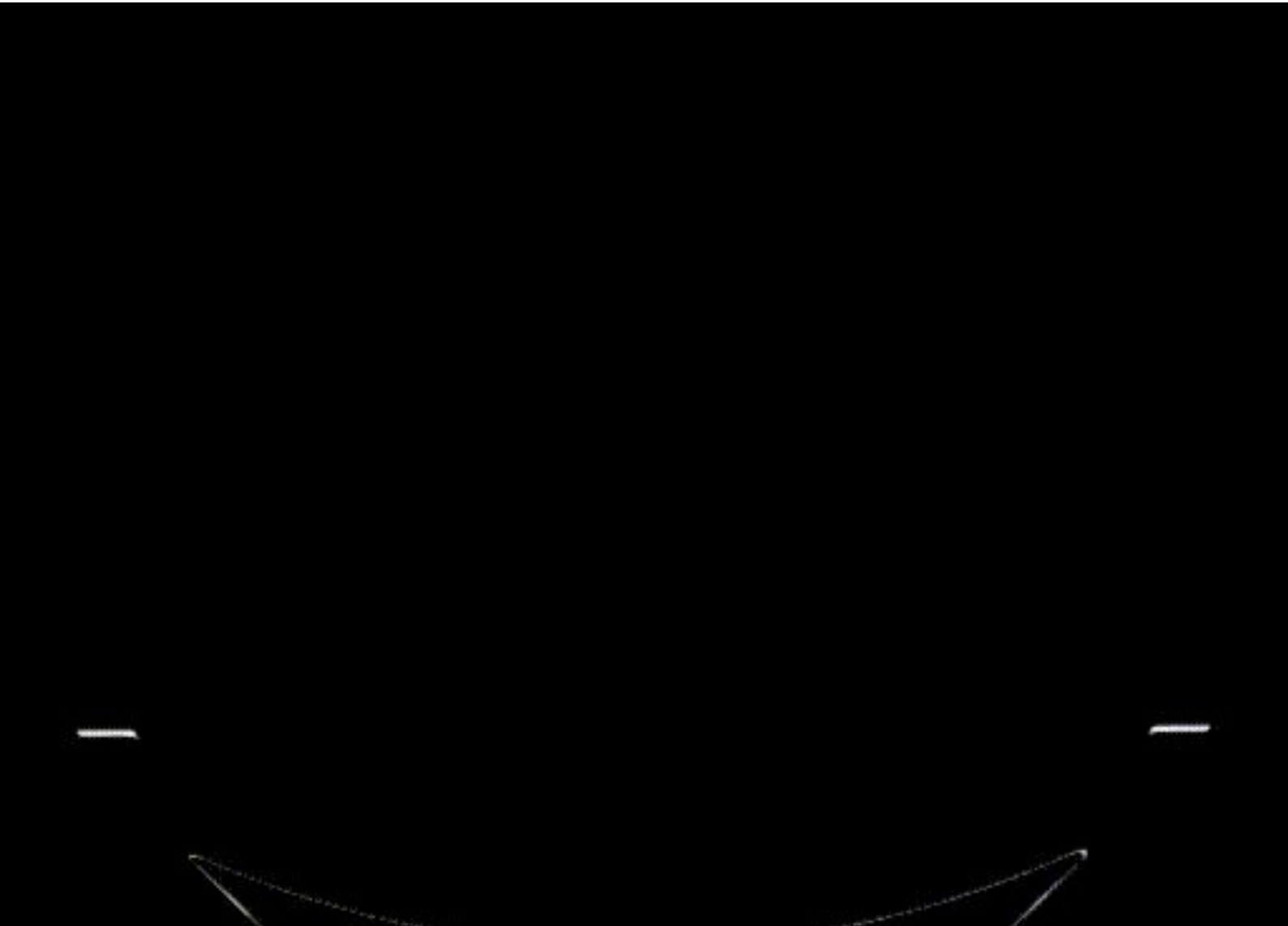
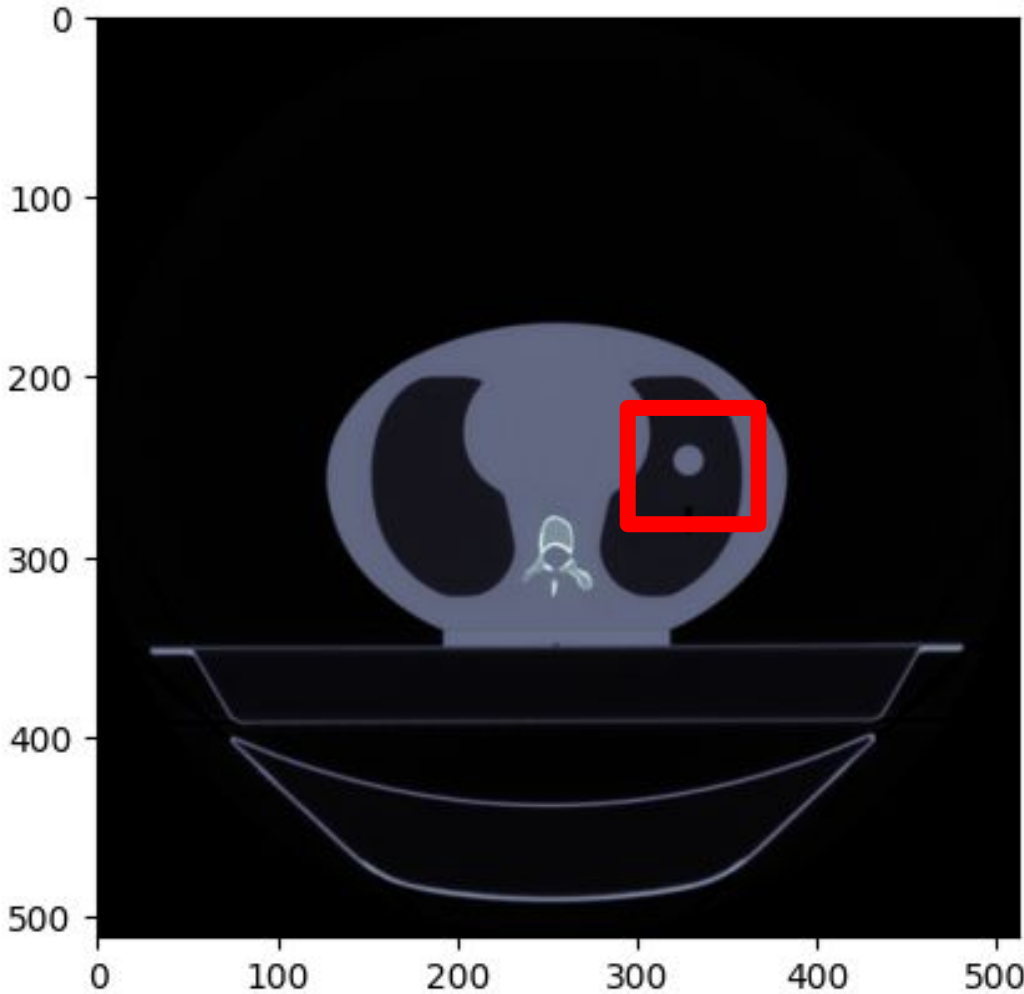


# CT images



(0008, 1150) Referenced SOP Class UID  
(0008, 1155) Referenced SOP Instance UID  
-----  
(0008, 2111) Derivation Description  
1, Point Transform = 0, compression ratio = 3.867400'  
(0010, 0010) Patient's Name  
(0010, 0020) Patient ID  
(0010, 0030) Patient's Birth Date  
(0010, 0040) Patient's Sex  
(0010, 1010) Patient's Age  
(0010, 2000) Medical Alerts  
(0018, 0022) Scan Options  
(0018, 0050) Slice Thickness  
(0018, 0060) KVP  
(0018, 0088) Spacing Between Slices  
(0018, 0090) Data Collection Diameter  
(0018, 1020) Software Versions  
(0018, 1030) Protocol Name  
(0018, 1100) Reconstruction Diameter  
(0018, 1120) Gantry/Detector Tilt  
(0018, 1130) Table Height  
(0018, 1140) Rotation Direction  
(0018, 1150) Exposure Time  
(0018, 1151) X-Ray Tube Current  
(0018, 1152) Exposure  
(0018, 1160) Filter Type  
(0018, 1210) Convolution Kernel  
(0018, 5100) Patient Position  
(0018, 9321) CT Exposure Sequence 1 item(s) ----  
  (0018, 9323) Exposure Modulation Type  
  (0018, 9324) Estimated Dose Saving  
  (0018, 9328) Exposure Time in ms  
  (0018, 9330) X-Ray Tube Current in mA  
  (0018, 9332) Exposure in mAs  
  (0018, 9345) CTDIvol

UI: CT Image Storage  
UI: 1.2.840.113704.1.111.2344.1669205515.21357  
  
ST: '[EMGDCM] Lossless Compression (JPEG p14), Selection Value =  
PN: 'CIRS Invivo statico^1cm sotto SW'  
LO: 'CIRSin vivo'  
DA: '20021122'  
CS: '0'  
AS: '020Y'  
LO: ''  
CS: 'HELIX'  
DS: '2.0'  
DS: '120.0'  
DS: '1.0'  
DS: '600.0'  
LO: '2.3.5'  
LO: 'TORACE 5mm/OncoBody'  
DS: '600.0'  
DS: '0.0'  
DS: '173.0'  
CS: 'CW'  
IS: '1230'  
IS: '317'  
IS: '390'  
SH: 'C'  
SH: 'C'  
CS: 'HFS'  
CS: 'ZDOM'  
FD: 0.0  
FD: 1.2302839116719242  
FD: 317.0  
FD: 390.0  
FD: 20.6

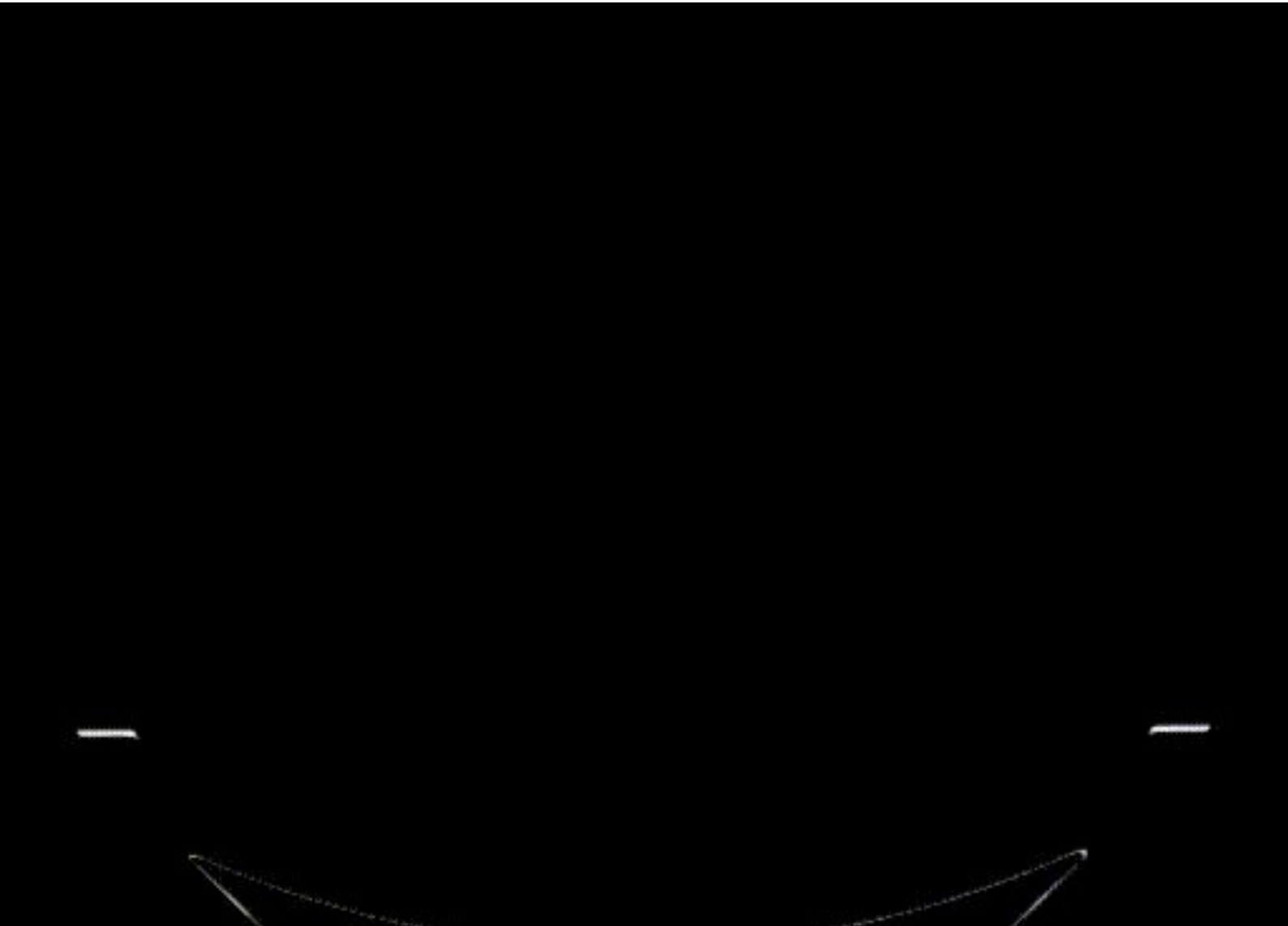
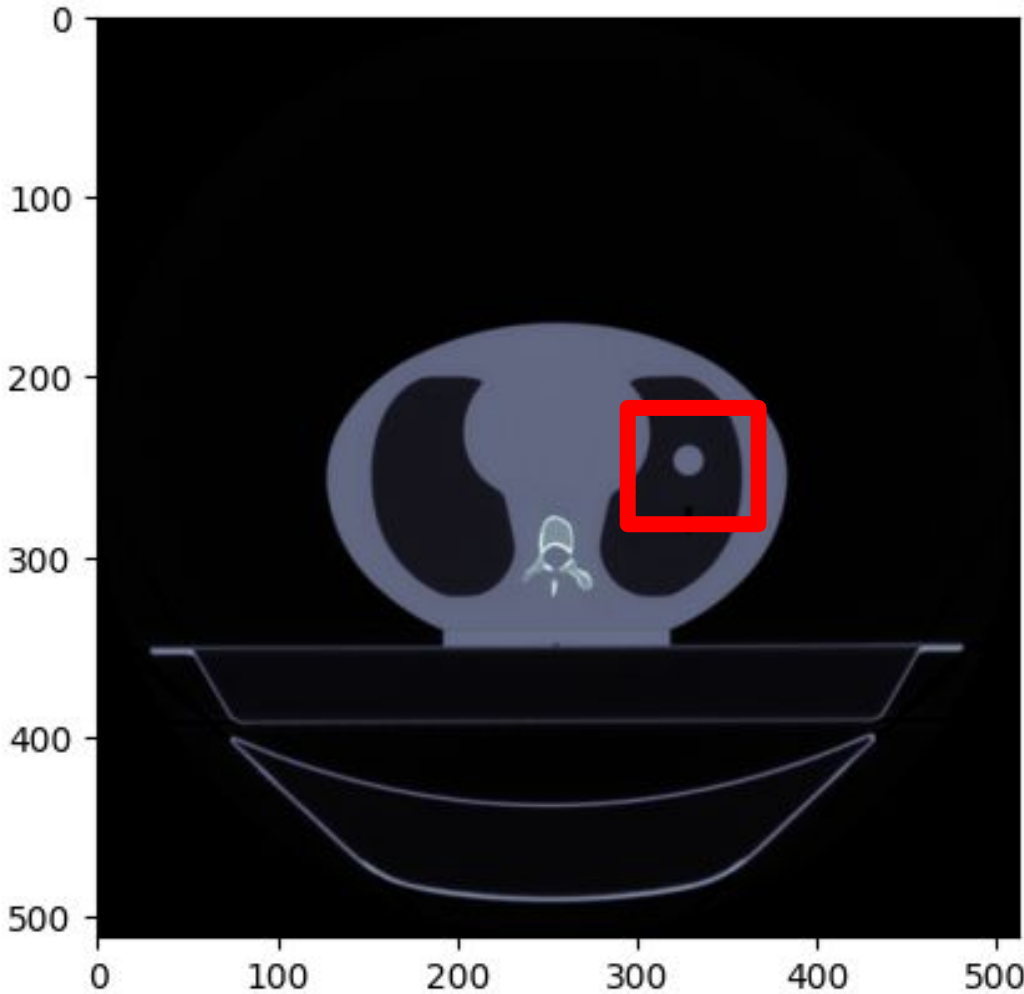




# CT images



(0018, 9323)	Exposure Modulation Type	CS: 'ZDOM'
(0018, 9324)	Estimated Dose Saving	FD: 0.0
(0018, 9345)	CTDIvol	FD: 20.6
(0020, 000d)	Study Instance UID	UI: 1.2.840.113704.1.111.3464.1669205363.54
(0020, 000e)	Series Instance UID	UI: 1.2.840.113704.1.111.6596.1669205536.6
(0020, 0010)	Study ID	SH: '56000'
(0020, 0011)	Series Number	IS: '2'
(0020, 0012)	Acquisition Number	IS: None
(0020, 0013)	Instance Number	IS: '19'
(0020, 0032)	Image Position (Patient)	DS: [-300, -218, -658]
(0020, 0037)	Image Orientation (Patient)	DS: [1, 0, 0, 0, 1, 0]
(0020, 0052)	Frame of Reference UID	UI: 1.2.840.113704.1.111.6596.1669205471.3
(0020, 0060)	Laterality	CS: ''
(0020, 1040)	Position Reference Indicator	LO: ''
(0020, 1041)	Slice Location	DS: '-658.0'
(0020, 4000)	Image Comments	LT: 'TORACE 5 MM'
(0028, 0002)	Samples per Pixel	US: 1
(0028, 0004)	Photometric Interpretation	CS: 'MONOCHROME2'
(0028, 0010)	Rows	US: 512
(0028, 0011)	Columns	US: 512
(0028, 0030)	Pixel Spacing	DS: [1.171875, 1.171875]
(0028, 0100)	Bits Allocated	US: 16
(0028, 0101)	Bits Stored	US: 12
(0028, 0102)	High Bit	US: 11
(0028, 0103)	Pixel Representation	US: 0
(0028, 1050)	Window Center	DS: [00050, 00050]
(0028, 1051)	Window Width	DS: [00350, 00350]
(0028, 1052)	Rescale Intercept	DS: '-1024.0'
(0028, 1053)	Rescale Slope	DS: '1.0'
(0032, 1070)	Requested Contrast Agent	LO: ''
(0040, 0012)	Pre-Medication	LO: ''
(0040, 0253)	Performed Procedure Step ID	SH: '5600096'
(00e1, 0010)	Private Creator	LO: 'ELSCINT1'
(00e1, 1001)	[Data Dictionary Version]	US: 1
(00e1, 1022)	[Presentation Relative Center]	DS: [00000, 00000]
(00e1, 1023)	[Presentation Relative Part]	DS: [00001, 00001]
(00e1, 1040)	[Image Label]	SH: 'TORACE 5 MM'
(00e1, 1042)	[Unknown]	LO: '453567-129831'

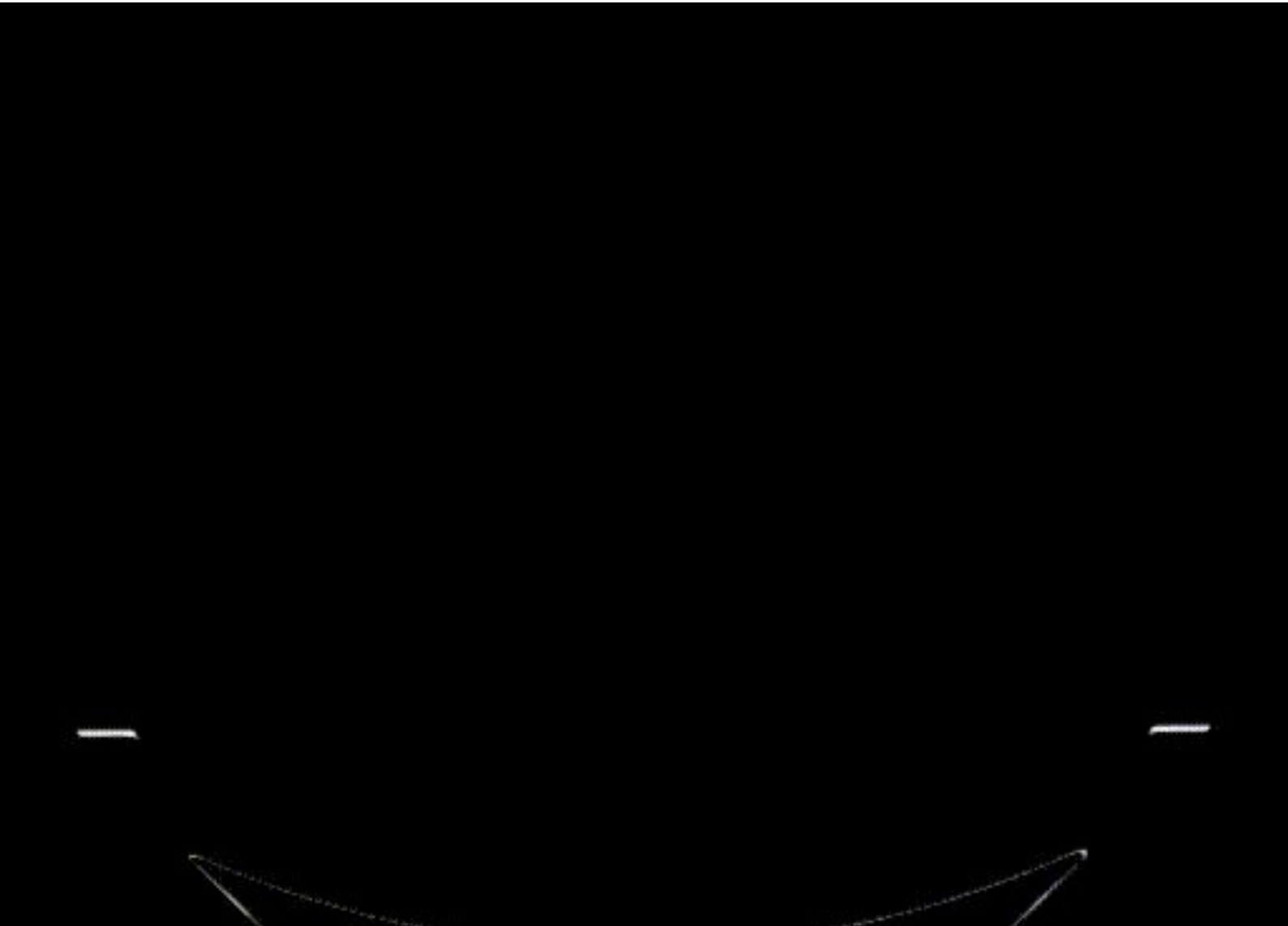
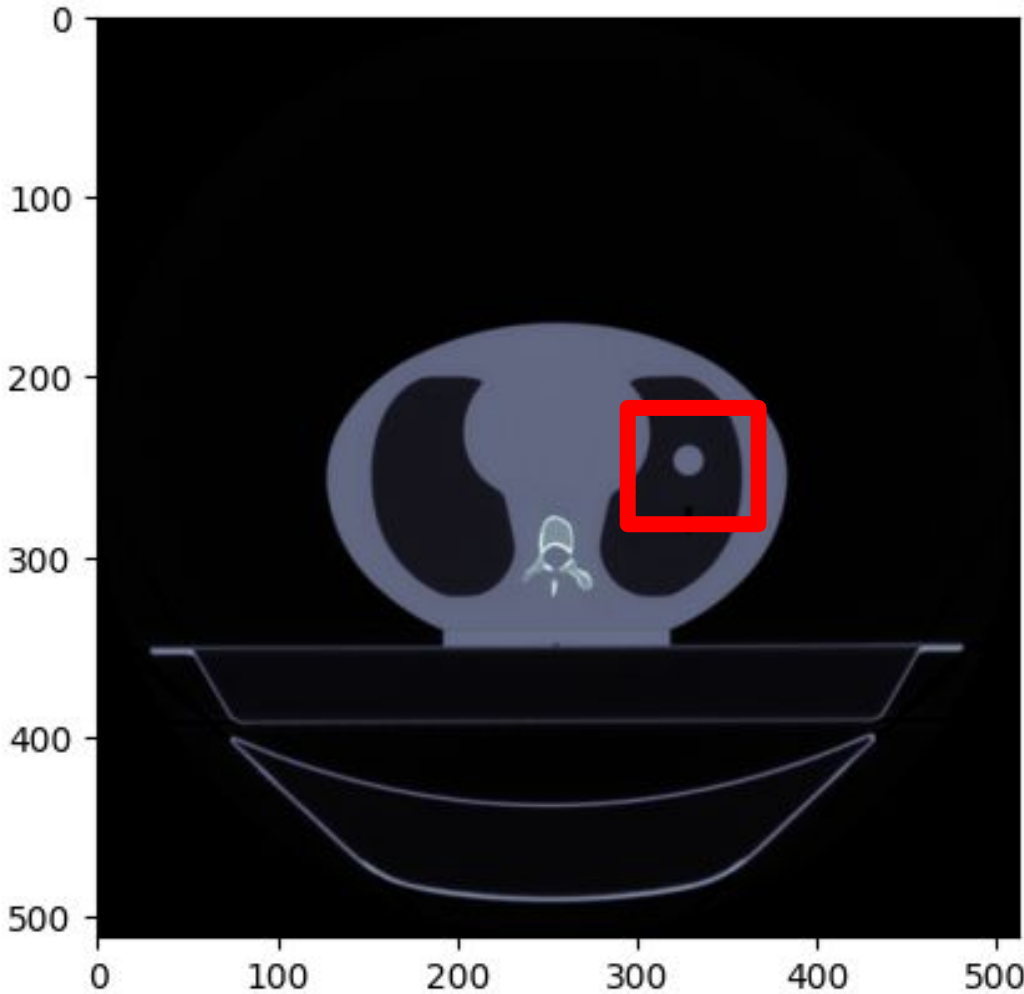




# CT images



(00e1, 1050)	[Acquisition Duration]	DS: '10.219'
(00e1, 1061)	[Protocol File Name]	LO: 'OncoBody_Multi_1063_usr.proc'
(00e1, 1063)	[Patient Language]	SH: 'ITALIAN'
(01e1, 0010)	Private Creator	LO: 'ELSCINT1'
(01e1, 1041)	Private tag data	UN: Array of 128 elements
(01f1, 0010)	Private Creator	LO: 'ELSCINT1'
(01f1, 1001)	[Acquisition Type]	CS: 'SPIRAL'
(01f1, 1002)	[Unknown]	CS: 'STANDARD'
(01f1, 1003)	[Concurrent Slices Generation]	CS: 'E'
(01f1, 1004)	[Angular Sampling Density]	CS: 'NORMAL'
(01f1, 1005)	[Reconstruction Arc]	DS: '180.0'
(01f1, 1007)	[Table Velocity]	DS: '19.5'
(01f1, 1008)	[Acquisition Length]	DS: '160.0'
(01f1, 100a)	[Edge Enhancement Weight]	US: 0
(01f1, 100c)	[Scanner Relative Center]	DS: [0, 0]
(01f1, 100d)	[Rotation Angle]	DS: '0.0'
(01f1, 100e)	[Unknown]	FL: 0.0
(01f1, 1026)	[Pitch]	DS: '0.813'
(01f1, 1027)	[Rotation Time]	DS: '1.0'
(01f1, 1032)	[Image View Convention]	CS: 'RIGHT_ON_LEFT'
(01f1, 1042)	[Unknown]	SH: 'No'
(01f1, 1044)	[Unknown]	OW: Array of 644 elements
(01f1, 1046)	[Unknown]	FL: 1.5
(01f1, 1047)	[Unknown]	SH: '3D'
(01f1, 1049)	[Unknown]	DS: '390.0'
(01f1, 104a)	[Unknown]	SH: 'DOM'
(01f1, 104b)	[Unknown]	SH: '16x1.5'
(01f1, 104c)	[Unknown]	SH: 'NO'
(01f1, 104d)	[Unknown]	SH: 'YES'
(01f1, 104e)	[Unknown]	LO: 'Body'
(01f7, 0010)	Private Creator	LO: 'ELSCINT1'
(01f7, 1010)	[Unknown]	OB: None
(01f7, 1011)	[Unknown]	OW: Array of 552 elements
(01f7, 1013)	[Unknown]	OW: Array of 136 elements
(01f7, 1014)	[Unknown]	OW: Array of 108 elements
(01f7, 1015)	[Unknown]	OW: Array of 188 elements
(01f7, 1016)	[Unknown]	OW: Array of 40 elements
(01f7, 1017)	[Unknown]	OW: b'\x00\x00\x00\x00\x00\x00\x00'
(01f7, 1018)	[Unknown]	OW: Array of 228 elements

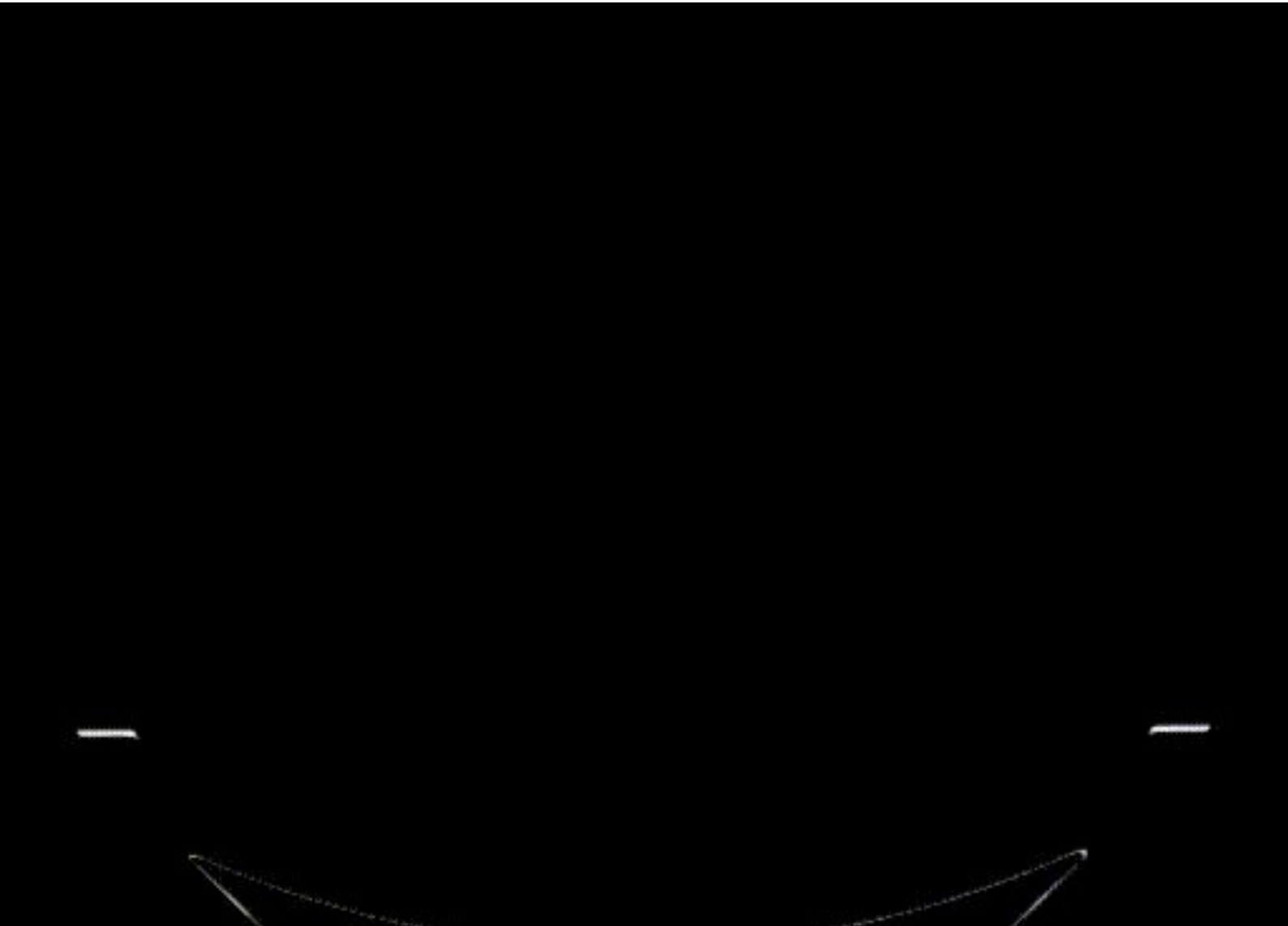
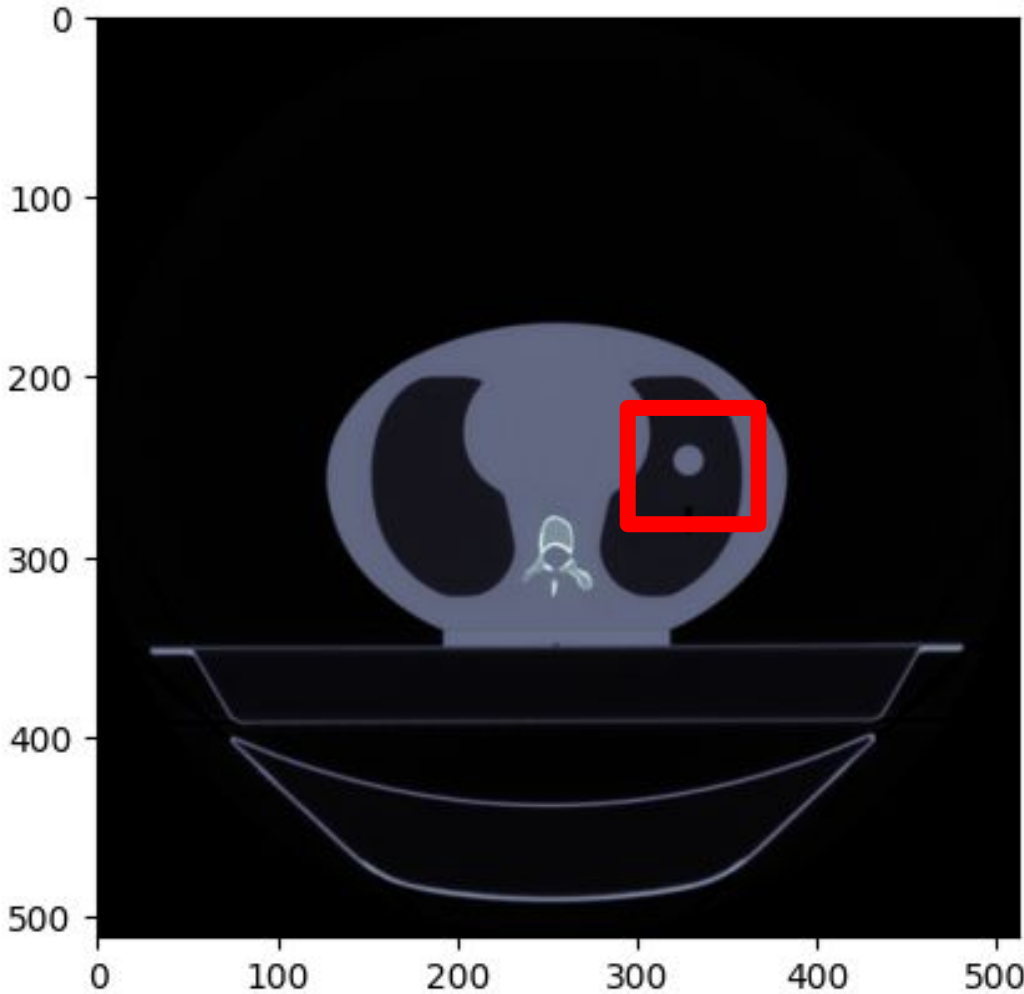




# CT images



(01f7, 1019) [Unknown]	OW: Array of 2160 elements
(01f7, 101a) [Unknown]	OW: Array of 28 elements
(01f7, 101b) [Unknown]	OW: Array of 1744 elements
(01f7, 101c) [Unknown]	OW: Array of 116 elements
(01f7, 101e) [Unknown]	OW: Array of 364 elements
(01f7, 101f) [Unknown]	OW: Array of 148 elements
(01f7, 1022) [Unknown]	UI: 1.2.840.113704.1.111.6596.1669205536.5.1.111111111111111111
(01f7, 1023) [Unknown]	OW: b'.\x00\x00\x00'
(01f7, 1025) [Unknown]	OW: b'\x01\x00\x00\x00\x00\x00\x00\x00p\xf2\x12\x00'
(01f7, 1026) [Unknown]	OW: Array of 10764 elements
(01f7, 1027) [Unknown]	OW: Array of 36 elements
(01f7, 1029) [Unknown]	OW: Array of 440 elements
(01f7, 102b) [Unknown]	OW: Array of 36 elements
(01f7, 102c) [Unknown]	OW: Array of 656 elements
(01f7, 102d) [Unknown]	OW: b'\x00\x00\x00\x00\x00\x00\x00\x00'
(01f7, 1030) [Unknown]	OW: b'\x00\x00\x00\x00\x00\x00\x00\x00\x00\x00\x00\x00\x00\x00\x00\x00'
(01f7, 1031) [Unknown]	OW: Array of 124 elements
(01f7, 1070) [Unknown]	OW: Array of 1312 elements
(01f7, 1074) [Unknown]	OW: Array of 288 elements
(01f7, 1075) [Unknown]	OW: Array of 136 elements
(01f7, 107f) Private tag data	UN: Array of 1312 elements
(07a1, 0010) Private Creator	L0: 'ELSCINT1'
(07a1, 1010) [Tamar Software Version]	L0: '3.5'
(7fdf, 0010) Private Creator	L0: 'ELSCINT1'
(7fdf, 10f0) [Unknown]	OB: None
(7fdf, 10ff) [Unknown]	SH: ''
(7fe0, 0010) Pixel Data	OW: Array of 524288 elements

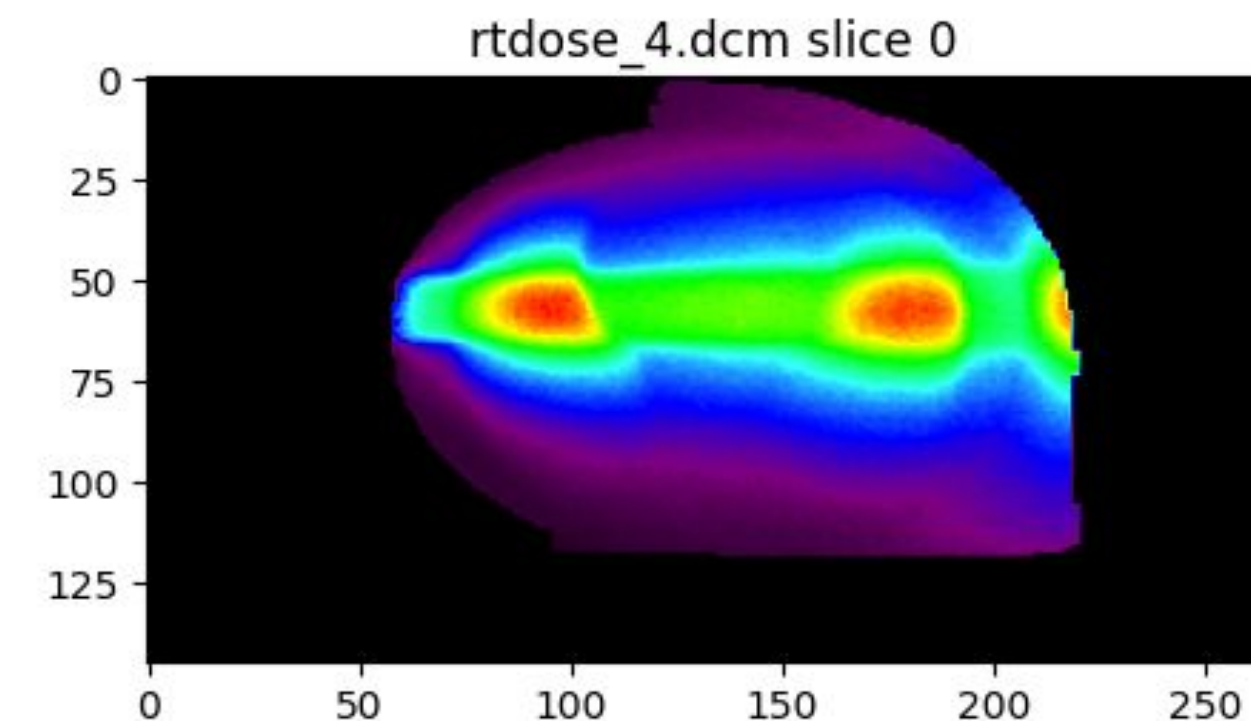
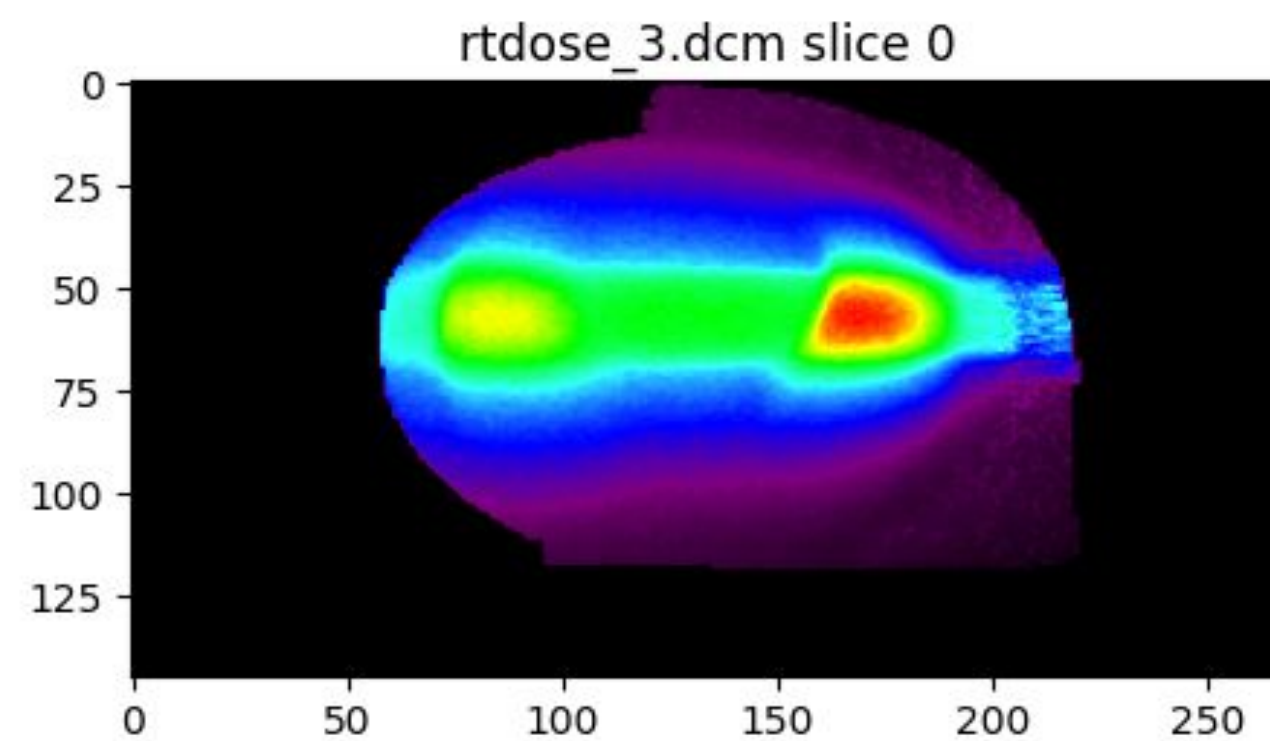
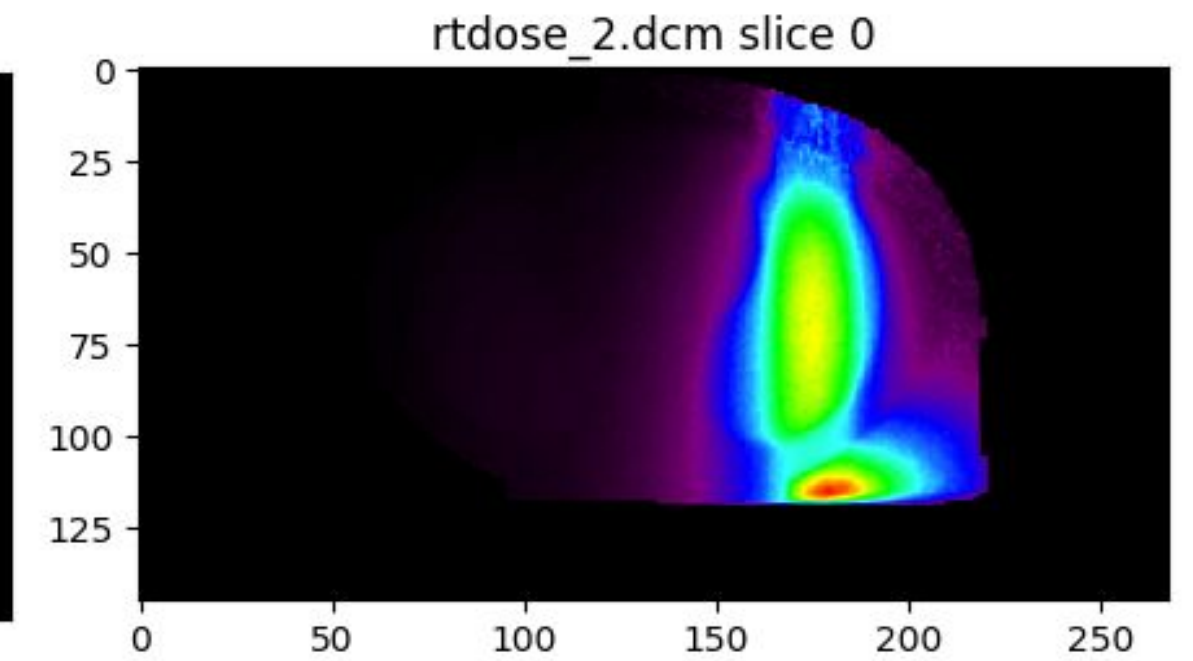
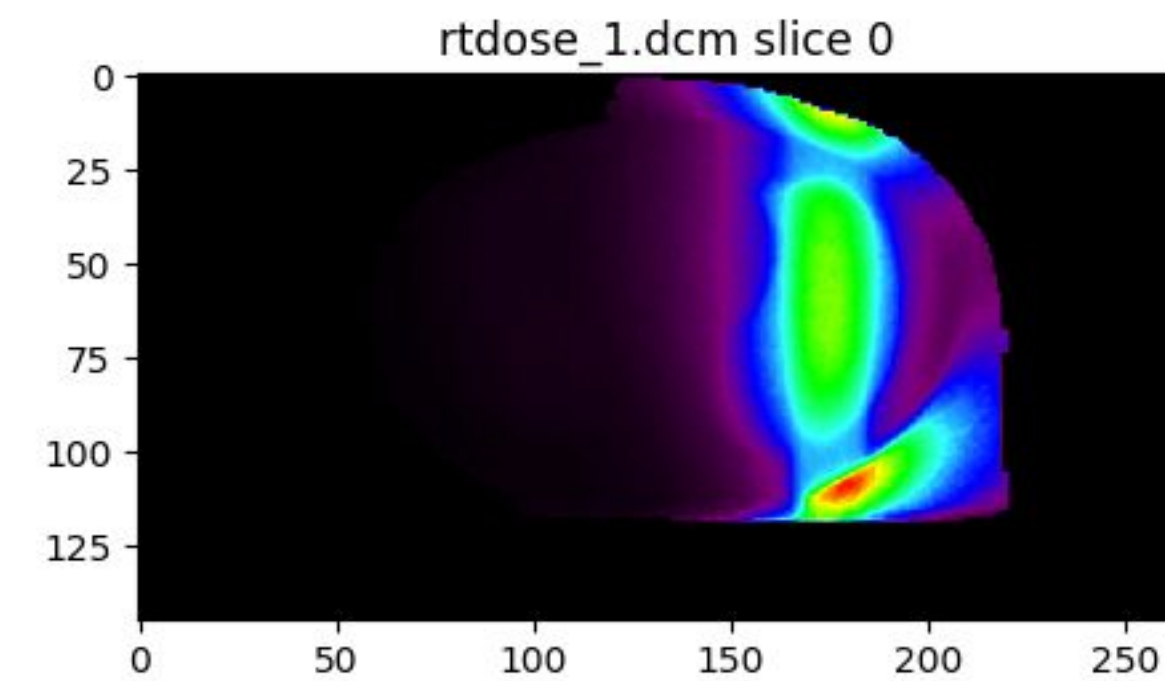
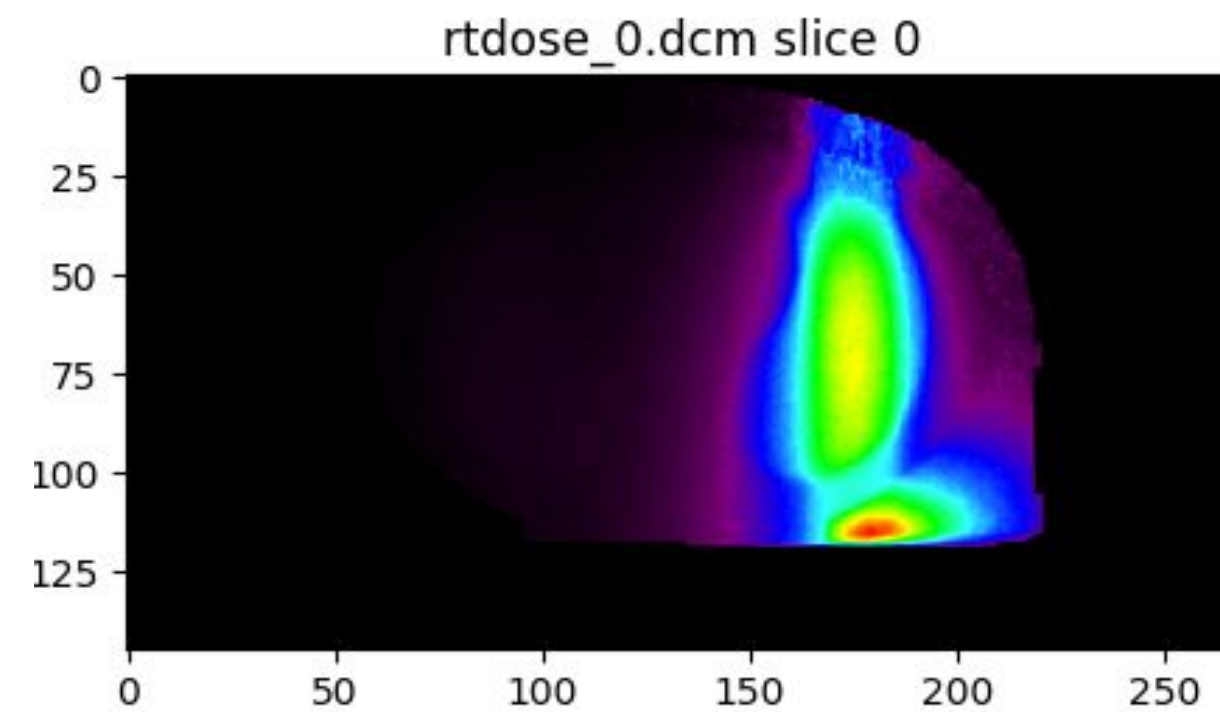




# RT dose data

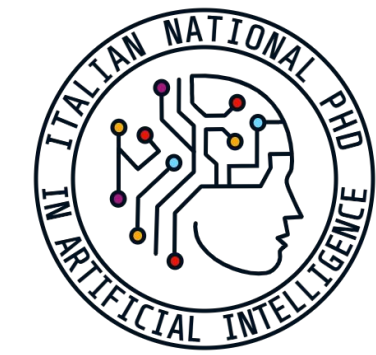
- CT
- EPID
- RT
- RTDOSE**
- RTPLAN

**Multi-frame dose images** → the dose distribution is encoded as a multi-frame image (79 frames)





# RT dose file: tags & attributes



```
Dataset.file_meta -----
(0002, 0000) File Meta Information Group Length  UL: 168
(0002, 0001) File Meta Information Version       OB: b'\x00\x01'
(0002, 0002) Media Storage SOP Class UID        UI: RT Dose Storage
(0002, 0003) Media Storage SOP Instance UID     UI: 2.16.840.1.114337.1.1.1682093615.0
(0002, 0010) Transfer Syntax UID               UI: Implicit VR Little Endian
(0002, 0012) Implementation Class UID          UI: 2.16.840.1.114362.1
(0002, 0013) Implementation Version Name       SH: 'MIM723M31402'
-----
(0008, 0005) Specific Character Set             CS: 'ISO_IR 100'
(0008, 0016) SOP Class UID                     UI: RT Dose Storage
(0008, 0018) SOP Instance UID                  UI: 2.16.840.1.114337.1.1.1682093615.0
(0008, 0020) Study Date                       DA: '20221123'
(0008, 0030) Study Time                       TM: '130930'
(0008, 0050) Accession Number                 SH: ''
(0008, 0060) Modality                         CS: 'RTDOSE'
(0008, 0070) Manufacturer                    LO: 'CMS, Inc.'
(0008, 0090) Referring Physician's Name       PN: ''
(0008, 1090) Manufacturer's Model Name        LO: 'Monaco'
(0010, 0010) Patient's Name                   PN: 'CIRS Invivo statico^1cm sotto SW'
(0010, 0020) Patient ID                      LO: 'CIRSin vivo'
(0010, 0030) Patient's Birth Date             DA: '20021122'
(0010, 0040) Patient's Sex                   CS: '0'
(0018, 0050) Slice Thickness                  DS: None
(0018, 1020) Software Versions                LO: '5.51.10'
(0020, 000d) Study Instance UID               UI: 1.2.840.113704.1.111.3464.1669205363.54
(0020, 000e) Series Instance UID             UI: 2.16.840.1.114337.1.1.1682093613.0.1.3
(0020, 0010) Study ID                        SH: '56000_6'
(0020, 0011) Series Number                   IS: None
(0020, 0013) Instance Number                 IS: '1'
(0020, 0032) Image Position (Patient)         DS: [-265.3, -41.9, -675]
(0020, 0037) Image Orientation (Patient)      DS: [1, 0, 0, 0, 1, 0]
(0020, 0052) Frame of Reference UID          UI: 1.2.840.113704.1.111.6596.1669205471.3
(0020, 1040) Position Reference Indicator     LO: ''
(0028, 0002) Samples per Pixel               US: 1
(0028, 0004) Photometric Interpretation       CS: 'MONOCHROME2'
```

```
(0028, 0008) Number of Frames                TS: '79'
(0028, 0009) Frame Increment Pointer          AT: (3004, 000c)
(0028, 0010) Rows                           US: 145
(0028, 0011) Columns                        US: 268
(0028, 0030) Pixel Spacing                  DS: [2, 2]
(0028, 0100) Bits Allocated                 US: 16
(0028, 0101) Bits Stored                    US: 16
(0028, 0102) High Bit                       US: 15
(0028, 0103) Pixel Representation           US: 0
(0028, 0106) Smallest Image Pixel Value     US: 0
(0028, 0107) Largest Image Pixel Value      US: 65535
(3004, 0002) Dose Units                      CS: 'GY'
(3004, 0004) Dose Type                      CS: 'PHYSICAL'
(3004, 000a) Dose Summation Type            CS: 'BEAM'
(3004, 000c) Grid Frame Offset Vector        DS: Array of 79 elements
(3004, 000e) Dose Grid Scaling               DS: '9.53212e-06'
(3004, 0014) Tissue Heterogeneity Correction CS: 'ROI_OVERRIDE'
(300c, 0002) Referenced RT Plan Sequence 1 item(s) ----
  (0008, 1150) Referenced SOP Class UID      UI: RT Plan Storage
  (0008, 1155) Referenced SOP Instance UID   UI: 2.16.840.1.114337.1.1.1682093613.0
  (300c, 0020) Referenced Fraction Group Sequence 1 item(s) ----
    (300c, 0004) Referenced Beam Sequence 1 item(s) ----
      (300c, 0006) Referenced Beam Number    IS: '2'
    -----
    (300c, 0022) Referenced Fraction Group Number IS: '1'
  -----
(300c, 0060) Referenced Structure Set Sequence 1 item(s) ----
  (0008, 1150) Referenced SOP Class UID      UI: RT Structure Set Storage
  (0008, 1155) Referenced SOP Instance UID   UI: 2.16.840.1.114337.1.1.1682093605.0
  -----
(7fe0, 0010) Pixel Data                     OW: Array of 6139880 elements
```



# RT dose file: tags & attributes



## Selection of specific tags

PatientName: CIRS Invivo statico^1cm sotto SW  
PatientID: CIRSinvivo  
StudyDate: 20221123  
Modality: RTDOSE  
SliceThickness: None  
Image Position: [-265.3, -41.9, -675]  
Image Orientation: [1, 0, 0, 0, 1, 0]  
NumberOfFrame: 79  
Rows: 145  
Columns: 268  
PixelSpacing: [2, 2]  
BitsAllocated: 16  
BitsStored: 16  
Dose Units: GY  
Dose Type: PHYSICAL  
DoseSummationType: BEAM  
GridFrameOffsetVector: [0, 2, 4, 6, 8, 10, 12, 14, 16, 18, 20, 22, 24, 26, 28, 30, 32, 34, 36, 38, 40, 42, 44, 46, 48, 50, 52, 54, 56, 58, 60, 62, 64, 66, 68, 70, 72, 74, 76, 78, 80, 82, 84, 86, 88, 90, 92, 94, 96, 98, 100, 102, 104, 106, 108, 110, 112, 114, 116, 118, 120, 122, 124, 126, 128, 130, 132, 134, 136, 138, 140, 142, 144, 146, 148, 150, 152, 154, 156]  
DoseGridScaling: 9.53212e-006

# Meaning of the tags

**(0008, 0060) Modality:** Type of equipment that originally acquired the data used to create the images in this Series.

**(0018, 0050) Slice Thickness:** Nominal slice thickness, in mm.

**(0020, 0032) Image Position (Patient):** The x, y, and z coordinates of the upper left hand corner (center of the first voxel transmitted) of the image, in mm.

**(0020, 0037) Image Orientation (Patient):** The direction cosines of the first row and the first column with respect to the patient. See <https://dicom.innolitics.com/ciods/rt-dose/image-plane/00200037>

**(0020, 0052) Frame of Reference UID:** Uniquely identifies the Frame of Reference for a Series. See <https://dicom.innolitics.com/ciods/rt-plan/frame-of-reference/00200052>

**(0028, 0008) Number of Frames:** Number of frames in a Multi-frame Image. A Multi-frame Image is defined as a Image whose pixel data consists of a sequential set of individual Image Pixel frames. A Multi-frame Image is transmitted as a single contiguous stream of pixels. Frame headers do not exist within the data stream.

**(0028, 0010) Rows:** Number of rows in the image.

**(0028, 0011) Columns:** Number of columns in the image.

**(0028, 0030) Pixel Spacing:** Physical distance in the patient between the center of each pixel, specified by a numeric pair - adjacent row spacing (delimiter) adjacent column spacing in mm.

**(0028, 0100) Bits Allocated:** Number of bits allocated for each pixel sample.

**(0028, 0101) Bits Stored:** Number of bits stored for each pixel sample.

**(3004, 0002) Dose Units:** Dose axis units.

**(3004, 0004) Dose Type:** Type of dose. Defined Terms:

- **PHYSICAL:** physical dose
- **EFFECTIVE:** physical dose after correction for biological effect using user-defined modeling technique
- **ERROR:** difference between desired and planned dose

**(3004, 000a) Dose Summation Type:** Type of dose summation. Defined Terms:

- **PLAN:** dose calculated for entire delivery of all fraction groups of RT Plan
- **MULTI\_PLAN:** dose calculated for entire delivery of 2 or more RT Plans
- **FRACTION:** dose calculated for entire delivery of a single Fraction Group within RT Plan
- **BEAM:** dose calculated for entire delivery of one or more Beams within RT Plan
- **BRACHY:** dose calculated for entire delivery of one or more Brachy Application Setups within RT Plan
- **FRACTION\_SESSION:** dose calculated for a single session ("fraction") of a single Fraction Group within RT Plan
- **BEAM\_SESSION:** dose calculated for a single session ("fraction") of one or more Beams within RT Plan
- **BRACHY\_SESSION:** dose calculated for a single session ("fraction") of one or more Brachy Application Setups within RT Plan
- **CONTROL\_POINT:** dose calculated for one or more Control Points within a Beam for a single fraction
- **RECORD:** dose calculated for RT Beams Treatment Record

**(3004, 000c) Grid Frame Offset Vector:** An array that contains the dose image plane offsets (in mm) of the dose image frames in a multi-frame dose. Required if multi-frame pixel data are present and Frame Increment Pointer (0028,0009) points to Grid Frame Offset Vector (3004,000C).

**(3004, 000e) Dose Grid Scaling:** Scaling factor that when multiplied by the dose grid data found in Pixel Data (7FE0,0010) Attribute of the Image Pixel Module, yields grid doses in the dose units as specified by Dose Units (3004,0002).



# Grid Frame Offset Vector

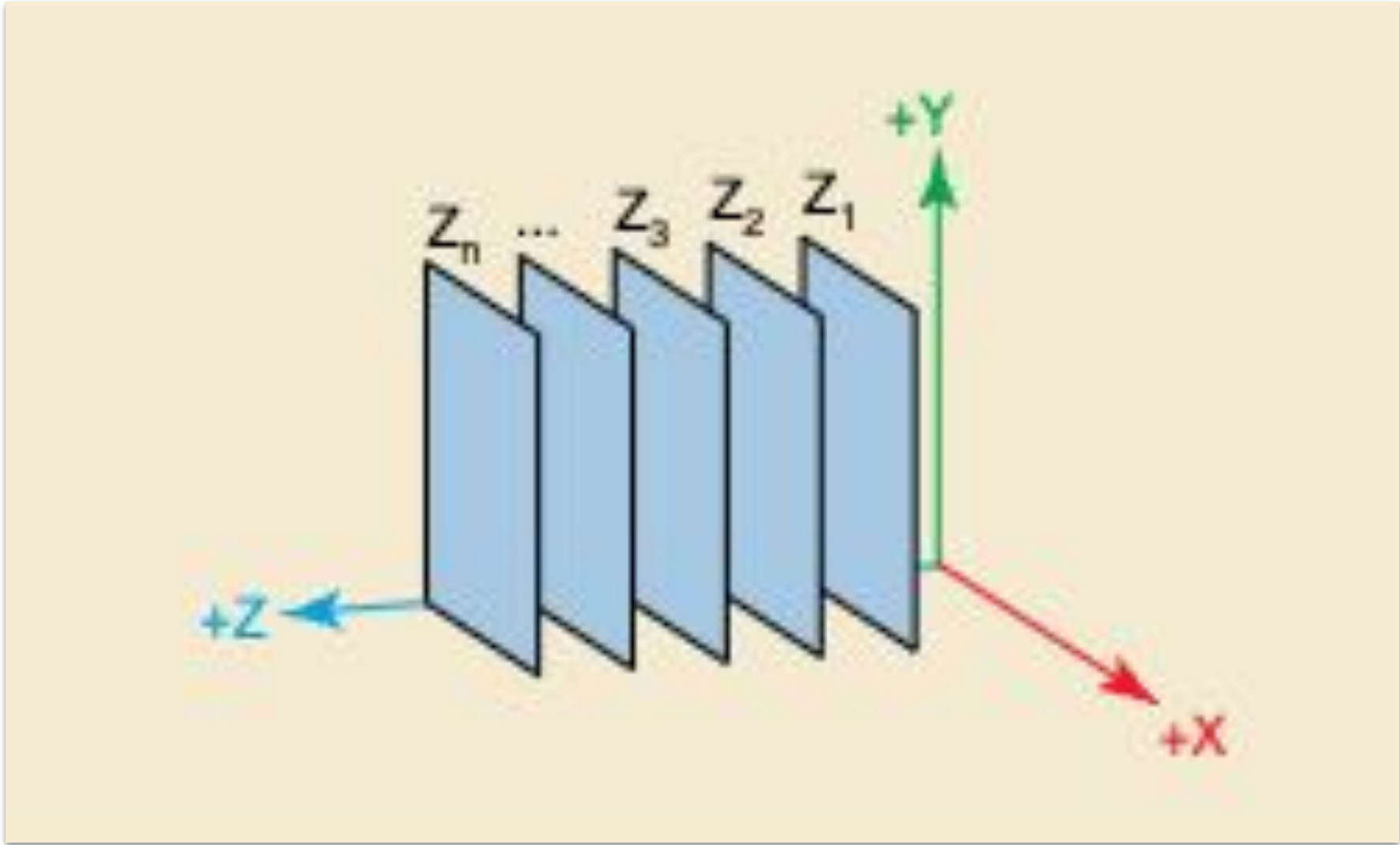
Dose Grid Frame example

## Grid Frame Offset Vector Attribute

Tag	(3004,000C)
Type	Conditionally Required (1C)
Keyword	GridFrameOffsetVector
Value Multiplicity	2-n
Value Representation	Decimal String (DS)

An array that contains the dose image plane offsets (in mm) of the dose image frames in a multi-frame dose. Required if multi-frame pixel data are present and Frame Increment Pointer (0028,0009) points to Grid Frame Offset Vector (3004,000C). See Section C.8.8.3.2.

<https://dicom.innolitics.com/ciods/rt-dose/rt-dose/3004000c>



Dose Units: GY  
Dose Type: PHYSICAL  
DoseSummationType: BEAM  
GridFrameOffsetVector: [0, 2, 4, 6, 8, 10, 12, 14, 16, 18, 20, 22, 24, 26, 28, 30, 32, 34, 36, 38, 40, 42, 44, 46, 48, 50, 52, 54, 56, 58, 60, 62, 64, 66, 68, 70, 72, 74, 76, 78, 80, 82, 84, 86, 88, 90, 92, 94, 96, 98, 100, 102, 104, 106, 108, 110, 112, 114, 116, 118, 120, 122, 124, 126, 128, 130, 132, 134, 136, 138, 140, 142, 144, 146, 148, 150, 152, 154, 156]  
DoseGridScaling: 9.53212e-006



# Grid Frame Offset Vector

## Section C.8.8.3.2

### C.8.8.3.2 Grid Frame Offset Vector

Grid Frame Offset Vector (3004,000C) shall be provided if a dose distribution is encoded as a multi-frame image. Values of the Grid Frame Offset Vector (3004,000C) shall vary monotonically and are to be interpreted as follows:

- a. If Grid Frame Offset Vector (3004,000C) is present and its first element is zero, this Attribute contains an array of n elements indicating the plane location of the data in the right-handed image coordinate system, relative to the position of the first dose plane transmitted, i.e., the point at which Image Position (patient) (0020,0032) is defined, with positive offsets in the direction of the cross product of the row and column directions.
- b. If Grid Frame Offset Vector (3004,000C) is present, its first element is equal to the third element of Image Position (patient) (0020,0032), and Image Orientation (patient) (0020,0037) has the value (1,0,0,0,1,0), then Grid Frame Offset Vector contains an array of n elements indicating the plane location (patient z coordinate) of the data in the Patient-Based Coordinate System.

In future implementations, use of option a) is strongly recommended.

This Attribute is conditional since the [RT Dose Module](#) may be included even if pixel doses are not being transmitted, or the image may be a single-frame image. If the [Multi-frame Module](#) is present, Frame Increment Pointer (0028,0009) shall have the Enumerated Value of 3004000C (Grid Frame Offset Vector).

#### Note

Option (a) can represent a rectangular-parallelepiped dose grid with any orientation with respect to the patient, while option (b) can only represent a rectangular-parallelepiped dose grid whose planes are in the transverse patient dimension and whose x- and y-axes are parallel to the patient x- and y-axes.

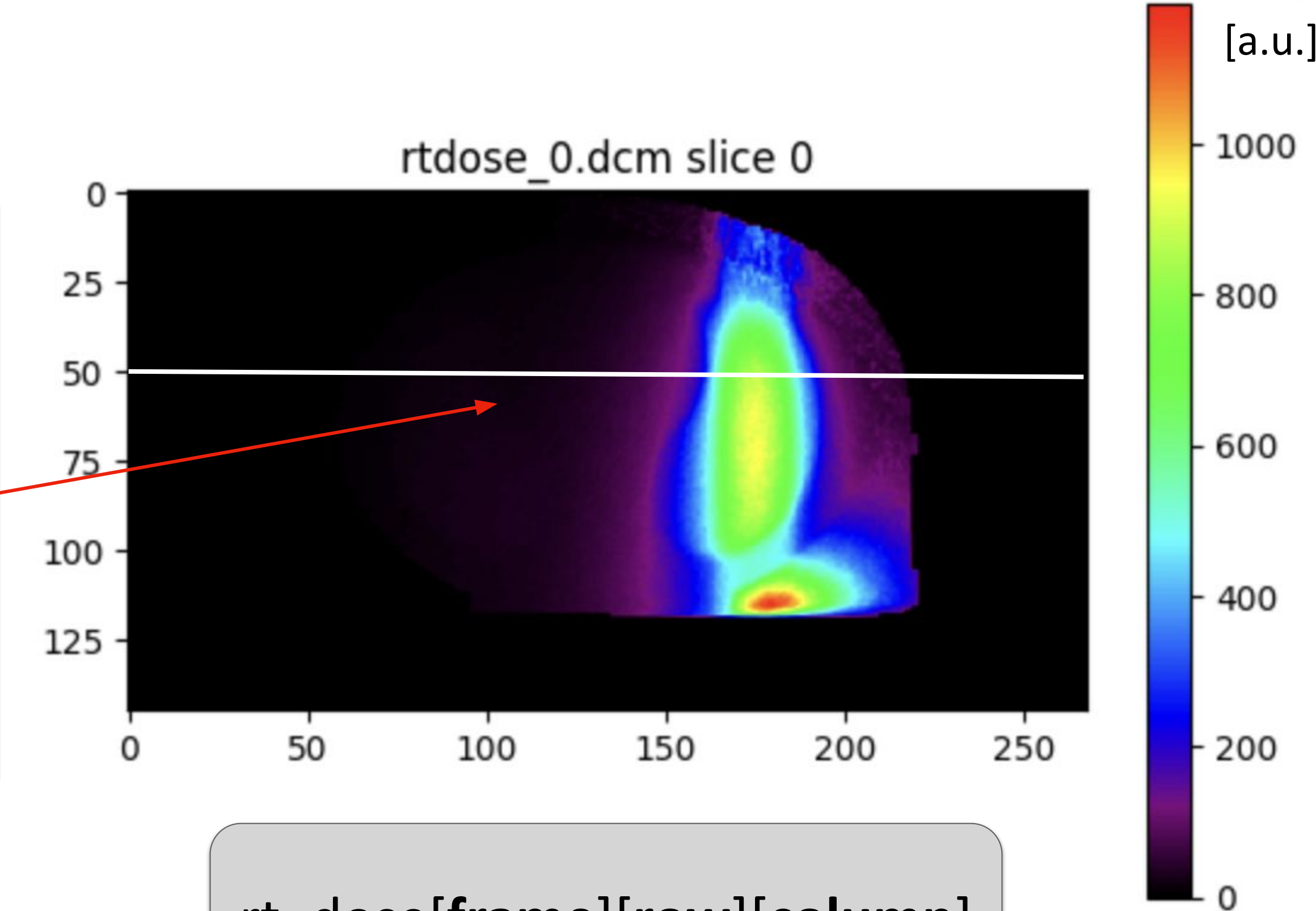
GridFrameOffsetVector: [0, 2, 4, 6, 8, 10, 12, 14, 16, 18, 20, 22, 24, 26, 28, 30, 32, 34, 36, 38, 40, 42, 44, 46, 48, 50, 52, 54, 56, 58, 60, 62, 64, 66, 68, 70, 72, 74, 76, 78, 80, 82, 84, 86, 88, 90, 92, 94, 96, 98, 100, 102, 104, 106, 108, 110, 112, 114, 116, 118, 120, 122, 124, 126, 128, 130, 132, 134, 136, 138, 140, 142, 144, 146, 148, 150, 152, 154, 156]



# RT dose

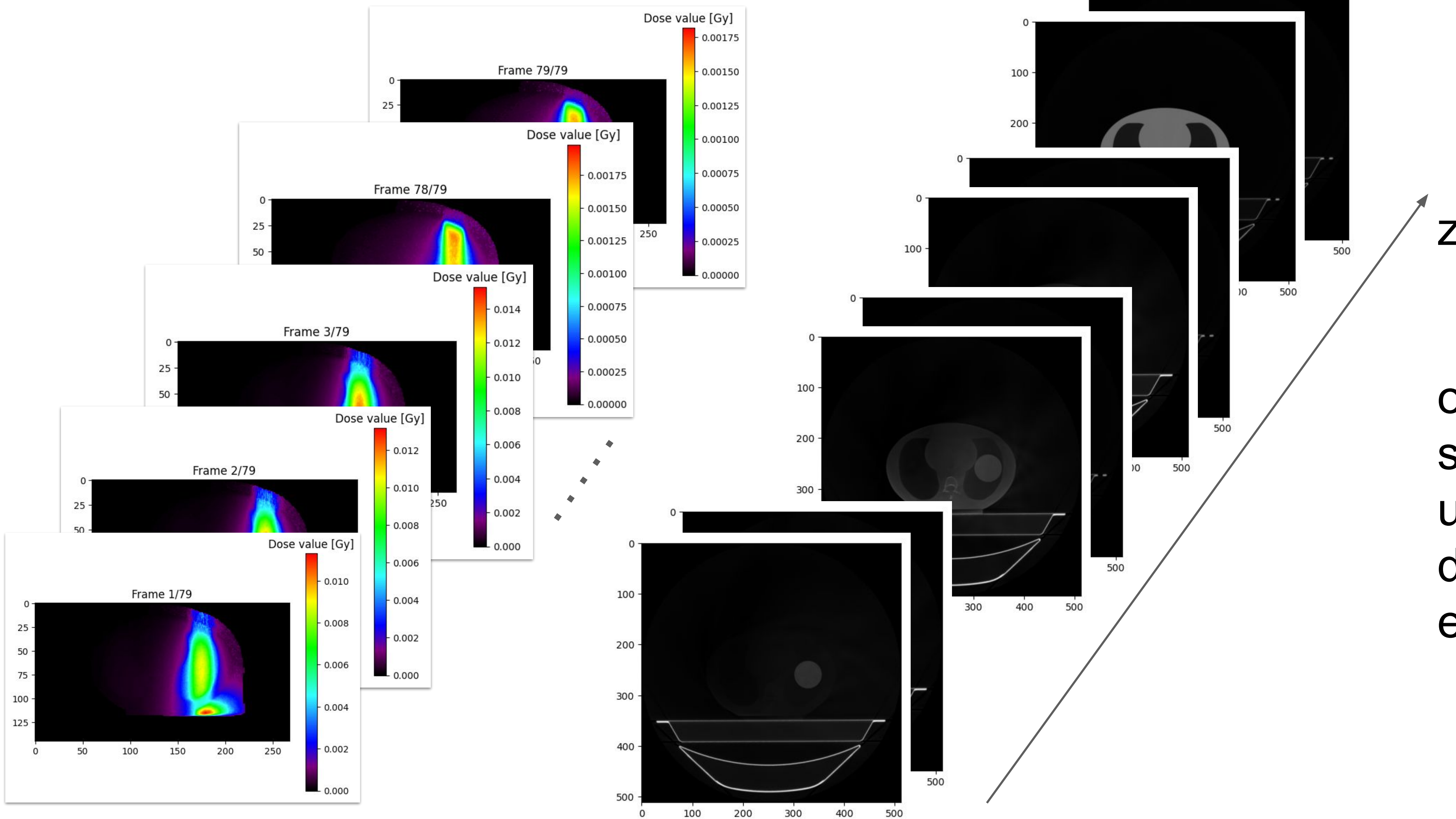
rt\_dose[0][50][:]

[	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	0	0	0	0	0	3	4	4	4	6	6	6	6	6	6	6	7	7
	7	9	9	10	10	13	13	13	13	15	15	15	15	15	17	17	17	17
	17	17	18	18	18	18	18	19	19	20	18	18	18	17	15	15	18	17
	15	17	19	18	19	19	20	20	21	21	21	23	24	24	25	26	26	30
	32	30	35	35	34	36	38	40	40	44	49	49	53	54	57	60	66	66
	72	74	81	85	93	95	99	108	108	112	122	129	143	160	177	189	219	258
	336	439	507	581	653	691	728	757	771	821	834	830	830	844	850	845	797	805
	774	744	720	663	627	553	487	467	402	346	320	267	217	173	151	137	126	108
	104	98	88	85	89	75	63	57	55	69	78	53	100	60	49	57	49	61
	57	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0



rt\_dose[frame][row][column]

# RT dose



ogni due  
slice CT ho  
una  
distribuzione  
e di dose



# RT dose & CT

PatientName: CIRS Invivo statico^1cm sotto SW

PatientID: CIRSinvivo

StudyDate: 20221123

Modality: RTDOSE

SliceThickness: None

Image Position: [-265.3, -41.9, -675]

Image Orientation: [1, 0, 0, 0, 1, 0]

NumberOfFrame: 79

Rows: 145

Columns: 268

PixelSpacing: [2, 2]

BitsAllocated: 16

BitsStored: 16

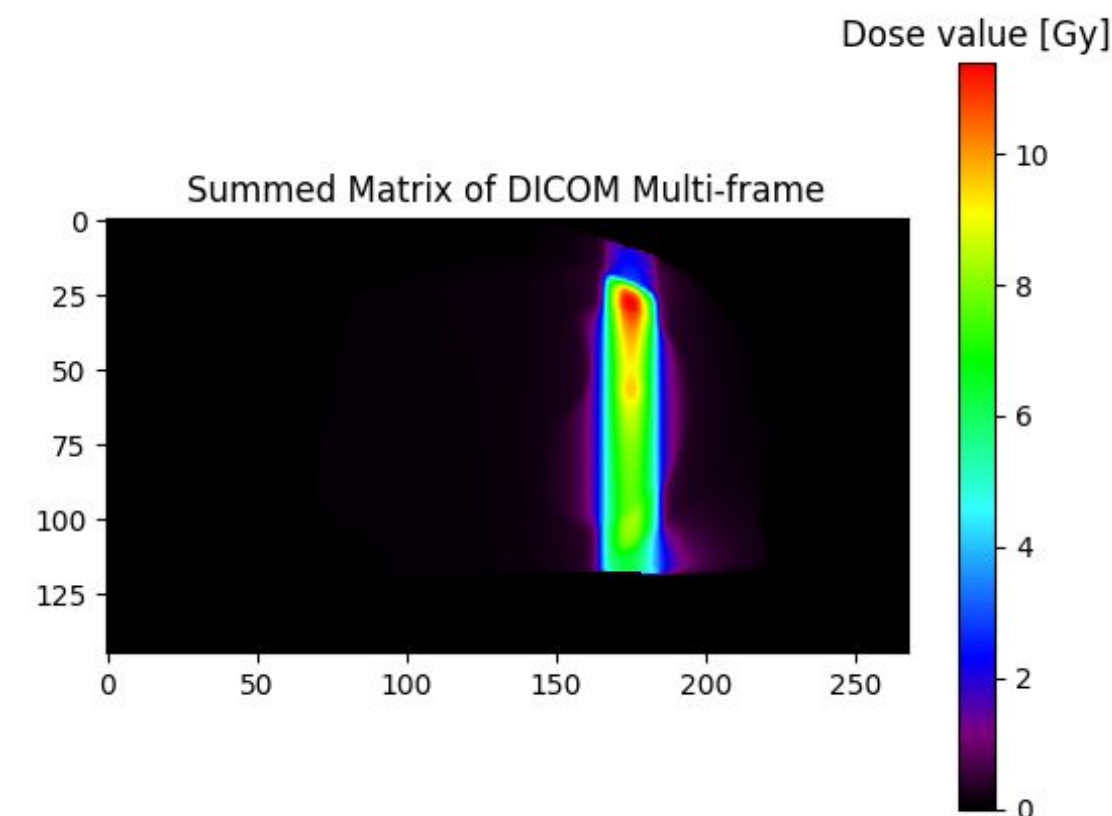
Dose Units: GY

Dose Type: PHYSICAL

DoseSummationType: BEAM

GridFrameOffsetVector: [0, 2, 4, 6, 8, 10, 12, 14, 16, 18, 20, 22, 48, 50, 52, 54, 56, 58, 60, 62, 64, 66, 68, 70, 72, 74, 76, 78, 80, 04, 106, 108, 110, 112, 114, 116, 118, 120, 122, 124, 126, 128, 130, 152, 154, 156]

DoseGridScaling: 9.53212e-006



Hanno stessa z → Provare a  
sovrapporre  
le due immagini

SLICE 1/158

Image Type: ['ORIGINAL', 'PRIMARY', 'AXIAL', 'HELIX']

Acquisition Time: 131232

Modality: CT

Slice Thickness: 2.00

Spacing Between Slices: 1.0

Table Height: 173.000000

Image Position (Patient): [-300, -218, -675]

Image Orientation (Patient): [1, 0, 0, 0, 1, 0]

Slice Location: -675.00

Rows: 512

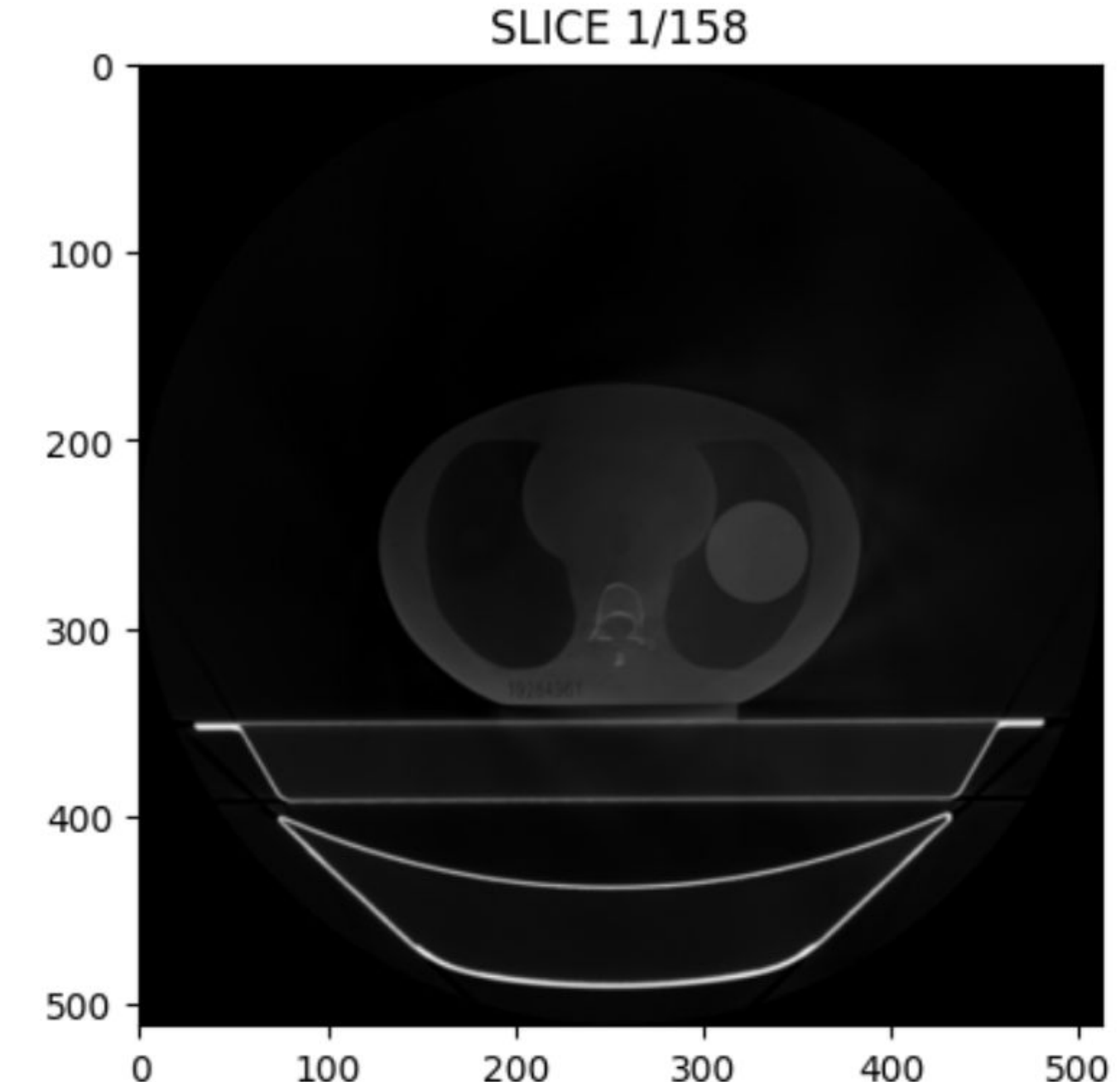
Columns: 512

Pixel Spacing: [1.171875, 1.171875]

Bits Allocated: 16

Window Center: [00050, 00050]

Window Width: [00350, 00350]



# How to overlay the RTDOSE and Image data in the correct position?

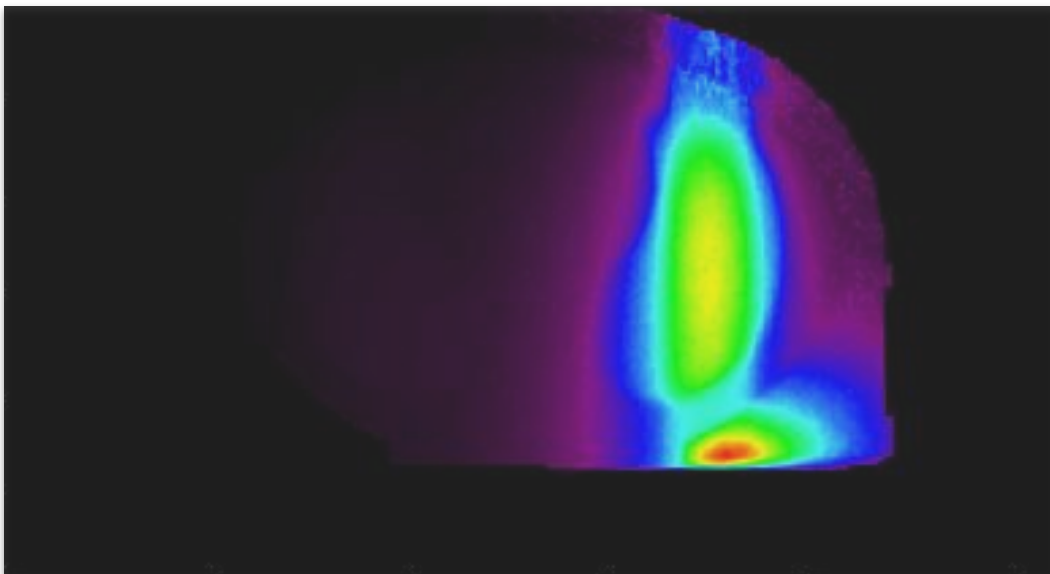
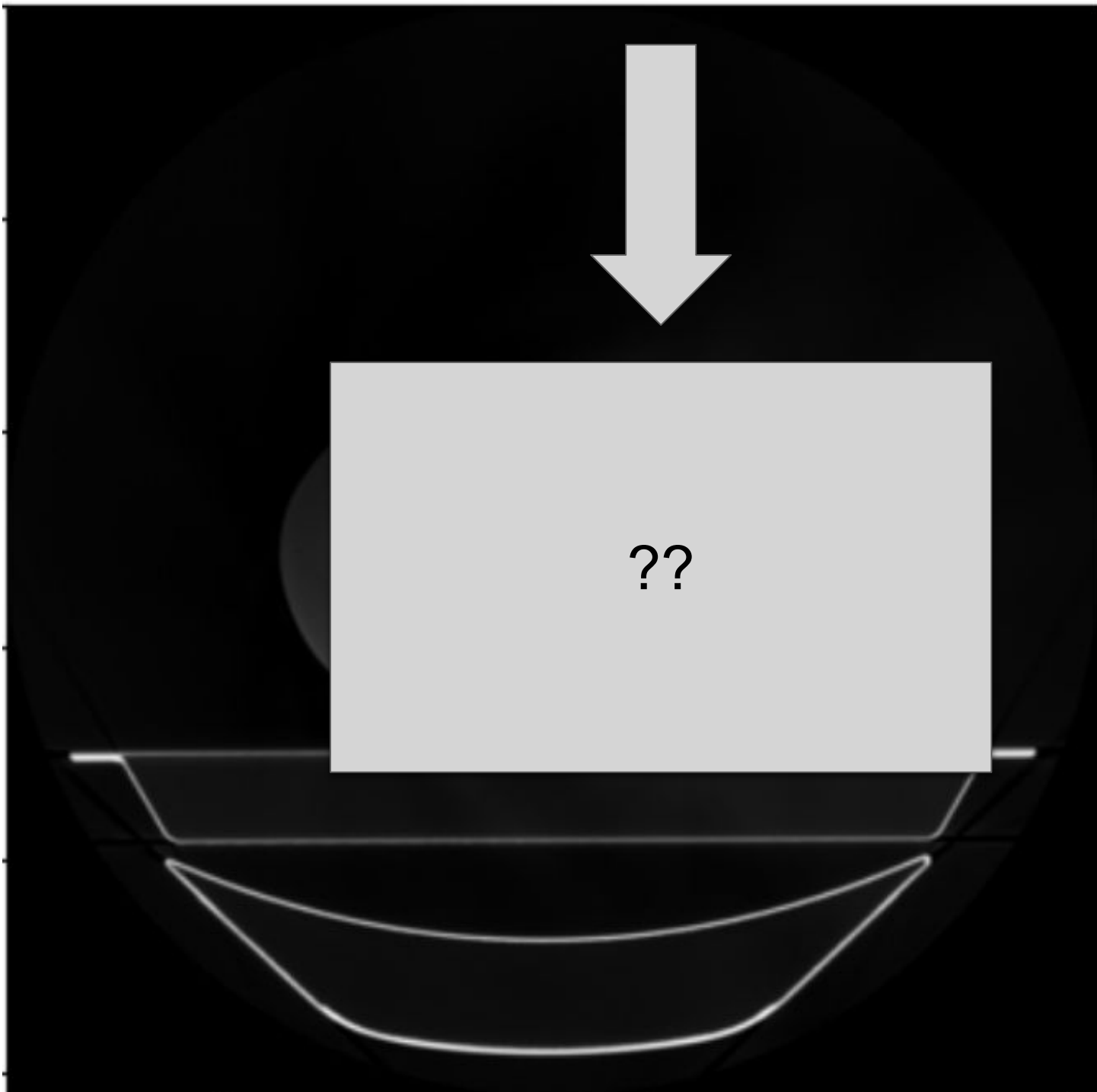
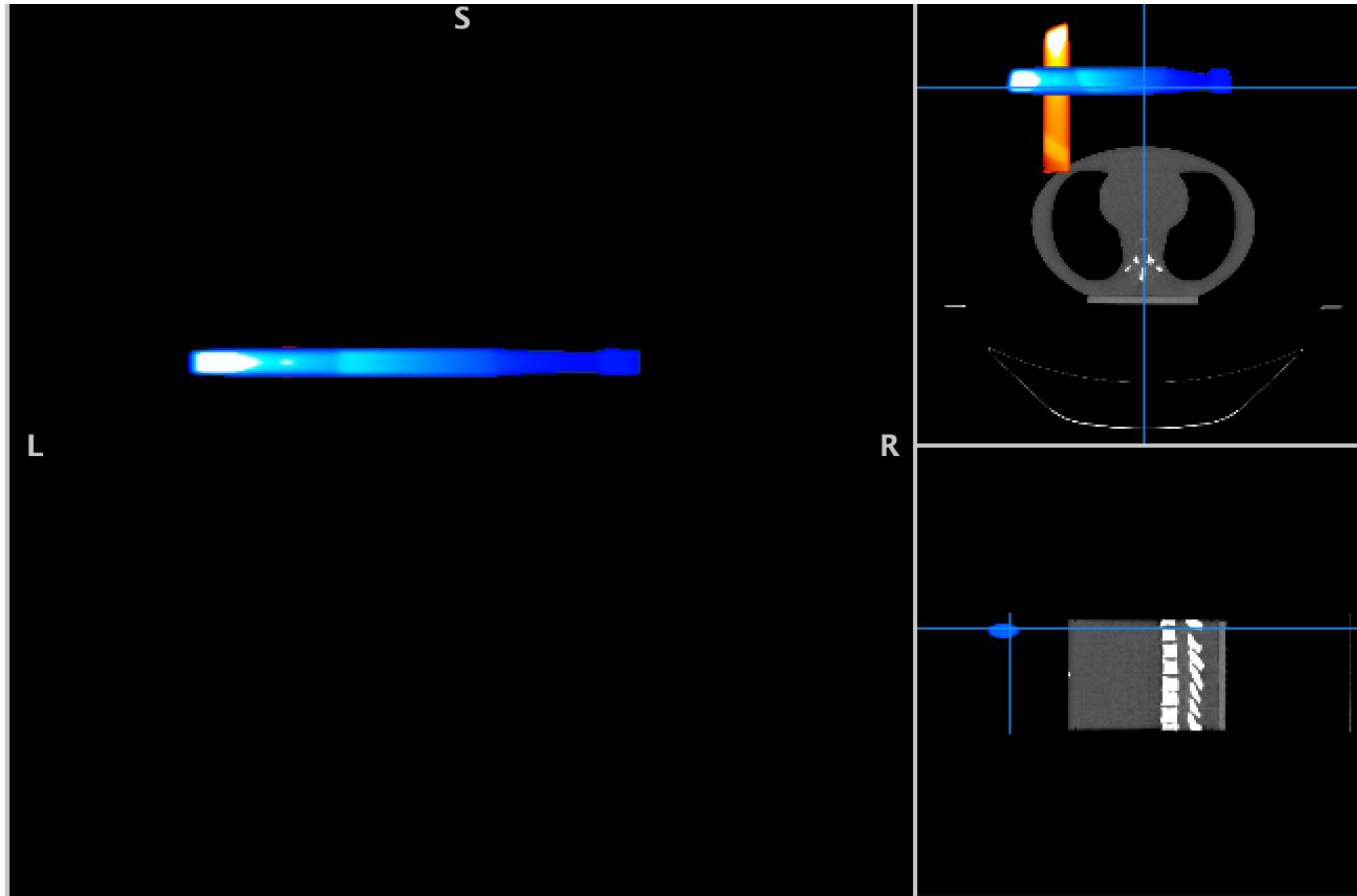


	Image Position (Patient) [mm]	Pixel Spacing [mm]
CT	[-300, -218, -675]	[1.171875, 1.171875]
RTDOSE	[-265.3, -41.9, -675]	[2, 2]







OverviewForumMangoiMangoPapayaMulti-image Analysis GUI

FeaturesDownloadsVideo TutorialsUsing MangoPublicationsOnline Web Viewer

Mango – short for Multi-image Analysis GUI – is a viewer for medical research images. It provides analysis tools and a user interface to navigate image volumes.

**Versions**

There are three versions of Mango, each geared for a different platform:

- Mango – **Desktop** – Mac OS X, Windows, and Linux
- Papaya – **Browser** – Firefox, Chrome, Safari and IE
- iMango – **Mobile** – Apple iPad

**Features**

- Support for *Analyze*, *DICOM*, *NEMA-DES*, *MINC*, *NIFTI* and *NIFTI2* image formats
- Support for *VTK (legacy)*, *GIFTI (.surf.gii)* and *BrainVisa* surface formats
- Partial support for TIFF, Concorde microPET, *AFNI* (legacy), *Stimulate*, and CTI ECAT
- Development:** Supports both *Java Plugin API* and *Python Script API* development
- Customizable: Create custom filters, color tables, file formats, and atlases
- Command-line integration:** open and process images from the command-line
- Web: *Custom protocol* and *Papaya JavaScript viewer*
- ROI Editing:** Threshold and component-based tools for painting and tracing ROIs
- Surface Rendering:** Interactive surface models supporting cut planes and overlays
- Image Registration:** Semi-automatic image coregistration and manual transform editing
- Image Stacking:** Threshold and transparency-based image overlay stacking
- Analysis:** Histogram, cross-section, time-series analysis, image and ROI statistics
- Processing:** Kernel and rank filtering, arithmetic/logic image and ROI calculators




**Credits**

Designed and developed by:

- Mohamad Habes, Ph.D.
- Jack L. Lancaster, Ph.D.
- Michael J. Martinez.

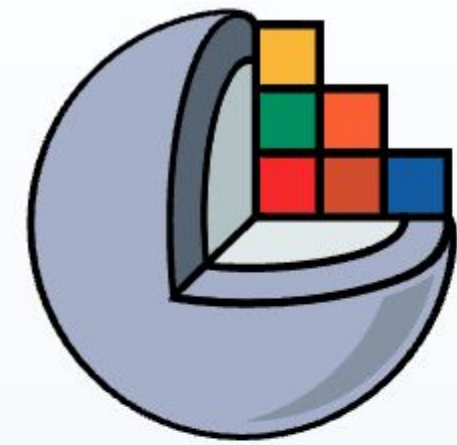
**Publications**

- Lancaster JL, Cykowski MD, McKay DR, Kochunov PV, Fox PT, Rogers W, Toga AW, Zilles K, Amunts K, Mazziotta J (2010). Anatomical Global Spatial Normalization. *Neuroinformatics*, 8:171–182. [\[PDF\]](#)
- Lancaster JL, Laird AR, Eickhoff SB, Martinez MJ, Fox PM, Fox, PT (2012). Automated regional behavioral analysis for human brain images. *Frontiers in Neuroinformatics*, 6, 23. [\[PDF\]](#)
- Lancaster JL, McKay DR, Cykowski MD, Martinez MJ, Tan X, Valaparla S, Zhang Y, Fox PT (2011). Automated analysis of fundamental features of brain structures. *Neuroinformatics*, 9(4):371-80. [\[PDF\]](#)

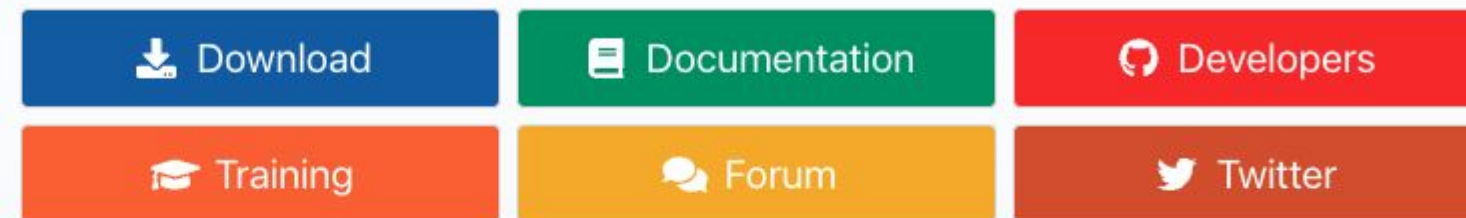


<https://mangoviewer.com/>





## 3D Slicer image computing platform



3D Slicer is a **free, open source** software for visualization, processing, segmentation, registration, and analysis of medical, biomedical, and other 3D images and meshes; and planning and navigating image-guided procedures.

### What is 3D Slicer ?

**Desktop software** to solve advanced image computing challenges with a focus on clinical and biomedical applications.

**Development platform** to quickly build and deploy custom solutions for research and commercial products, using free, open source software.

**Community** of knowledgeable users and developers working together to improve medical computing.

<https://www.slicer.org/>

<https://slicer.readthedocs.io/en/latest/>

<https://pypi.org/project/pynrrd/>

<https://www.slicer.org/w/index.php/Documentation/Nightly/Extensions/SlicerRT#Tutorials>

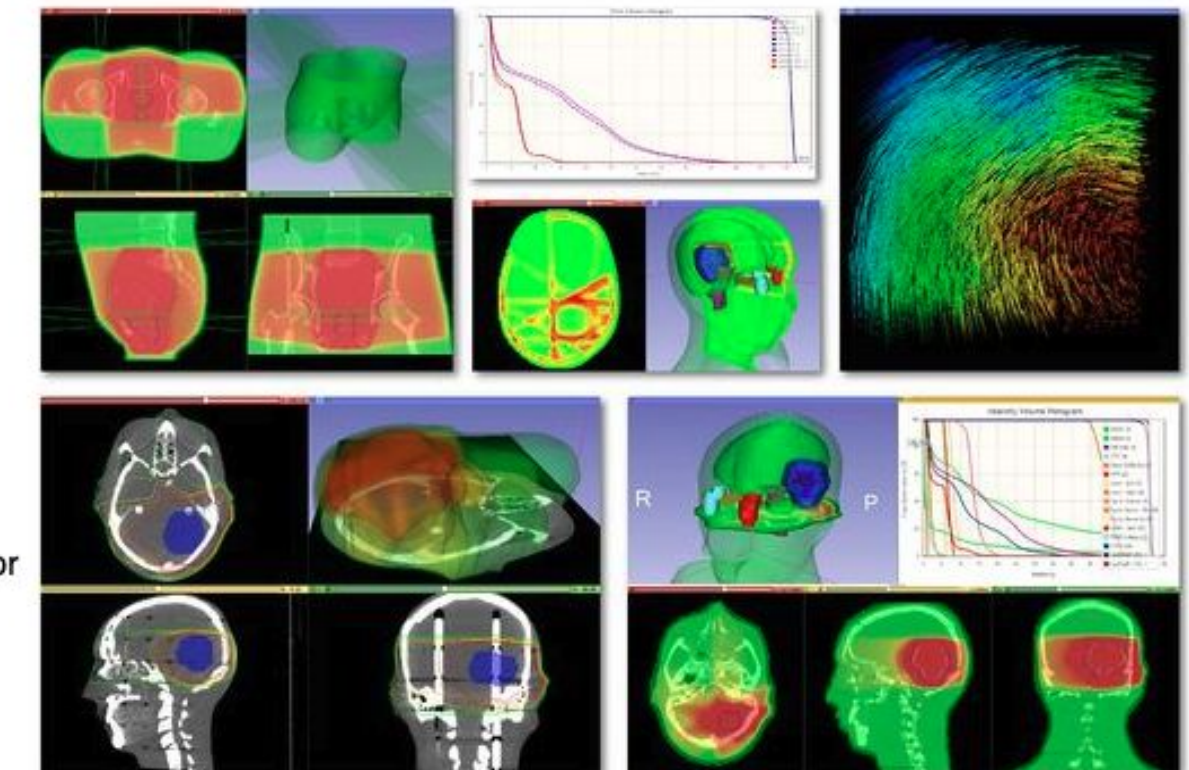
<https://discourse.slicer.org/t/how-does-3d-slicer-overlay-the-rtdose-and-image-data-in-the-correct-position/8924>



- SlicerRT is a radiation therapy toolkit for 3D Slicer, containing generic RT features for import/export, analysis, visualization, aiming to make 3D Slicer a powerful radiotherapy research platform. SlicerRT development is currently funded by CANARIE.  
SlicerRT was originally created via funding by Cancer Care Ontario and the Ontario Consortium for Adaptive Interventions in Radiation Oncology (OCAIRO) to provide free, open-source toolset for radiotherapy and related image-guided interventions.
- The SlicerRT extension incorporates [Plastimatch](#) modules and algorithms.
- Additional information for users can be found on the [User's Guide](#) page

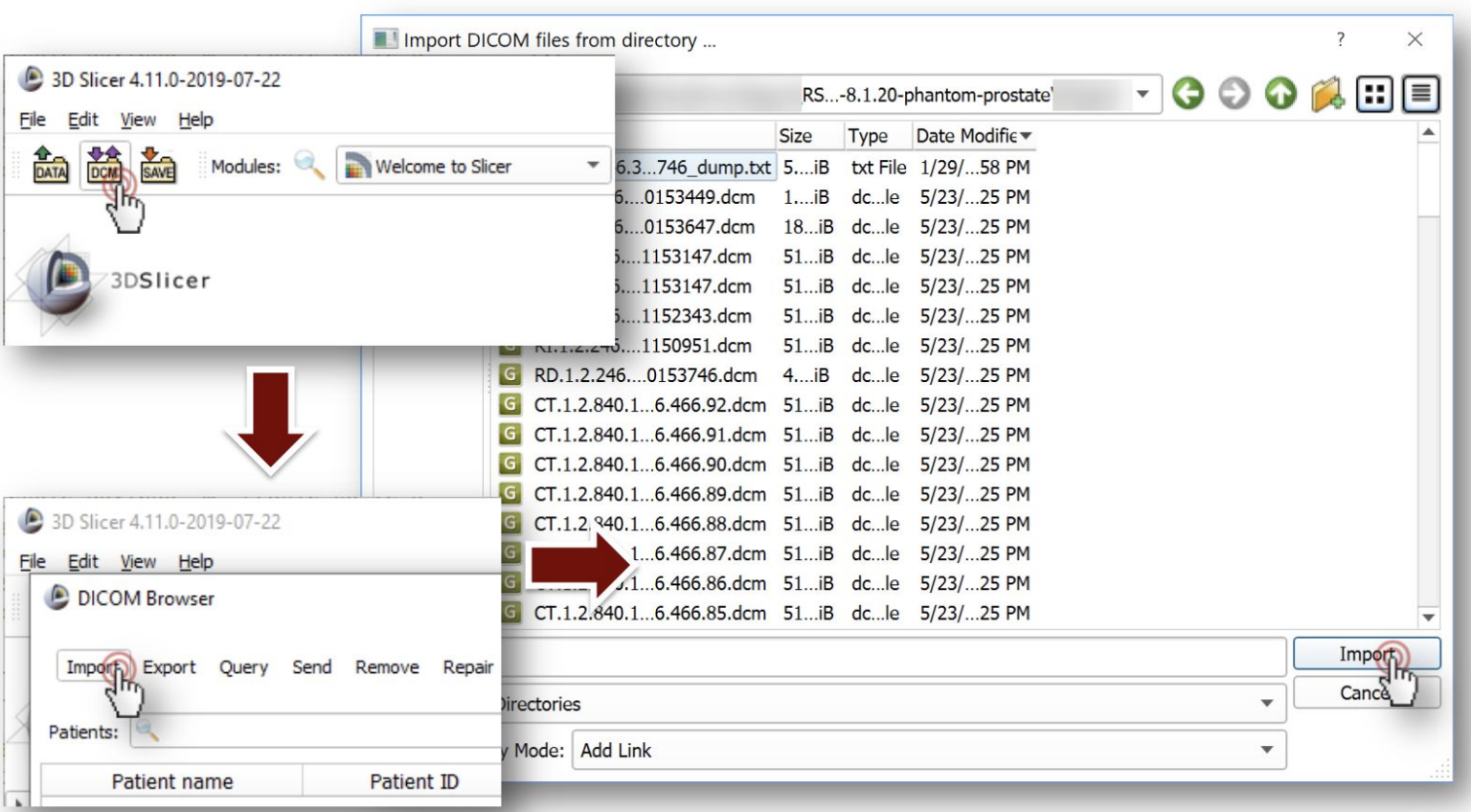
### Modules

- [External Beam Planning](#) (Treatment planning)
- Dose analysis
  - [Dose volume histogram](#)
  - [Dose accumulation](#)
  - [Dose comparison](#) (Gamma dose similarity index)
  - [Isodose line and surface display](#)
- Contour analysis
  - [Segment comparison](#) (Dice Similarity Coefficient, Hausdorff distances)
  - [Segment morphology](#) (Add/remove margin, Unify, Intersect, etc.)
- I/O
  - [DICOM-RT import, export](#) (handles datasets of types RT Structure Set, RT Dose, RT Plan, RT Image)
  - [DICOM-SRO import/export](#) (handles DICOM Spatial Registration object, both rigid and deformable)
- [Batch processing scripts](#) (currently only one is available for command-line conversion of RTSS to volume nodes)
- Modules from [Plastimatch](#)
  - [Plastimatch Automatic deformable image registration](#)
  - [Plastimatch LANDWARP Landmark](#)

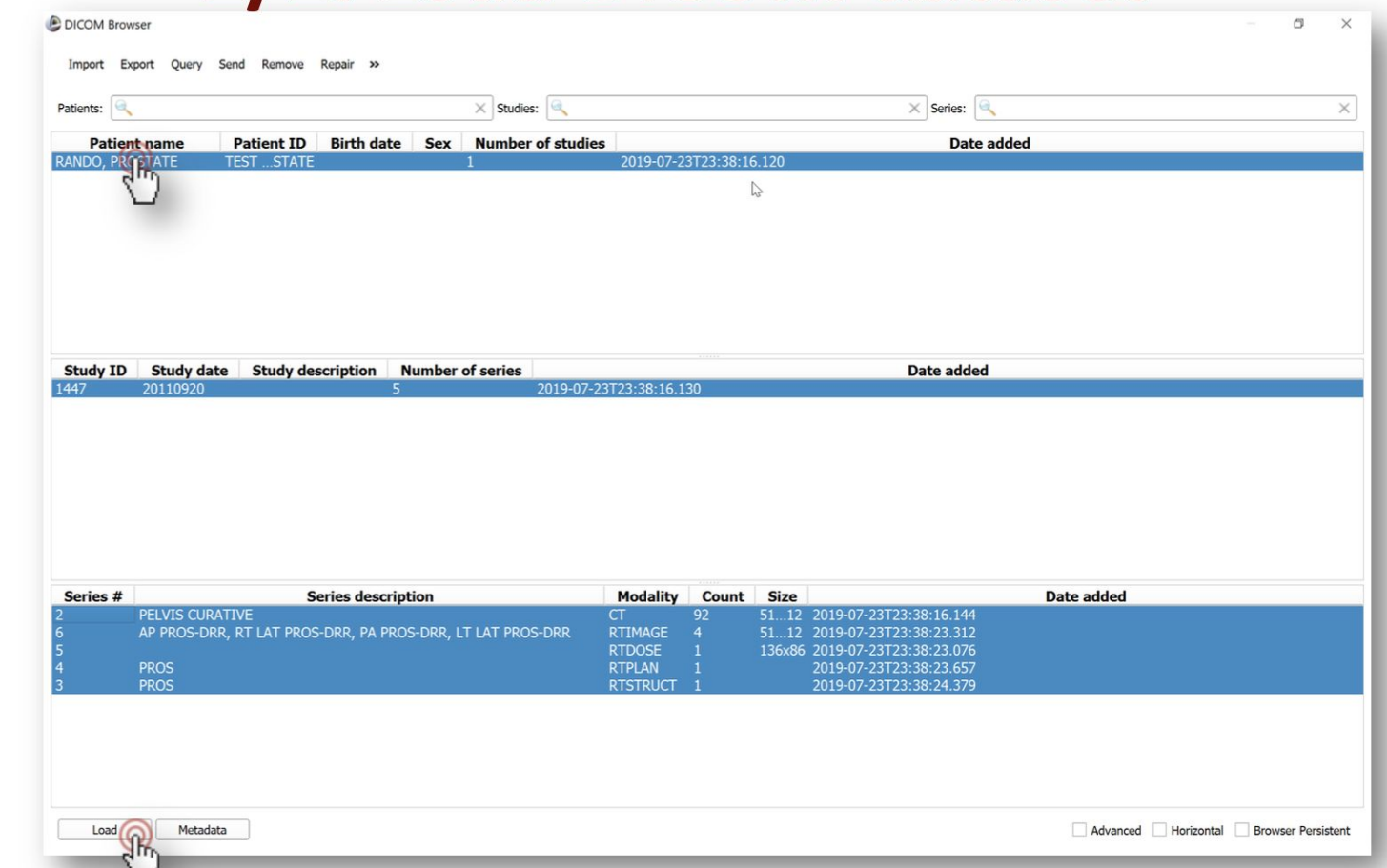




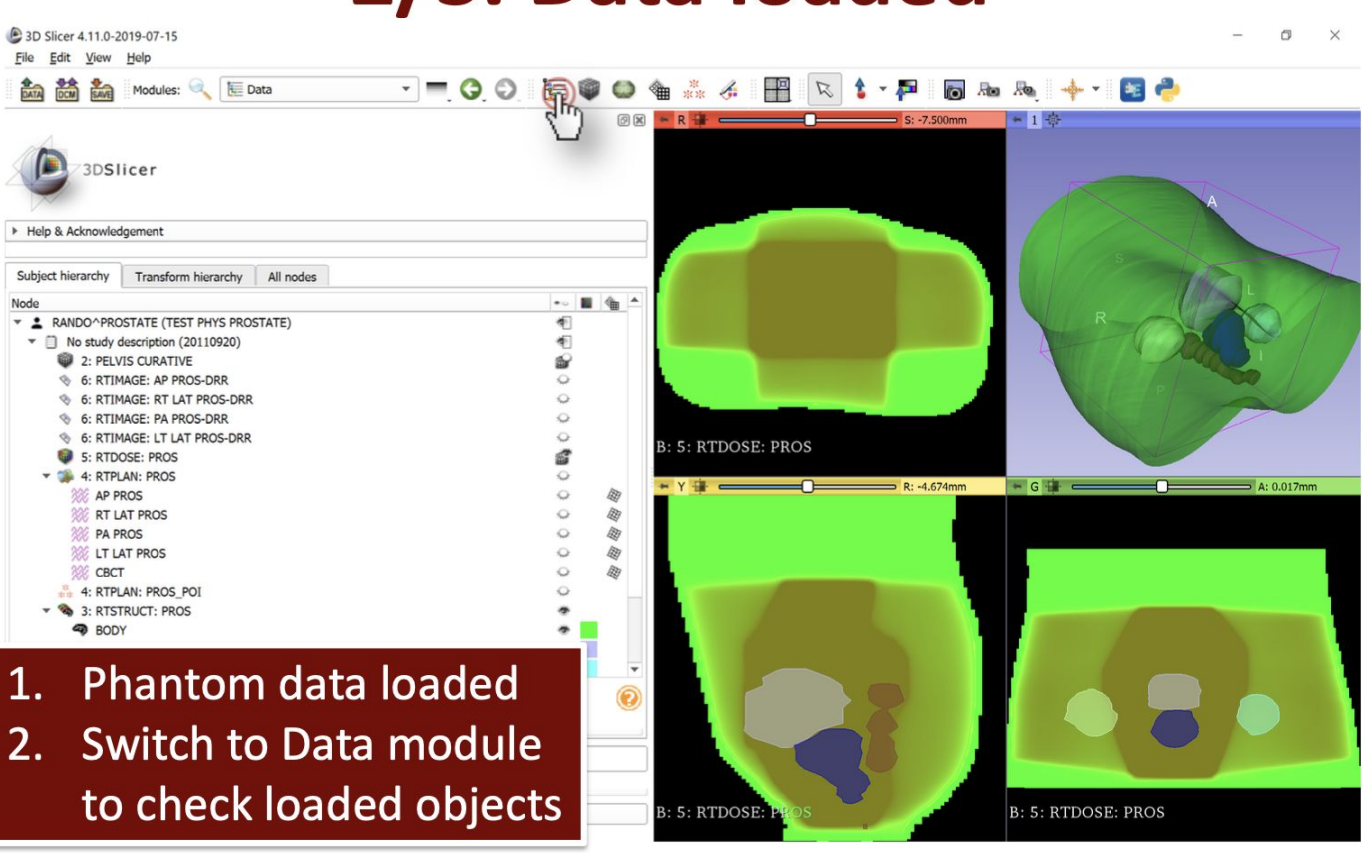
# 2/1. Import DICOM dataset



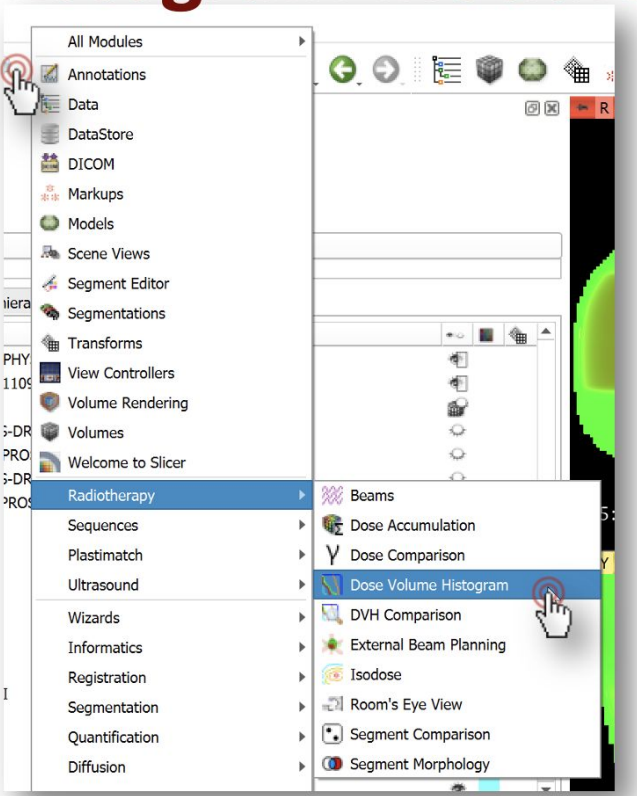
# 2/2. Load DICOM dataset



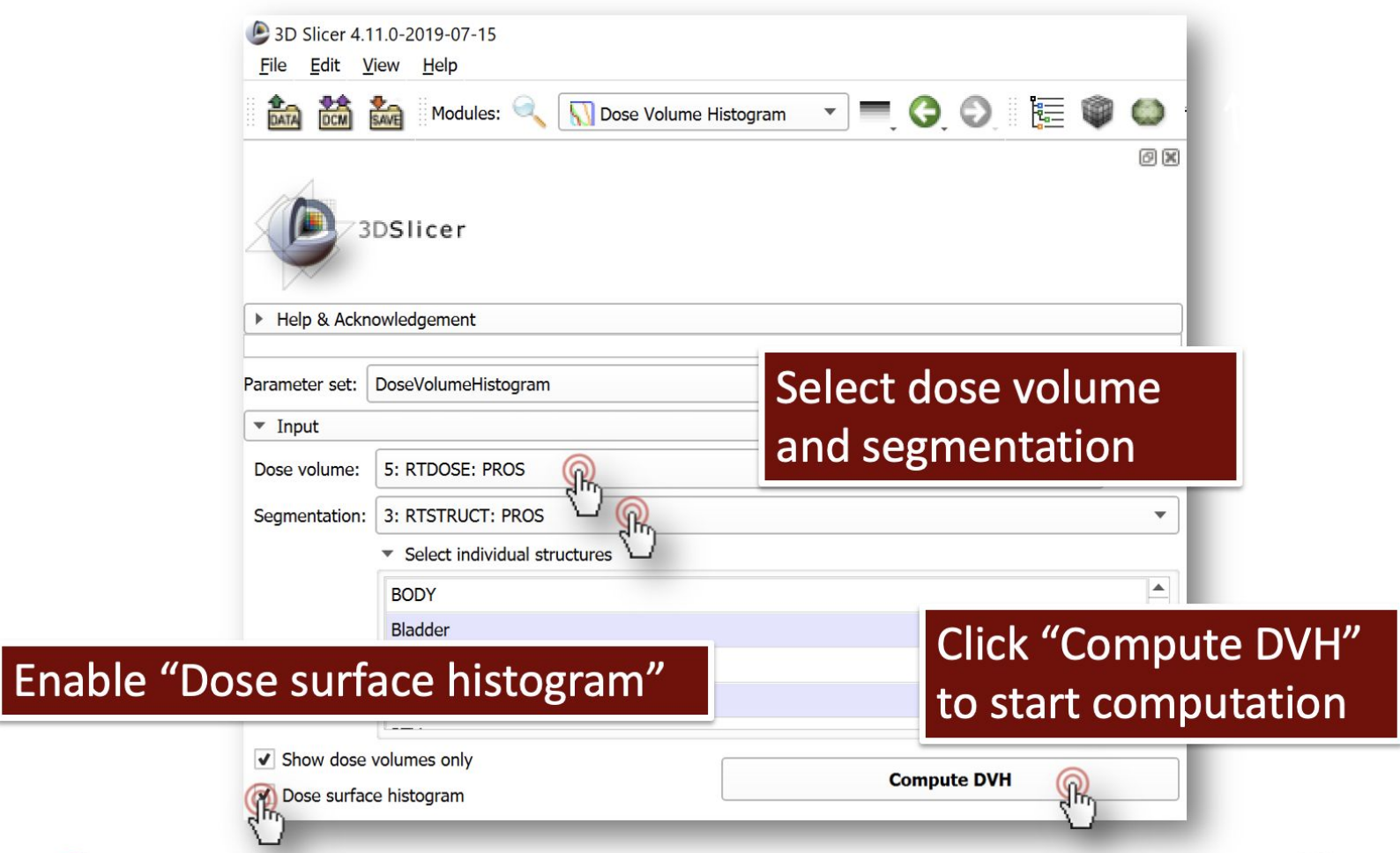
# 2/3. Data loaded



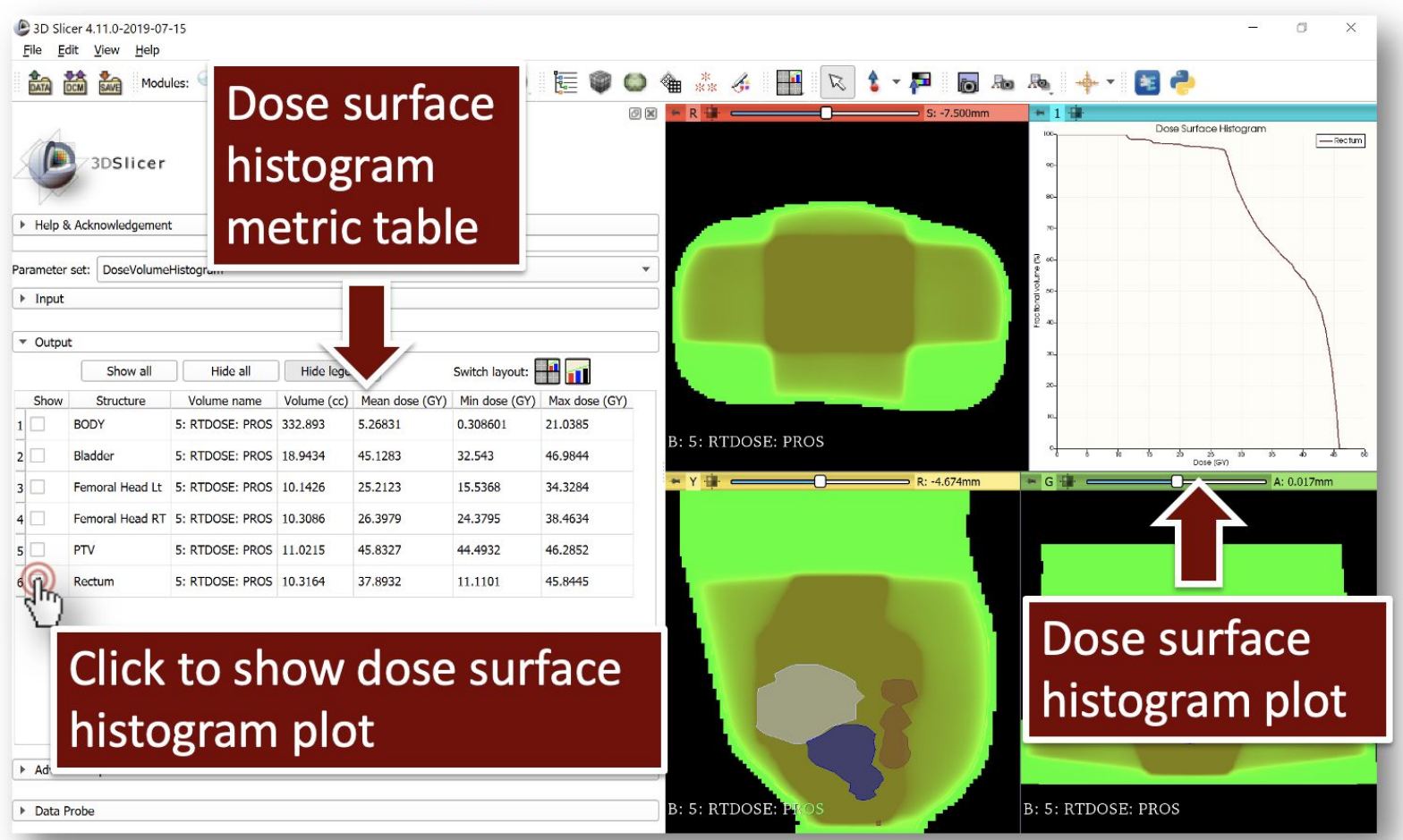
# 3/1. Switch to Dose Volume Histogram module



# 3/2. Set inputs and compute

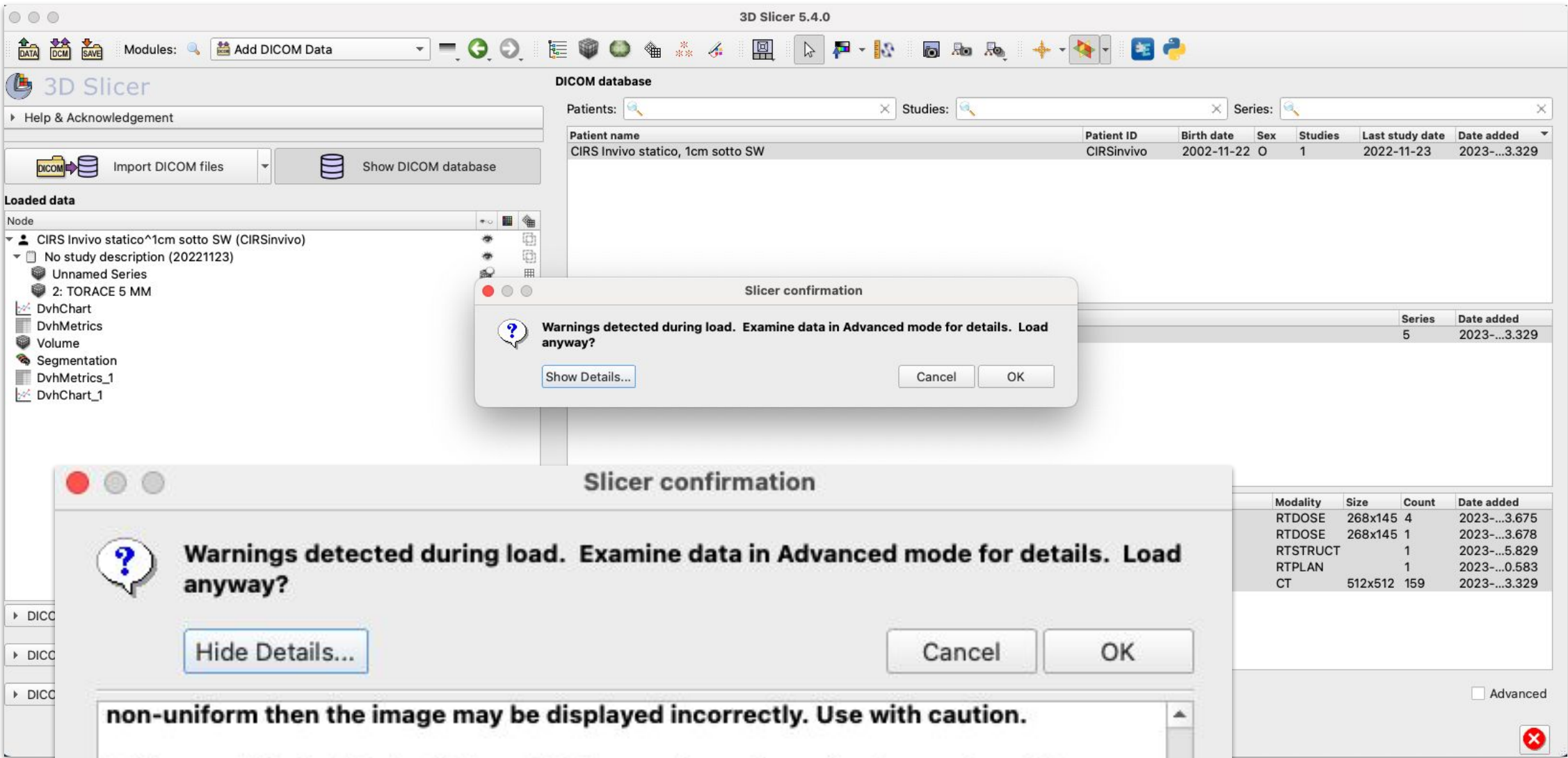
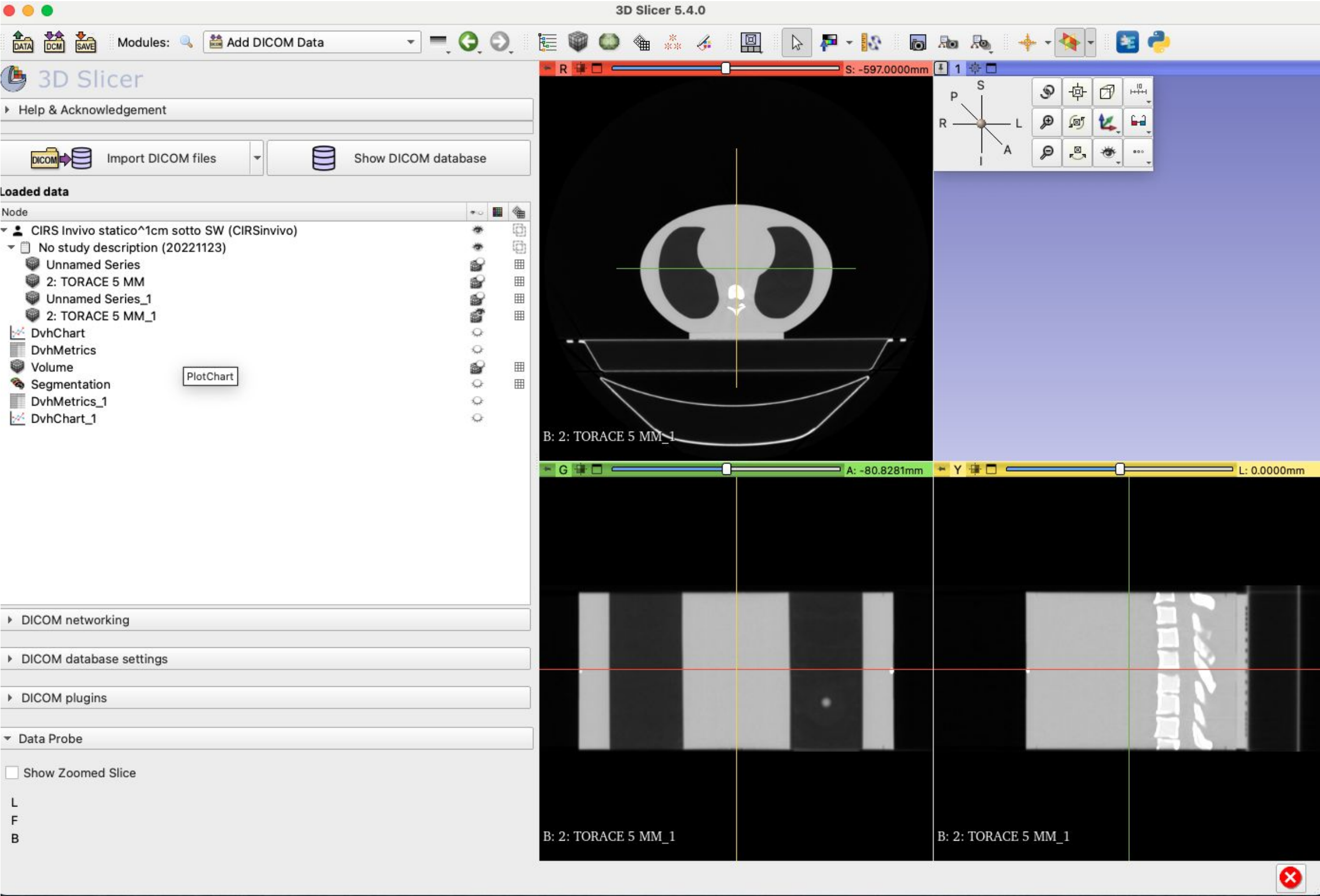


# 3/3. Output dose is visualized





# How to overlay the RTDOSE and Image data in the correct position?





# How to overlay the RTDOSE and Image data in the correct position?



When I click on “Load selection to Slicer” I get an error message “Could not load ... as a scalar volume”

A common cause of loading failure is corruption of the DICOM files by incorrect anonymization. Patient name, patient ID, and series instance UID fields should not be empty or missing (the anonymizer should replace them by other valid strings). Try to load the original, non-anonymized sequence and/or change your anonymization procedure.

If none of the above helps then check the Slicer error logs and report the error on the [Slicer forum](#). If you share the data (e.g., upload it to Dropbox and add the link to the error report) then Slicer developers can reproduce and fix the problem faster.

[https://slicer.readthedocs.io/en/latest/user\\_guide/modules/dicom.html#when-i-click-on-load-selection-to-slicer-i-get-an-error-message-could-not-load-as-a-scalar-volume](https://slicer.readthedocs.io/en/latest/user_guide/modules/dicom.html#when-i-click-on-load-selection-to-slicer-i-get-an-error-message-could-not-load-as-a-scalar-volume)

Series #	Series description	Modality	Size	Count	Date added
0		RTDOSE	268x145	4	2023-...3.675
0		RTDOSE	268x145	1	2023-...3.678
1	STRCTRNAME	RTSTRUCT		1	2023-...5.829
1	4beams	RTPLAN		1	2023-...0.583
2	TORACE 5 MM	CT	512x512	159	2023-...3.329



# RT plan



- CT
- EPID
- RT
- RTDOSE
- RTPLAN**

## RT Plan

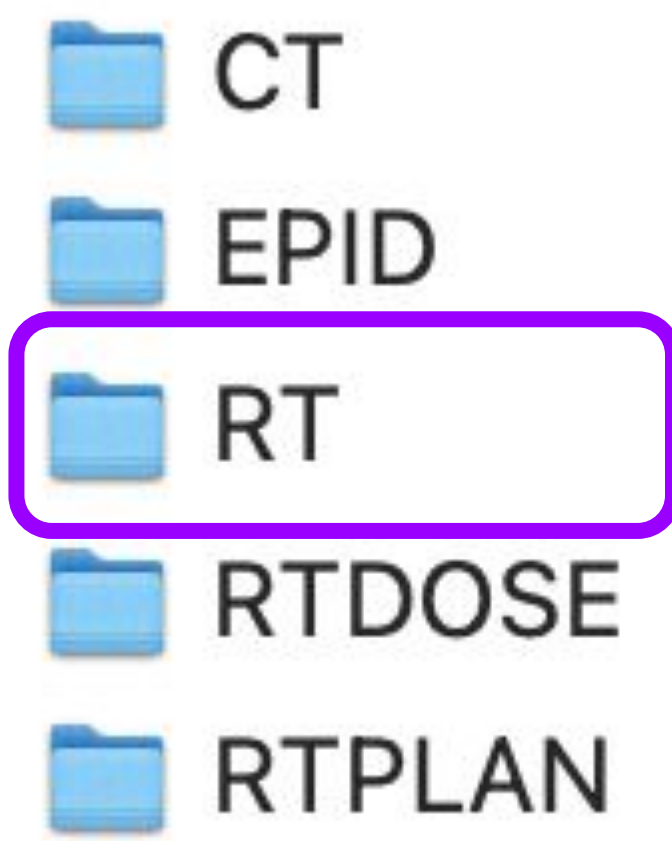
The focus for this Radiotherapy Plan IOD (RT Plan IOD) is to address the **requirements for transfer of treatment plans generated by** manual entry, a virtual simulation system, or a **treatment planning system** before or during a course of treatment. Such plans may **contain fractionation information**, and **define external beams** and/or brachytherapy application setups.

```
Out[63]: Dataset.file_meta -----
(0002, 0000) File Meta Information Group Length  UL: 168
(0002, 0001) File Meta Information Version      OB: b'\x00\x01'
(0002, 0002) Media Storage SOP Class UID       UI: RT Plan Storage
(0002, 0003) Media Storage SOP Instance UID    UI: 2.16.840.1.114337.1.1.1682093613.0
(0002, 0010) Transfer Syntax UID               UI: Implicit VR Little Endian
(0002, 0012) Implementation Class UID         UI: 2.16.840.1.114362.1
(0002, 0013) Implementation Version Name      SH: 'MIM723M31402'

(0008, 0005) Specific Character Set            CS: 'ISO_IR 100'
(0008, 0012) Instance Creation Date            DA: '20230421'
(0008, 0013) Instance Creation Time            TM: '181333'
(0008, 0014) Instance Creator UID             UI: 2.16.840.1.114337
(0008, 0016) SOP Class UID                    UI: RT Plan Storage
(0008, 0018) SOP Instance UID                 UI: 2.16.840.1.114337.1.1.1682093613.0
(0008, 0020) Study Date                       DA: '20221123'
(0008, 0030) Study Time                       TM: '130930'
(0008, 0050) Accession Number                 SH: ''
(0008, 0060) Modality                         CS: 'RTPLAN'
(0008, 0070) Manufacturer                     LO: 'CMS, Inc.'
(0008, 0090) Referring Physician's Name       PN: ''
(0008, 1070) Operators' Name                  PN: ''
(0008, 1090) Manufacturer's Model Name        LO: 'Monaco'
(0010, 0010) Patient's Name                   PN: 'CIRS Invivo statico^1cm sotto SW'
(0010, 0020) Patient ID                       LO: 'CIRSin vivo'
```

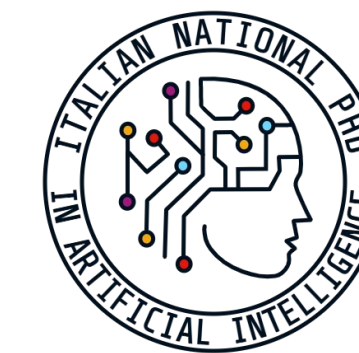


# RT



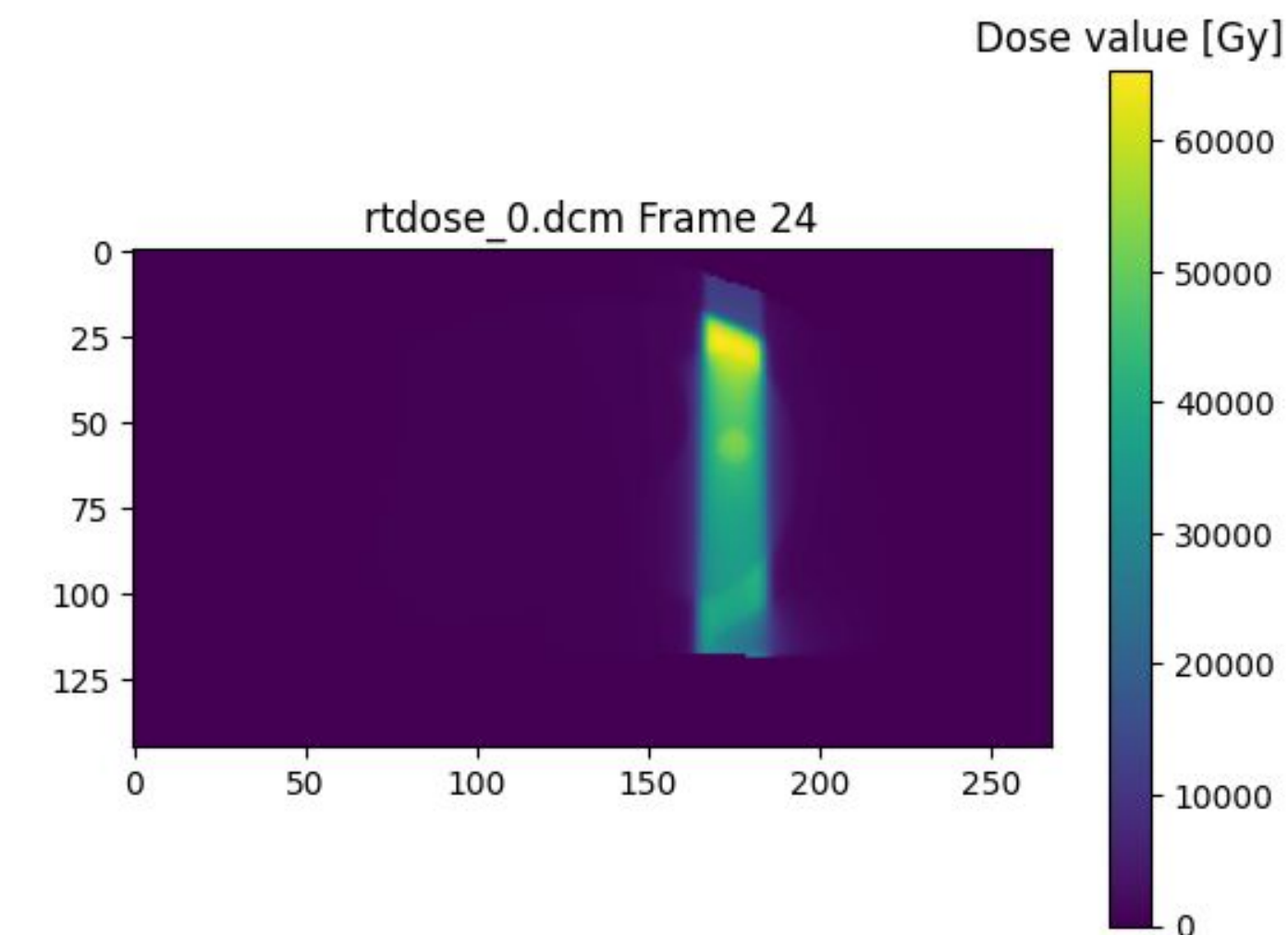
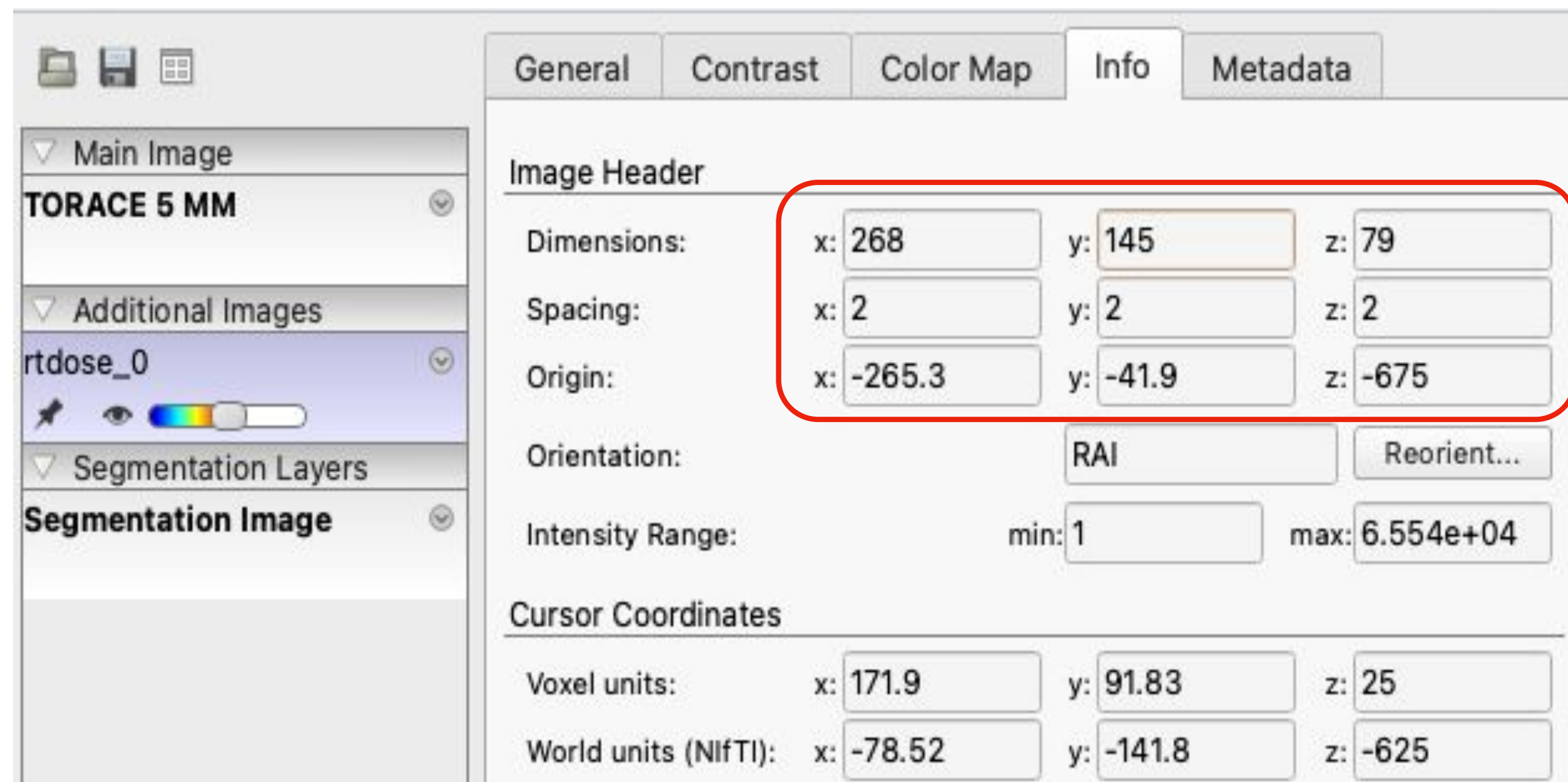
## RT Structure Set CIOD

The focus for this Radiotherapy Structure Set IOD (RT Structure Set IOD) is to address the **requirements for transfer of Patient structures and related data defined on CT scanners**, virtual simulation workstations, treatment planning systems and similar devices.

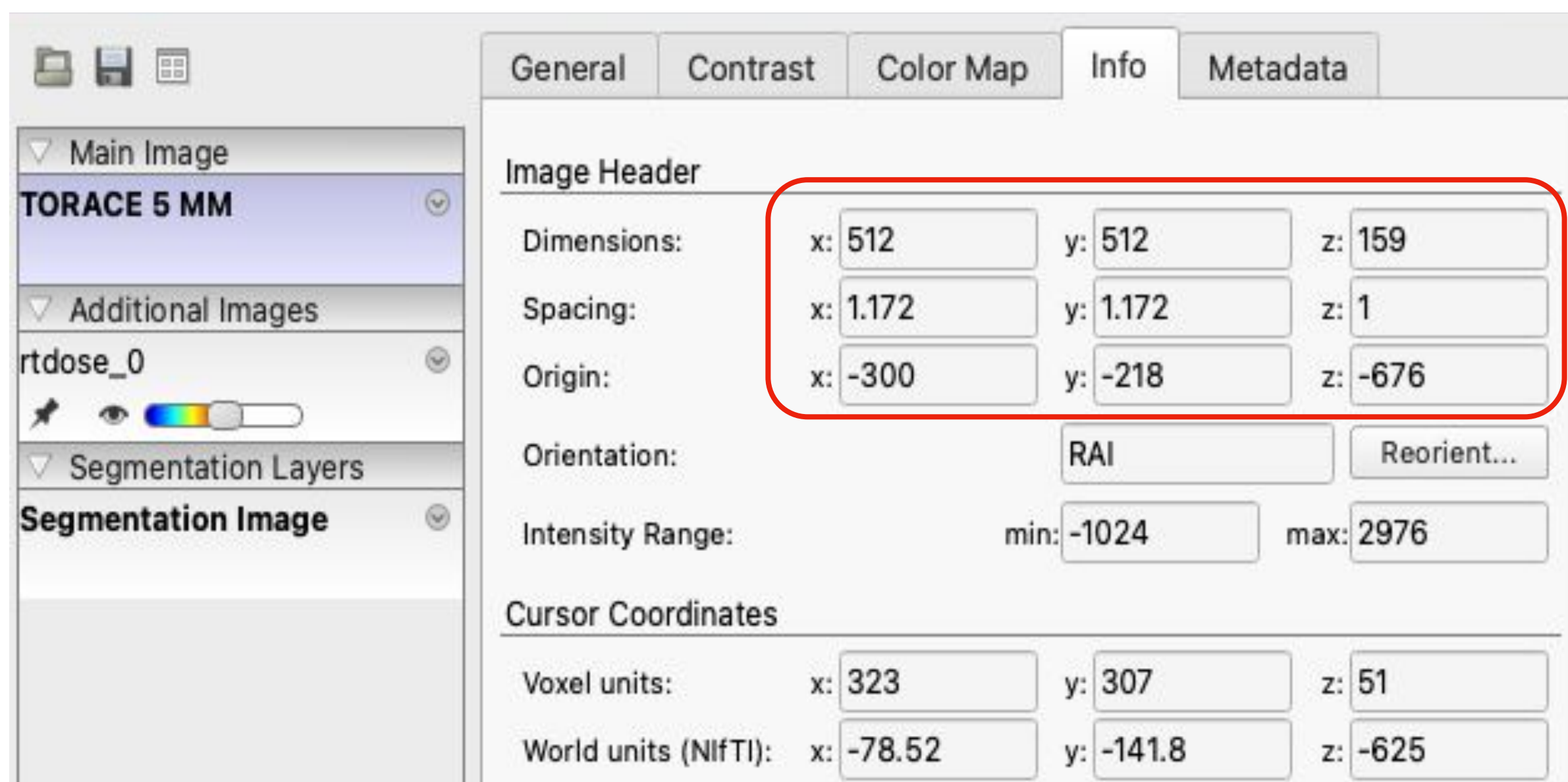
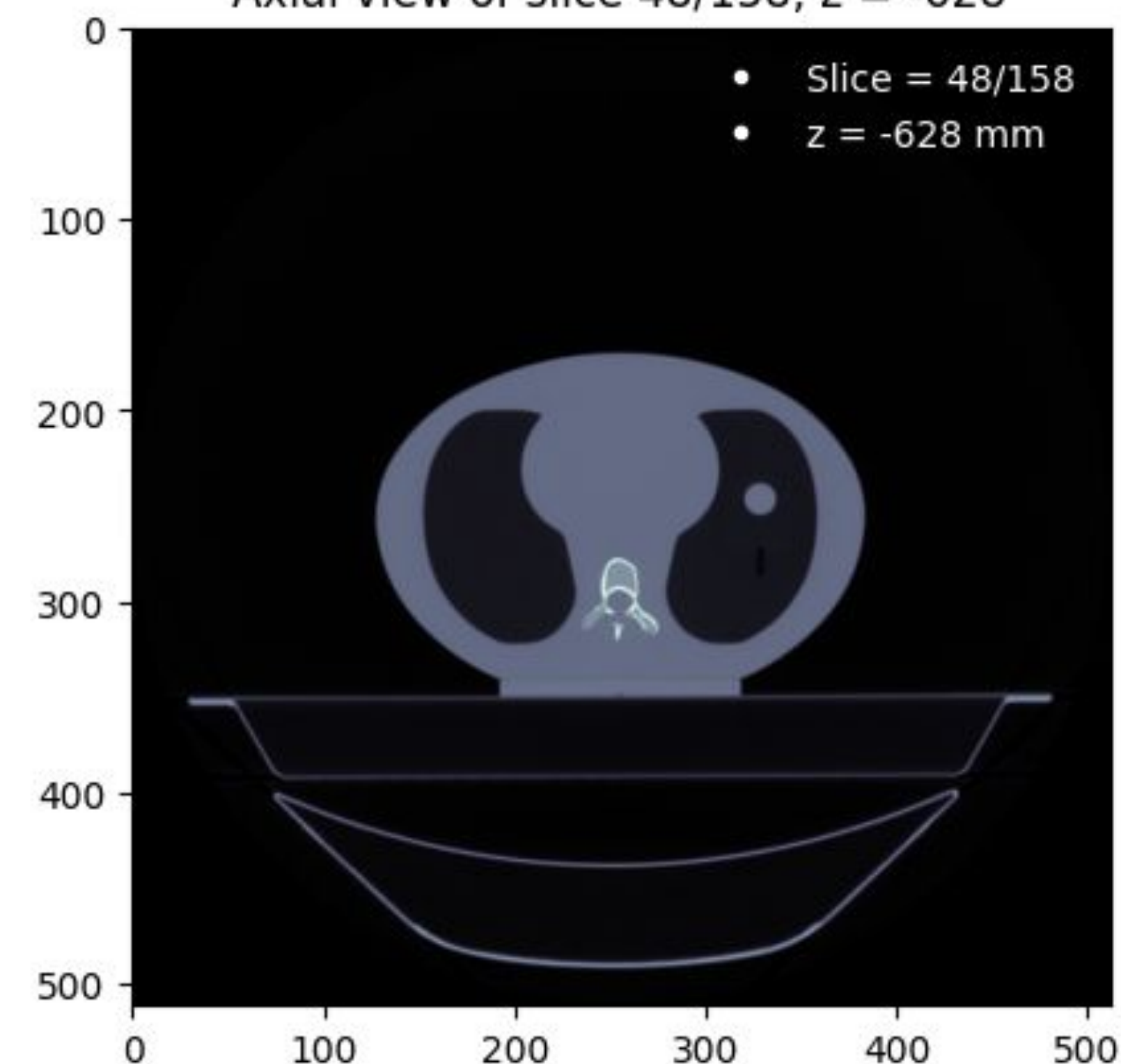


Dataset.file_meta -----		
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(0002, 0001)	File Meta Information Version	OB: b'\x00\x01'
(0002, 0002)	Media Storage SOP Class UID	UI: RT Structure Set Storage
(0002, 0003)	Media Storage SOP Instance UID	UI: 2.16.840.1.114337.1.1.1682093605.0
(0002, 0010)	Transfer Syntax UID	UI: Implicit VR Little Endian
(0002, 0012)	Implementation Class UID	UI: 2.16.840.1.114362.1
(0002, 0013)	Implementation Version Name	SH: 'MIM723M31402'
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(0008, 0005)	Specific Character Set	CS: 'ISO_IR 100'
(0008, 0012)	Instance Creation Date	DA: '20230421'
(0008, 0013)	Instance Creation Time	TM: '181329.000000'
(0008, 0014)	Instance Creator UID	UI: 2.16.840.1.114337
(0008, 0016)	SOP Class UID	UI: RT Structure Set Storage
(0008, 0018)	SOP Instance UID	UI: 2.16.840.1.114337.1.1.1682093605.0
(0008, 0020)	Study Date	DA: '20221123'
(0008, 0030)	Study Time	TM: '130930'
(0008, 0050)	Accession Number	SH: '1'
(0008, 0060)	Modality	CS: 'RTSTRUCT'
(0008, 0070)	Manufacturer	LO: 'Computerized Medical Systems'
(0008, 0090)	Referring Physician's Name	PN: 'Unknown'
(0008, 1040)	Institutional Department Name	LO: 'EpiGray'
(0008, 1090)	Manufacturer's Model Name	LO: 'Monaco'
(0010, 0010)	Patient's Name	PN: 'CIRS Invivo statico^1cm sotto SW'
(0010, 0020)	Patient ID	LO: 'CIRSin vivo'
(0010, 0030)	Patient's Birth Date	DA: '20021122'

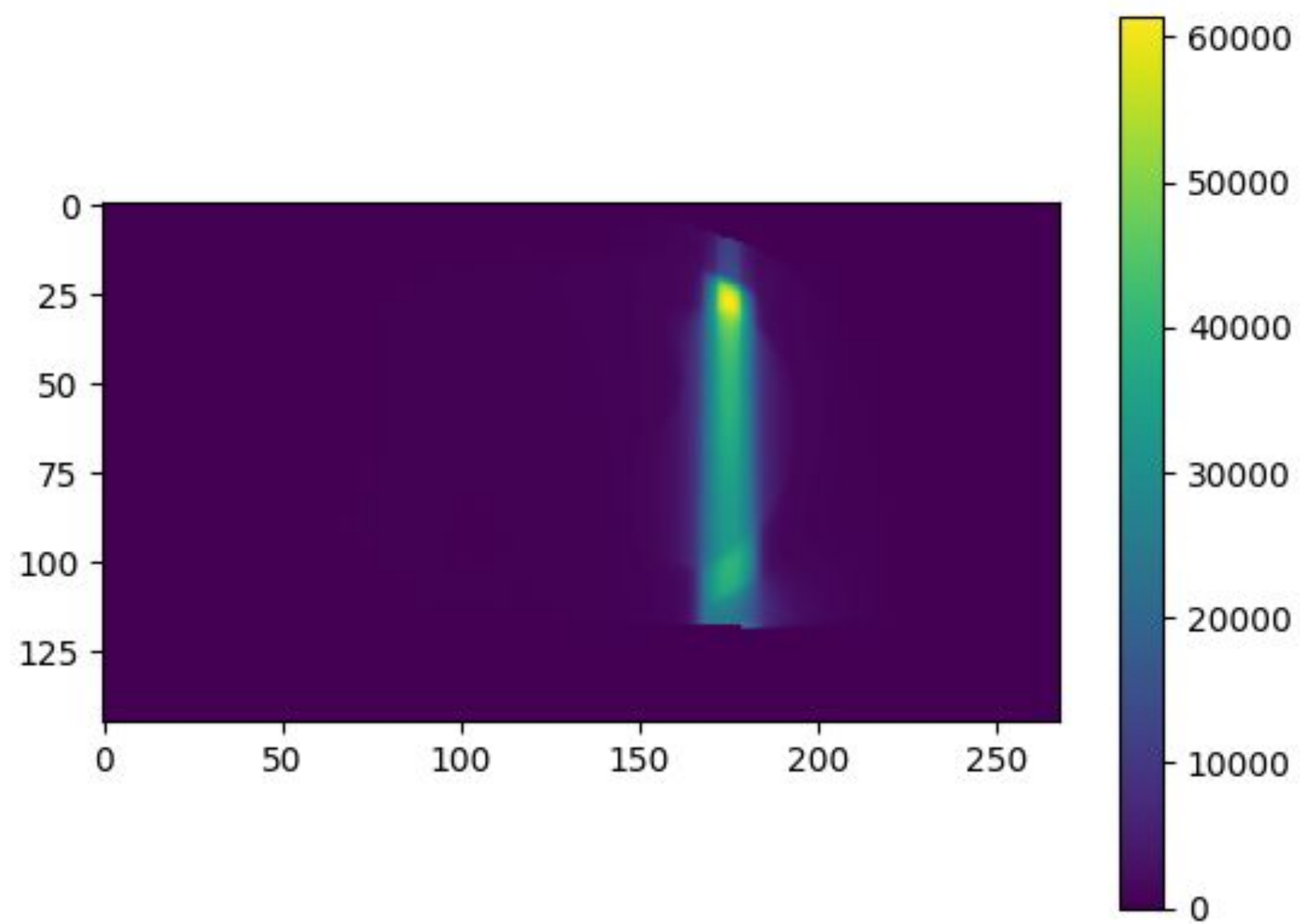




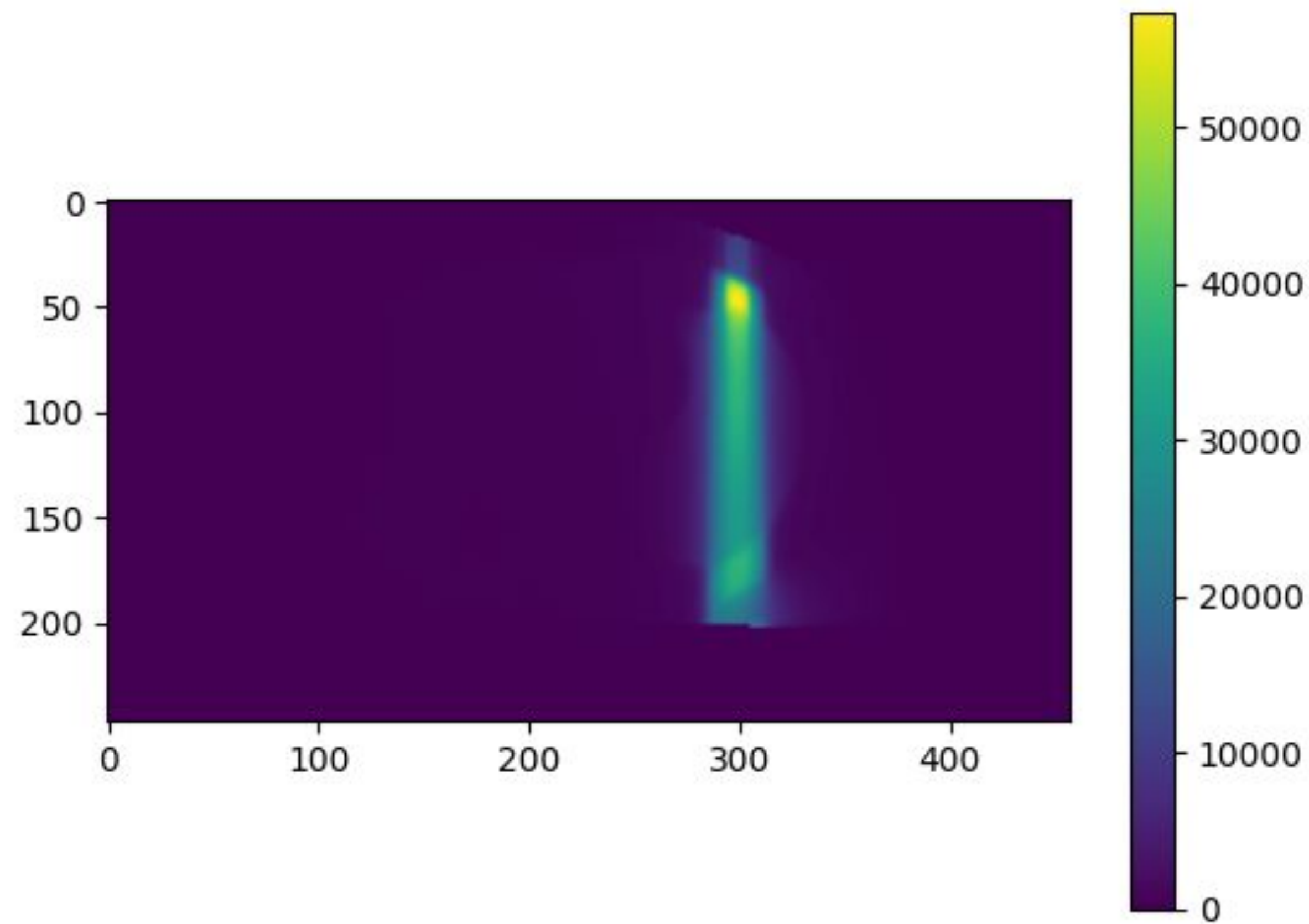
Axial view of slice 48/158, z = -628



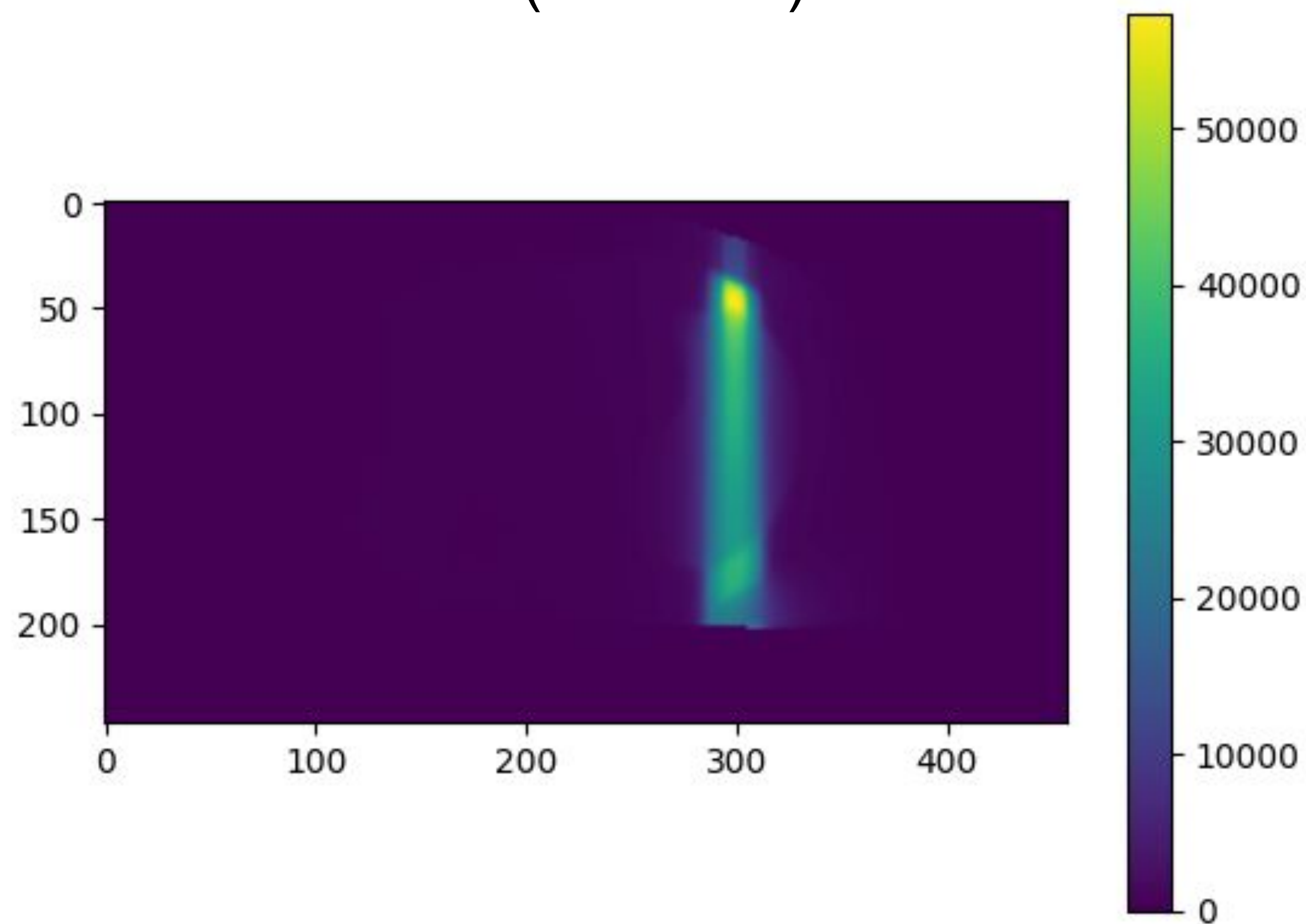
Original image (128x268)



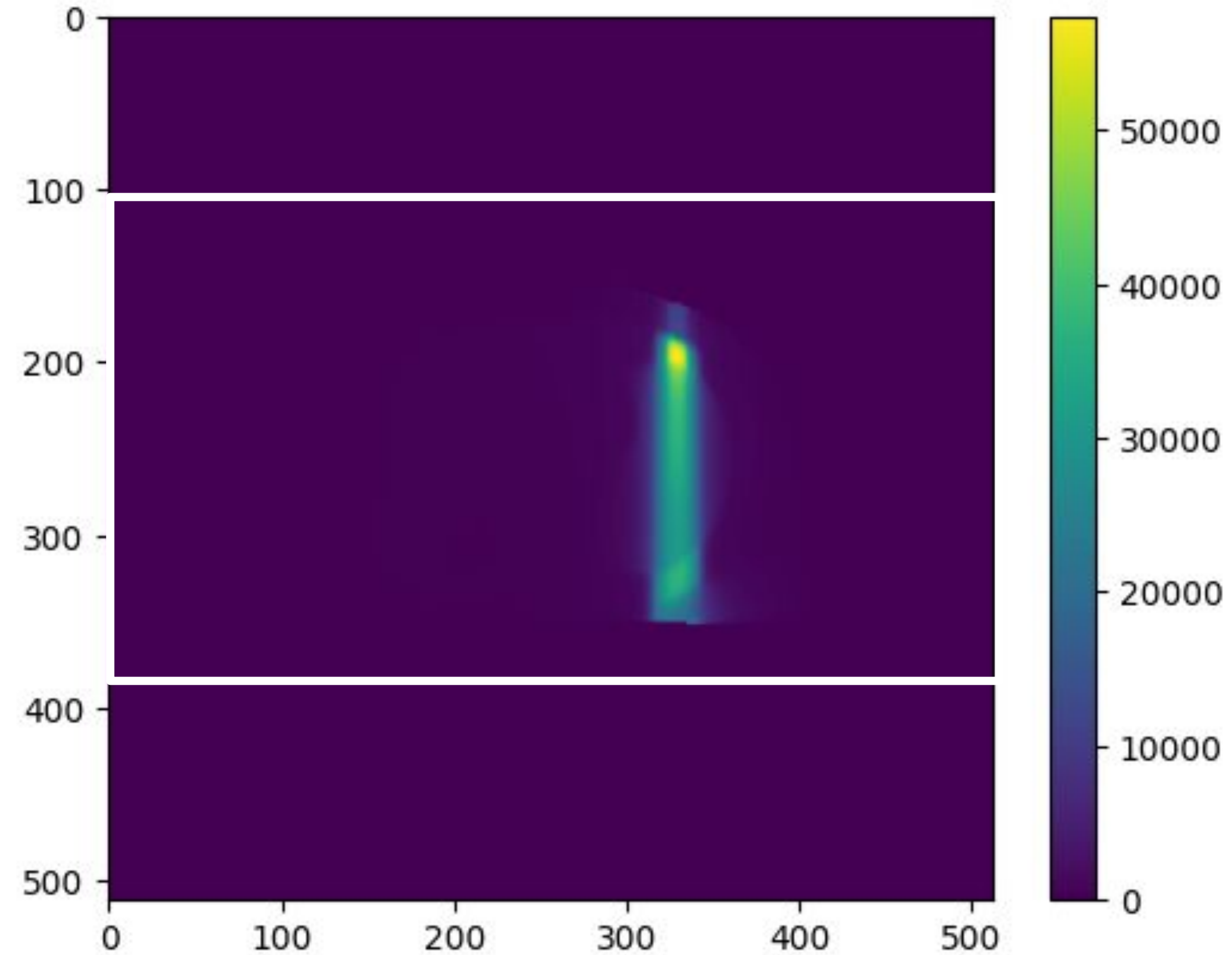
Resized (247x457)



Resized (247x457)

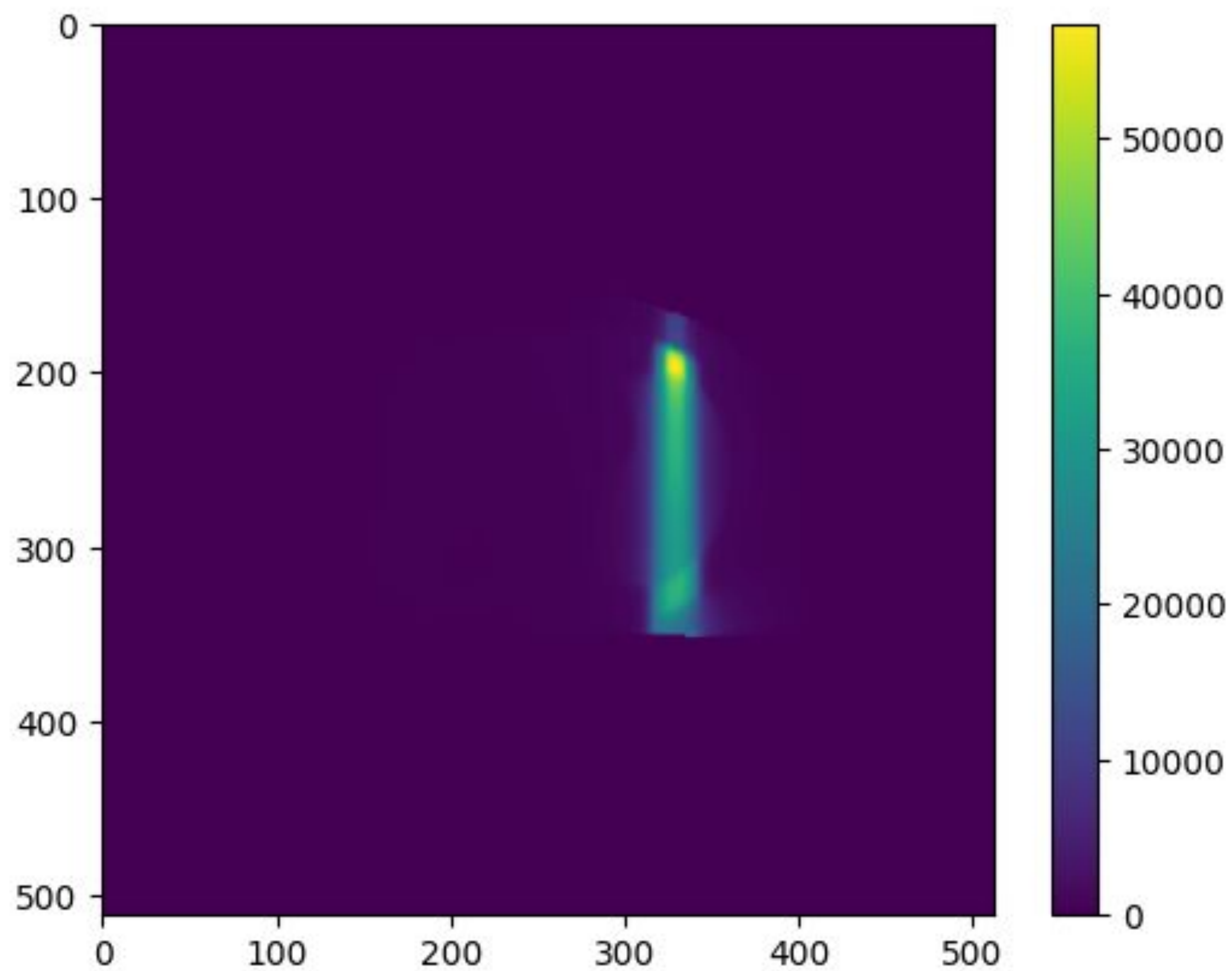


Overlay on a zeros matrix (512x512)





Overlay on a zeros matrix (512x512)



Overlay on a CT slice (512x512)

