

Invited plenary talk on April 13, 2022  
2022 annual meeting of  
the Council on Ionizing Radiation Measurements and Standards (CIRMS)

# **Radiotherapy Plan QA using Deep-CNN based Multi-OAR Autosegmentation and GPU- accelerated Monte Carlo Dose Check**

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# Disclosures:

Developer of commercial software tools:



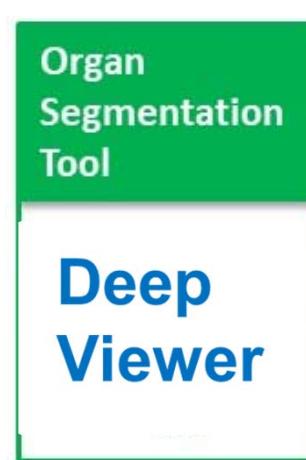
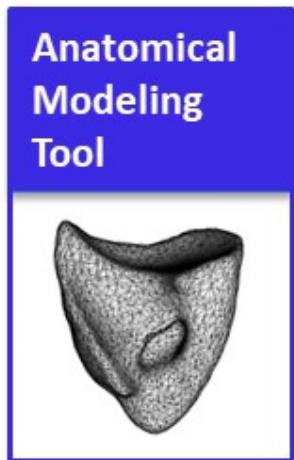
<https://www.virtual-dose.com>

## DeepViewer



<http://www.wisdom-tech.online/>

# Outline of the presentation

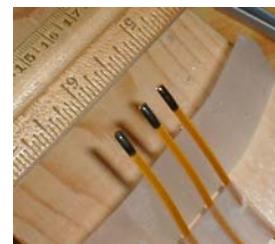


Over the years, we have developed a number of major software tools

# Two Ways to Determine Organ Doses

## Measurements

- Dosimeters
- Physical phantom



## Monte Carlo Simulations

- Computational phantoms
- Monte Carlo codes



AP



PA



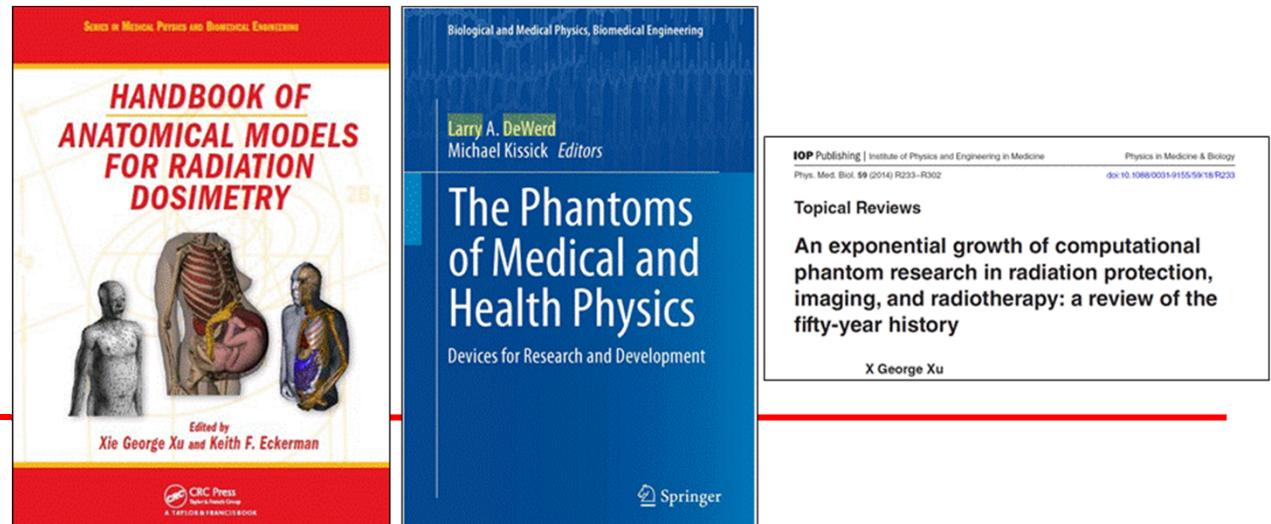
RLAT



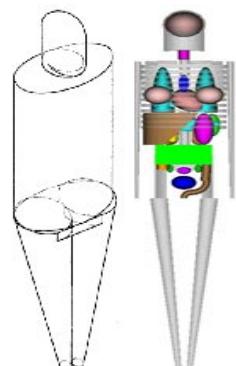
LLAT

# 60-Year History of Computational Phantoms

- Radiation Protection
- Medical Imaging
- Radiotherapy



## 1<sup>st</sup> Generation STYLIZED



ORNL family models  
1960-1980s

## 2nd Generation VOXEL

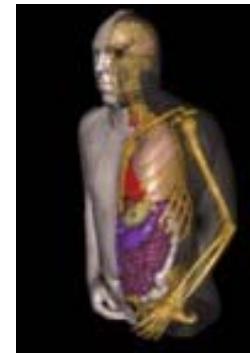
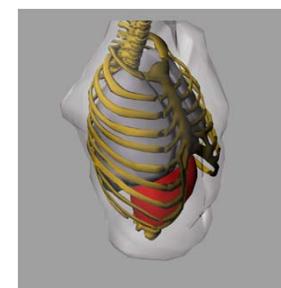


Image-based  
1980s-present

## 3rd Generation BREP



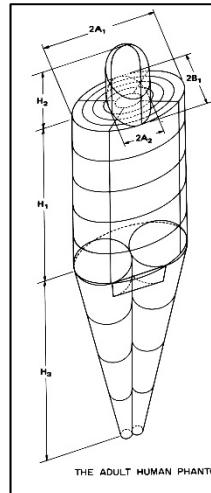
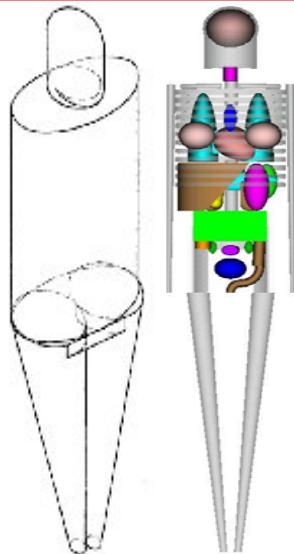
Deformable 4D models  
2000s-present

- Personalization
- Multi-scale  
(voxel – DNA)

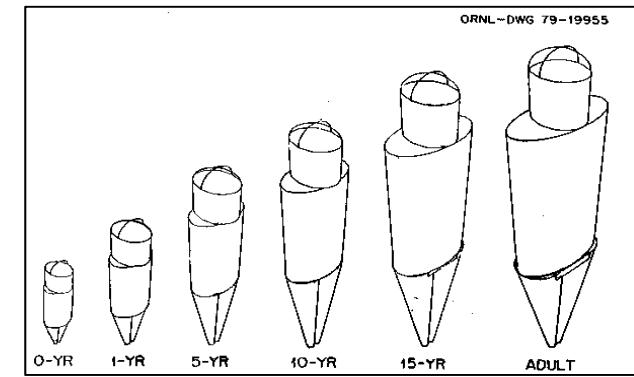
# 1st-Generation “Stylized” Phantoms

(Society of Nuclear Medicine’s MIRD Committee)

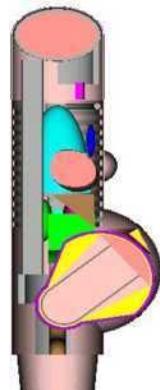
*Anatomically simple and friendly for computers prior to 1980s*



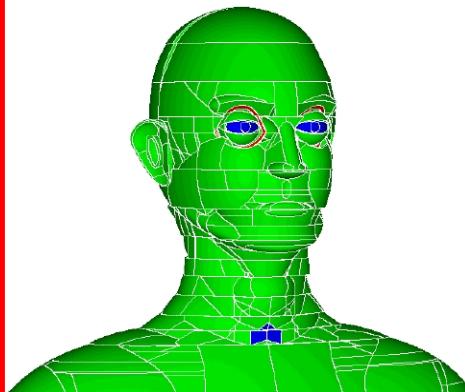
PHANTOM DIMENSIONS AND DOSE REGIONS							
Age (yr)	Weight (kg)	H <sub>1</sub> (cm)	H <sub>2</sub> (cm)	H <sub>3</sub> (cm)	A <sub>1</sub> (cm)	B <sub>1</sub> (cm)	A <sub>2</sub> (cm)
0	3.148	23	13	16	5.5	5	4.5
1	9.112	33	16	28.8	8	7	6.5
5	18.12	45	20	46	11	7.5	6.5
10	30.57	54	22	64	14	8	6.5
15	53.95	65	23	78	18	9	7
Adult	69.88	70	24	80	20	10	7



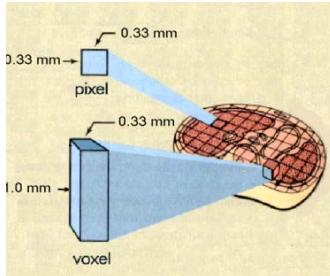
Snyder *et al* (1978), Cristy (1980), and Cristy and Eckerman (1987), plus ADAM/Eva phantoms by Kramer *et al* (1982) from Germany



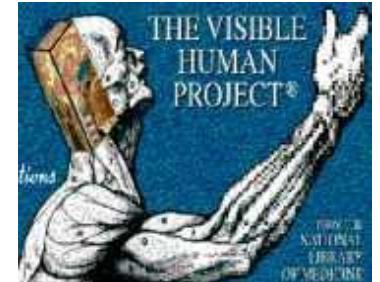
At the end of each trimester of pregnancy  
(Stabin *et al* 1995)



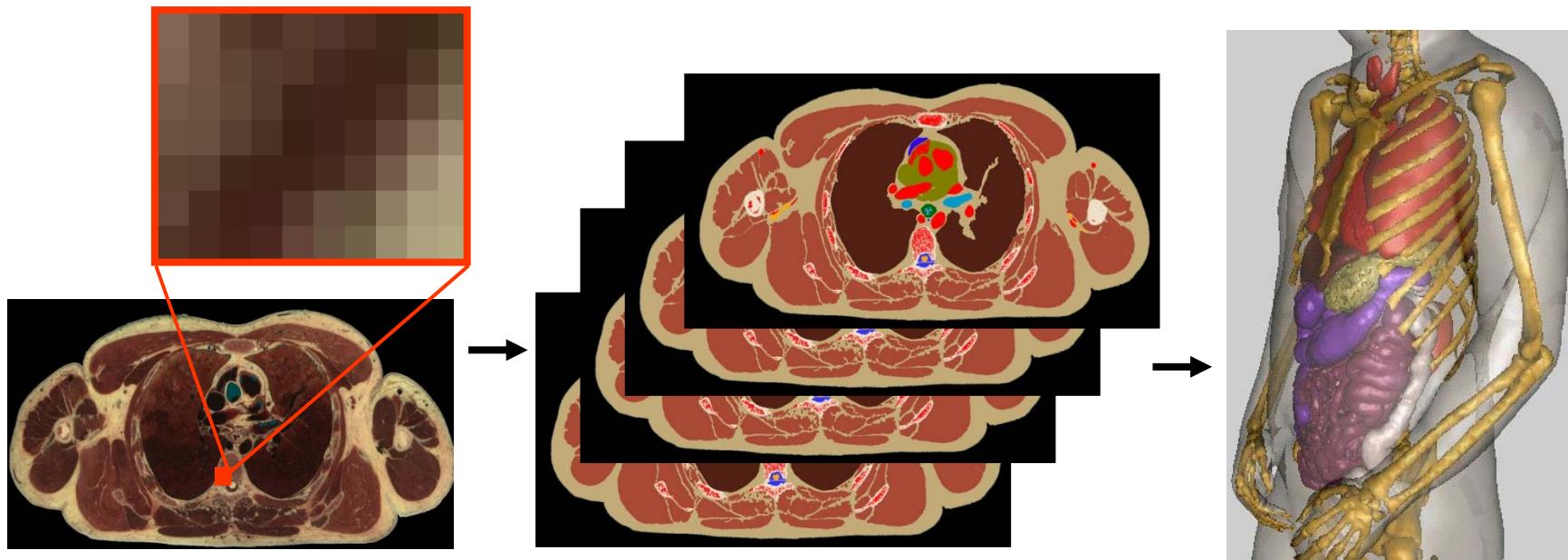
- The Computational Anatomical Man (CAM) and
- CAF (Computerized Anatomical Female)
- by Billings and Yucker (1973)
- Used exclusively by NASA



## 2<sup>nd</sup>-Generation “Voxel” Phantoms - Example of the VIP-Man (1997-2000)



Xu XG, Chao TC, Bozkurt A. VIP-Man: An image-based whole-body adult male model constructed from color photographs of the Visible Human Project for multi-particle Monte Carlo calculations. *Health Phys.*, 78(5):476-486, 2000. ***One of the most cited (450+)***



Identification of organs in each slice of a 2D pixel map

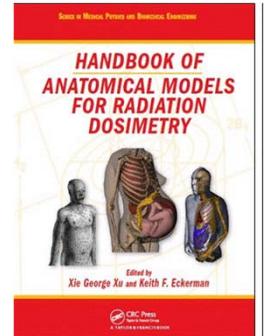
Registration of all slices

Finished 3D voxel VIP-Man

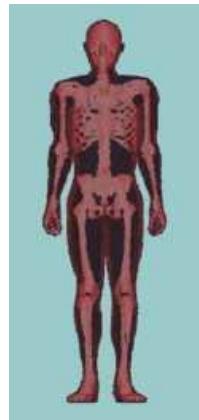
*At the time, organ segmentation was done manually*

# <<Handbook of Anatomical Models for Radiation Dosimetry>> by Xu and Eckerman 2009

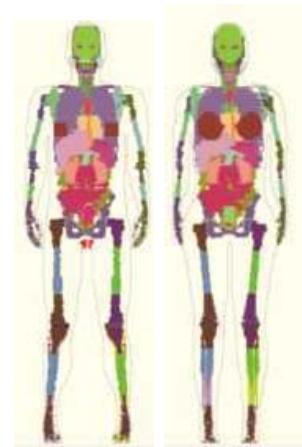
*- curtsey images*



REX & REGINA (ICRP)



NORMAN



MAX06 FAX06



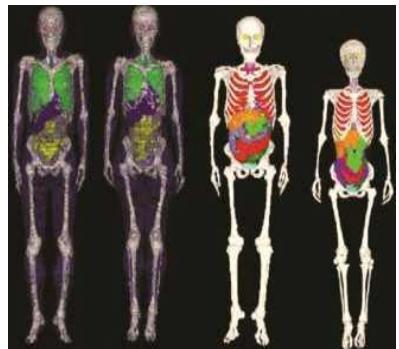
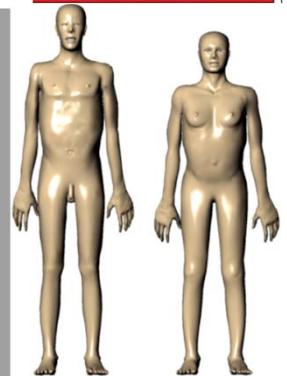
Zubal



NCAT



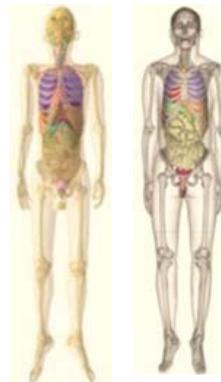
VIP-Man, Pregnant, Adult M/F



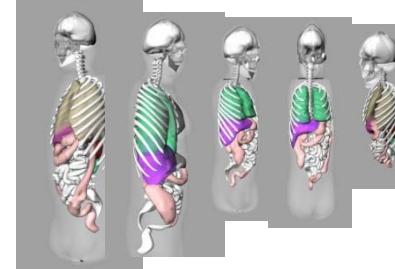
Otoko Onago JM KF



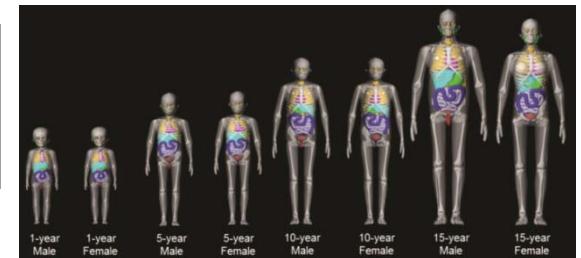
KTMAN 1, 2



CNMAN VCH



Vanderbilt Family



UF Family

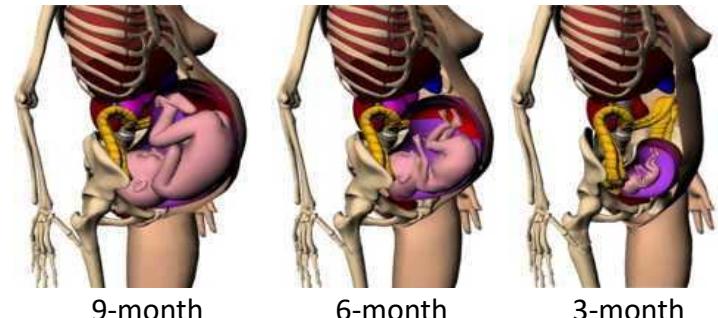
*At the time, organ segmentation was done manually*

# 3rd-Generation “BREP” Phantoms (NURBS or Meshes)

## Pregnant Phantoms

Xu X G, Taranenko V, Zhang J, Shi C. A boundary-representation method for designing whole-body radiation dosimetry models: pregnant females representing three gestational periods, RPI-P3, -P6 and -P9. *Phys. Med. Biol.* (2007)

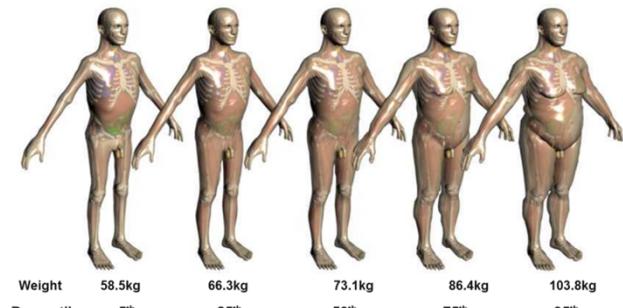
*The Best 10 papers by PBM in 2007*



## Size and Weight Adjustable Phantoms

Na YH, Zhang\* B, Zhang\* J, Caracappa PF, Xu XG. Deformable Adult Human Phantoms for Radiation Protection Dosimetry: Anatomical Data for Covering 5th- 95th Percentiles of the Population and Software Algorithms. *Phys. Med. Biol.* 55: 3789-3811 (2010).

• Same height (e.g. 176cm Male), but different weights:



The skin meshes from an open source software, MakeHuman™ version 0.9.1 RC11 (<http://www.makehuman.org/>)

## Obese Individuals

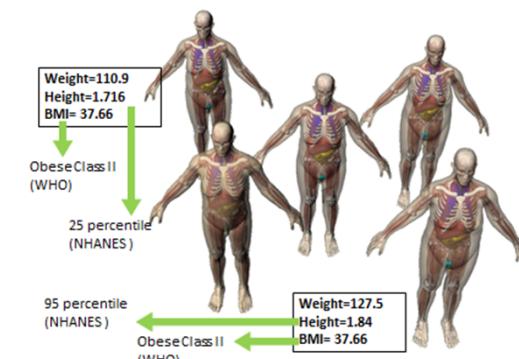
Ding A, Mille MM, Liu T, Caracappa PF and Xu XG. *Phys. Med. Biol.* 57:2441–2459 (2012).

*One of the most downloaded in 2012*

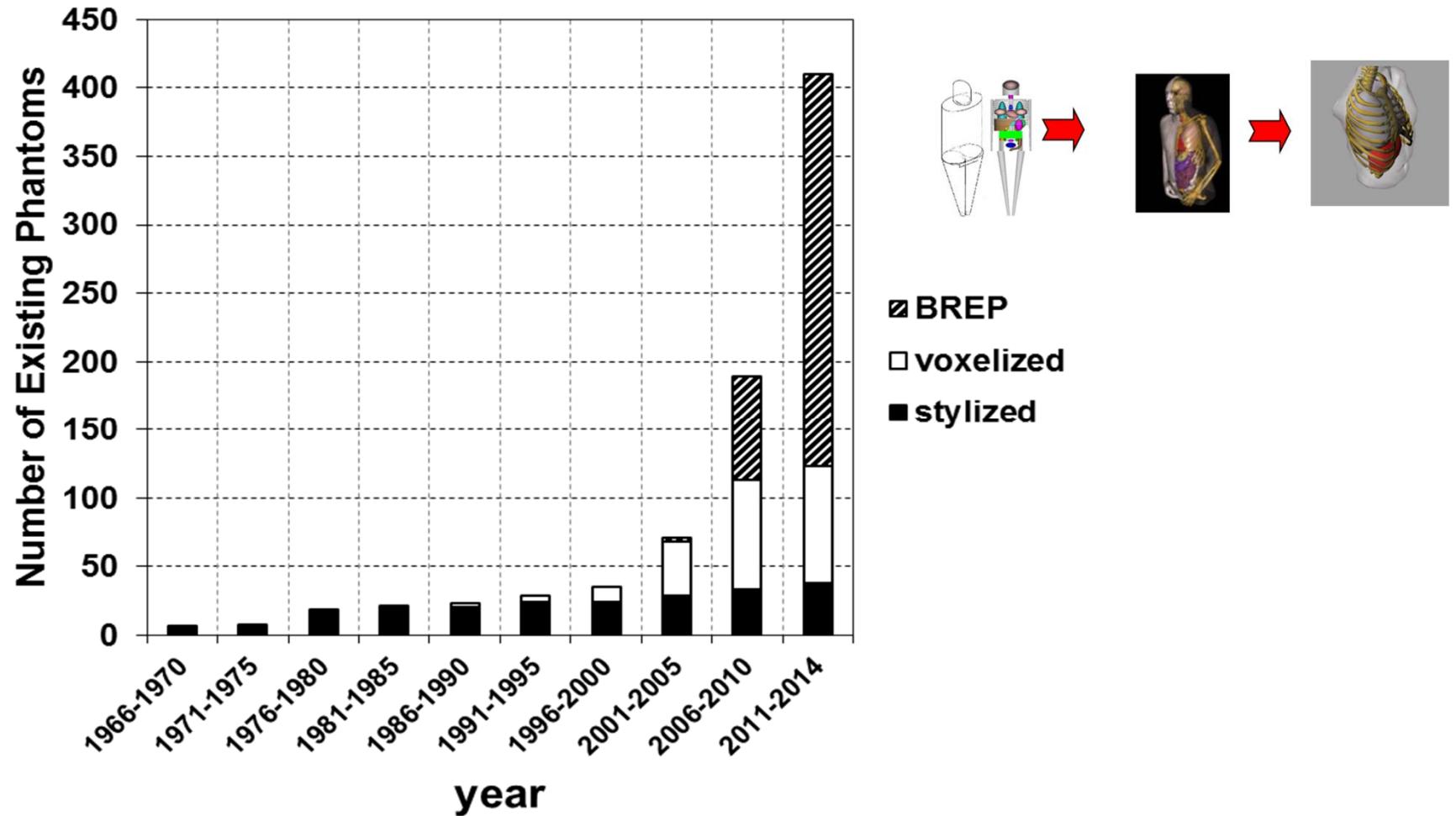


Weight Category	BMI
Underweight	< 18.5
Normal Weight	18.5-24.9
Overweight	25-29.9
Obese (I, II)	30-34.9, 34.9-39.9
Morbidly Obese	>40

## Obese Phantoms

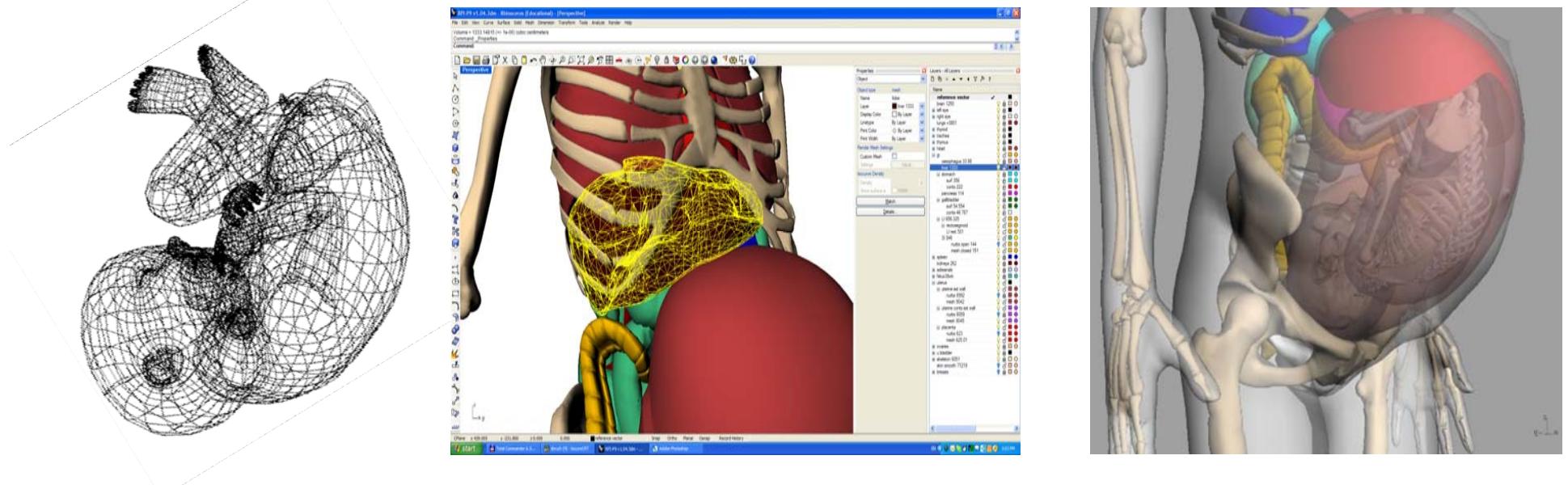
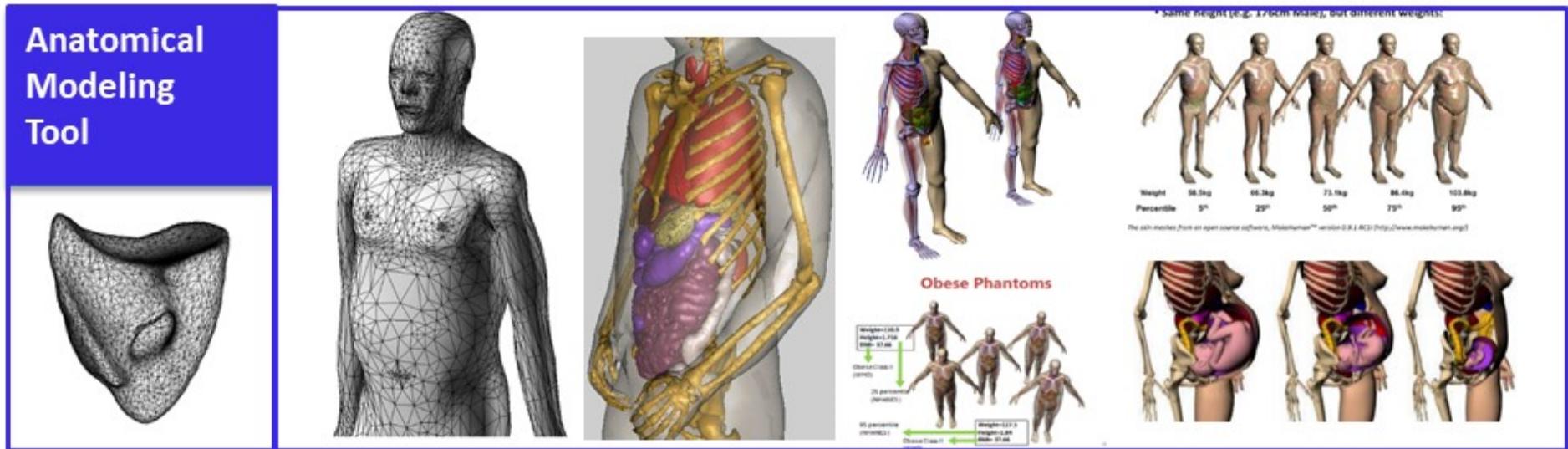


# The number of phantoms in existence since 1966 shows a surprising exponential growth



X. George Xu, "An exponential growth of computational phantom research in radiation protection, imaging, and radiotherapy: a review of the fifty-year history," [Physics in Medicine & Biology](#), 59(R233-R302) 2014.

# Patient geometric modeling tools



# Future Radiobiological Predictive Modeling at Multi-Scale

## - To Bridge the Gap Between Voxel and DNA

### Geometric modeling, DNA

★ Offer complex cell geometries from whole cell to nuclear base

★ Easy to combine geometries to simulate desired accuracy

Nucleus

Cells

Osteocyte within the bone matrix

Osteoblasts

Osteoclast with multiple nuclei

Bone matrix

5

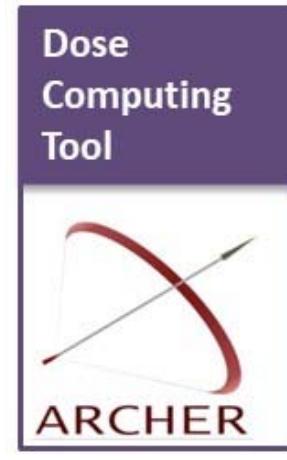
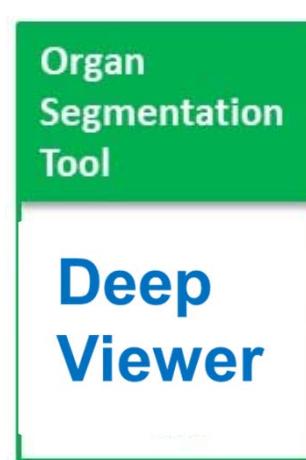
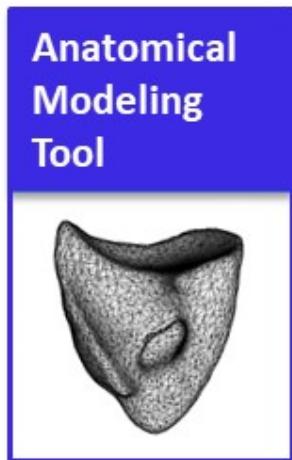
A. McNamara et al.  
PMB 2018;63(17).

J. Schuemann, AAPM 2019

MASSACHUSETTS GENERAL HOSPITAL RADIATION ONCOLOGY HARVARD MEDICAL SCHOOL

Courtesy slide from Jan Schuemann

# Outline of the presentation



# Cancer risks of pediatrics from CT

*by Pearce, et al. Lancet 2012*

- 178,604 young patients
- CT scans from 1985 -2002
- 81 hospitals in Great Britain
- Cancer data from 1985 - 2008



	Male patients		Female patients	
	Brain dose (mGy)	Red bone marrow dose (mGy)	Brain dose (mGy)	Red bone marrow dose (mGy)
<b>Age at brain CT</b>				
0 years	28	8	28	8
5 years	28	9	28	9
10 years	35	6	35	6
15 years	43	4	44	6
20 years	35	2	42	2
<b>Age at chest CT</b>				
0 years	0.4	4	0.4	4
5 years	0.3	3	0.3	3
10 years	0.3	3	0.3	3
15 years	0.2	4	0.3	4
20 years	0.2	4	0.3	4
<b>Age at abdominal CT</b>				
0 years	0.2	3	0.2	3
5 years	0.1	2	0.1	2
10 years	0.1	3	0.1	3
15 years	0.0	3	0.0	3
20 years	0.0	3	0.0	4
<b>Age at extremity CT</b>				
0 years	0.0	1	0.0	1
5 years	0.0	0.2	0.0	0.2
10 years	0.0	0.1	0.0	0.1
15 years	0.0	0.0	0.0	0.0
20 years	0.0	0.0	0.0	0.0

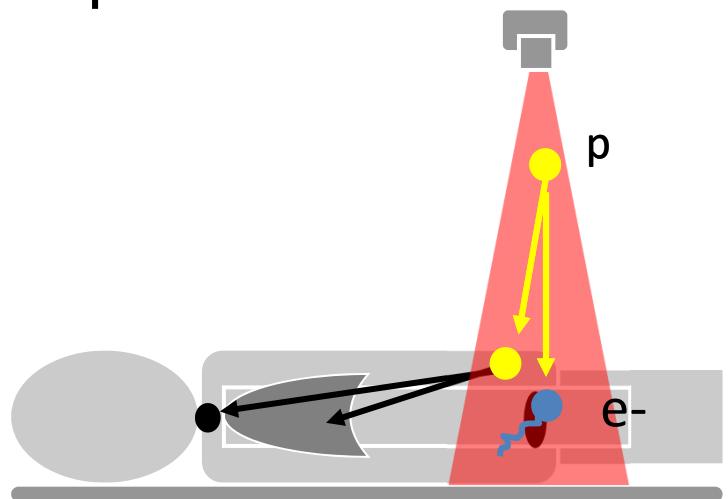
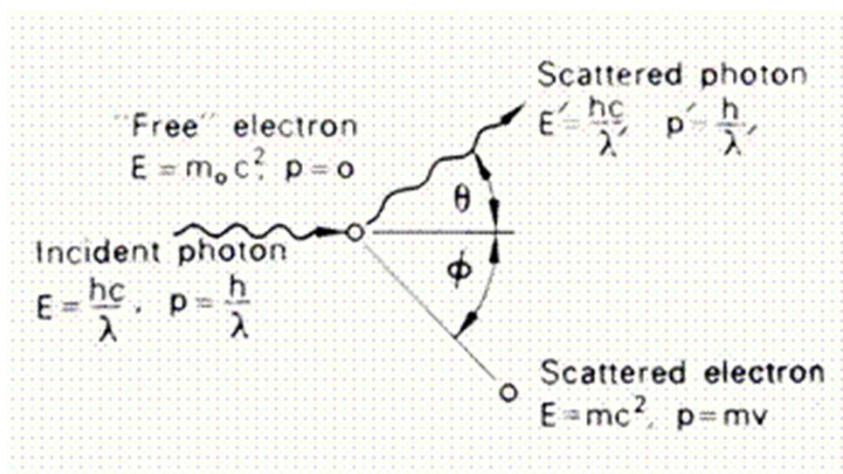
Table 1: Estimated radiation doses to the brain and red bone marrow from one CT scan, by scan type, sex, and age at scan, as used in this study for scans after 2001

# Radiation Physics for CT Scans

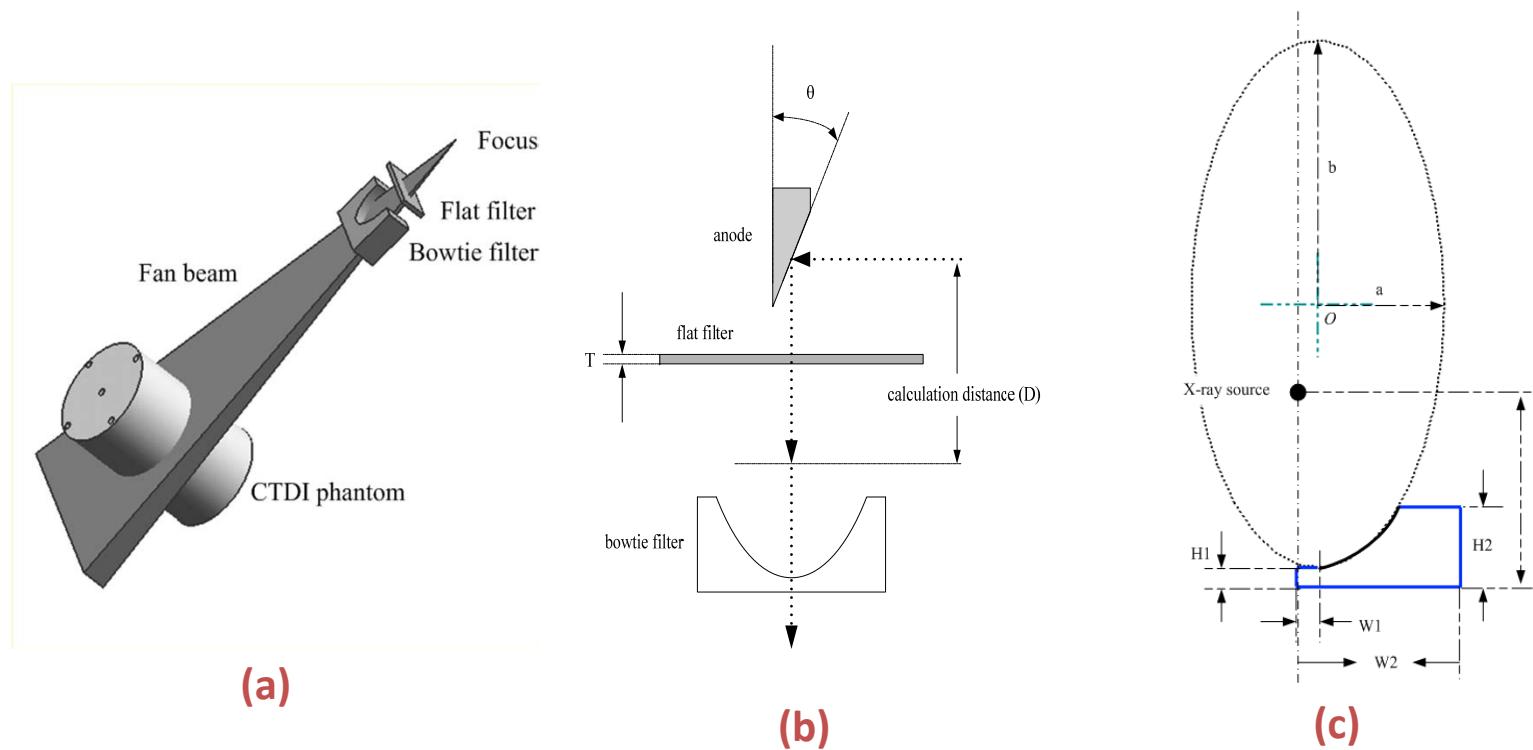
## X-ray photon interactions (< 160 keV)

✓ Photoelectric effect

✓ Compton scattering     $E_{pe} = hf - \phi$



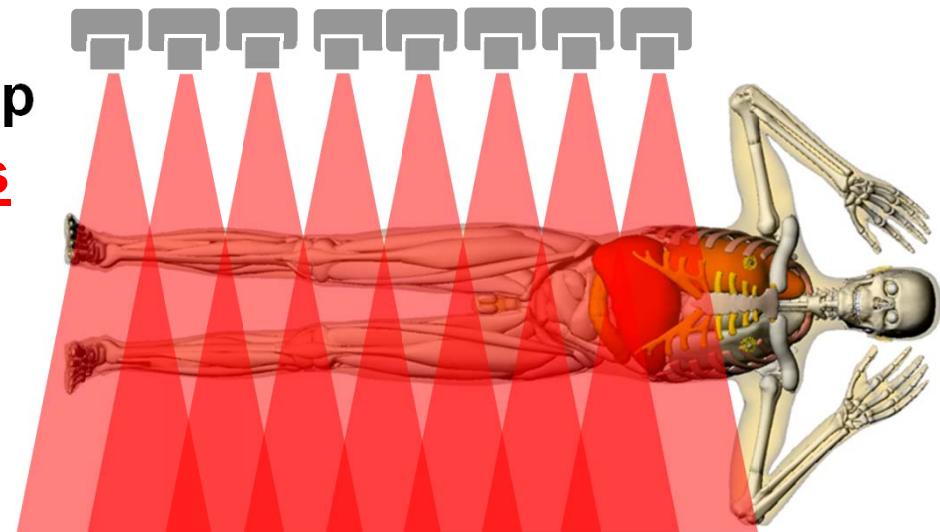
# Method: CT Scanner modeling methodology



Ding A, Gao Y, Liu H , Caracappa PF, Long DJ, Bolch WE, Liu Bob, Xu XG. VirtualDose: A New CT Dose Reporting Software for Adult and Pediatric Patients. Phys. Med. Biol.. 60(14):5601-5625 (2015).

# A Comprehensive Slice-by-slice Organ Dose Database

- Axial scan simulations in MCNPX
- Contiguous scans from the top to the bottom of 27 phantoms
- CT technical parameters
  - 4 different tube voltages: 80, 100, 120, and 140 kVp
  - 4 different beam collimations: 1.25, 5, 10 and 20 mm
  - Using both the head and body bowtie filters



Ding A, Gao Y, Liu H , Caracappa PF, Long DJ, Bolch WE, Liu Bob, Xu XG. VirtualDose: A New CT Dose Reporting Software for Adult and Pediatric Patients. Phys. Med. Biol.. 60(14):5601-5625 (2015).

# Database Hosted in SQL Data Server

<http://www.virtual-dose.com>

Organ name

Slice position

The screenshot shows a Microsoft SQL Server Management Studio window with a table titled 'ROSE\_VDataProv\_ady\_10mm\_140kvp' and 'ROSE\_VDataProv\_Body\_20mm\_80kvp'. The table has 15 columns labeled F2 through F16. The first column, 'Organ/tissue', contains various organ names like Adrenals, Bladder\_wall, Brain, Breasts, Esophagus, Eye\_lens, Eyeballs, Fetal\_brain, Fetal\_skeleton, Fetal\_soft\_tiss., Fetus\_total, Gallbladder, Heart\_wall, Kidneys, LI\_ctons, LI\_wall, Liver, Lungs, Ovaries, Pancreas, Placenta, Remainder, SI\_wall\_and..., Skeleton, Skin, Spleen, Stomach\_wall, and Thymus. The data values are represented in scientific notation. Red arrows point from the 'Organ name' text to the 'Organ/tissue' column and from the 'Slice position' text to the F12-F16 columns. A blue arrow points to the text 'Organ dose results' located in the bottom right corner of the table area.

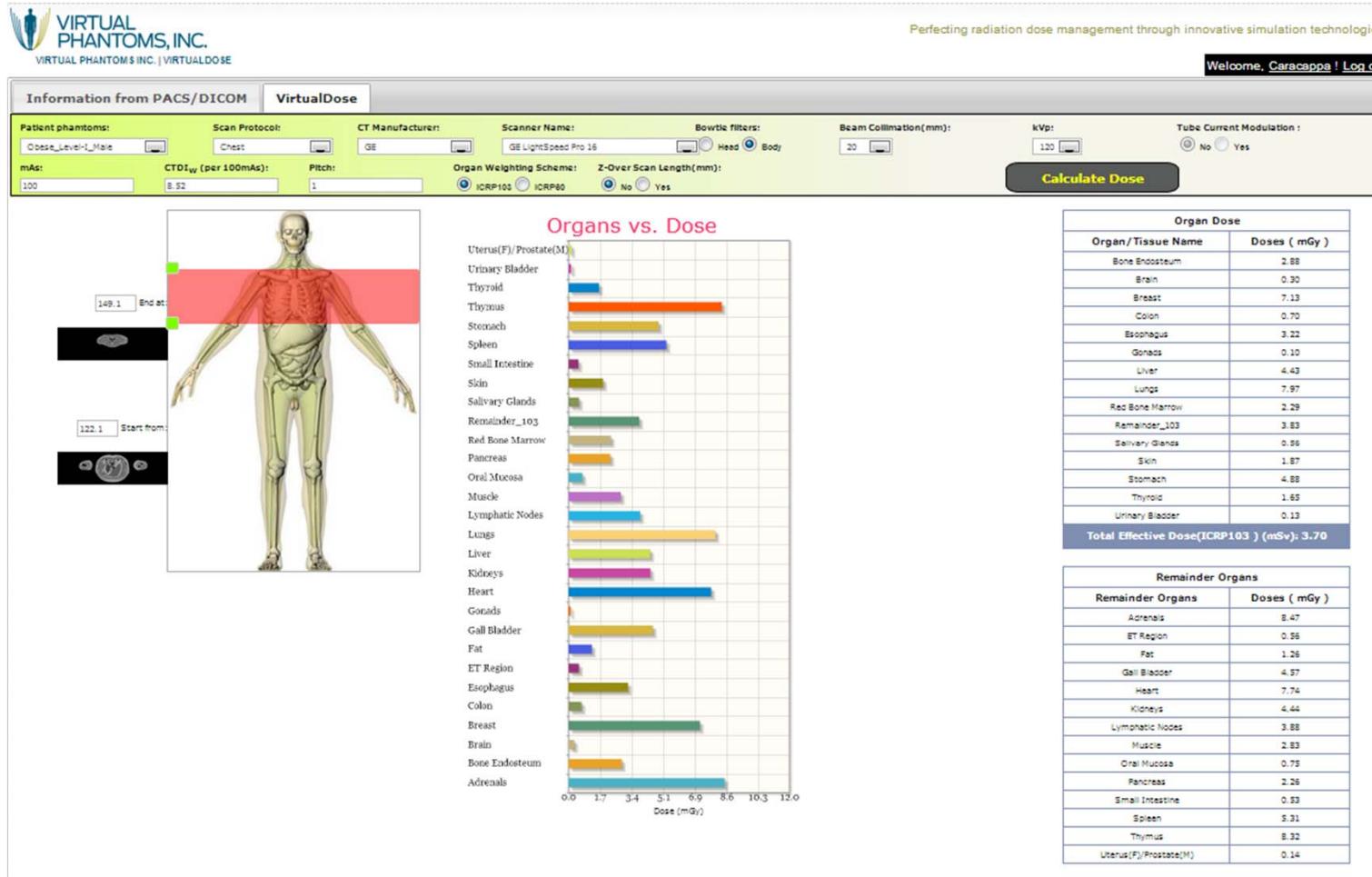
Organ/tissue	F2	F3	F4	F5	F6	F7	F8	F9	F10	F11	F12	F13	F14	F15
Adrenals	3.8E-10	2.64E-10	3.35E-10	3.6E-10	4.01E-10	3.09E-10	4.29E-10	3.64E-10	3.07E-10	2.7E-10	6.02E-10	5.4E-10	6.64E-10	5.85E-10
Bladder_wall	4E-09	5.04E-09	5.28E-09	4.91E-09	5.4E-09	5.35E-09	4.81E-09	5.25E-09	5.08E-09	5.31E-09	5.36E-09	5.41E-09	5.59E-09	5.92E-09
Brain	6.44E-11	9.77E-11	7.37E-11	7.42E-11	1.03E-10	1.13E-10	8.91E-11	1.04E-10	1.03E-10	1.31E-10	1.06E-10	1.04E-10	9.46E-11	8.99E-11
Breasts	6.84E-10	6.97E-10	7.1E-10	7.27E-10	7.3E-10	7.72E-10	7.95E-10	7.71E-10	7.99E-10	8.2E-10	8.48E-10	8.84E-10	9.38E-10	8.84E-10
Esophagus	2.14E-10	3.58E-10	2.4E-10	1.54E-10	1.74E-10	2.34E-10	2.56E-10	1.71E-10	2.69E-10	2.97E-10	2.49E-10	2.79E-10	3.02E-10	3.17E-10
Eye_lens	9.26E-10	4.41E-10	7.68E-10	1.11E-09	7.7E-10	1.41E-09	1.73E-09	0	1.29E-09	1.31E-09	5.12E-10	3.14E-10	0	1.45E-10
Eyeballs	3.67E-10	3.04E-10	2.89E-10	3.66E-10	4.15E-10	2.93E-10	3.44E-10	2.53E-10	4E-10	2.73E-10	3.46E-10	4.13E-10	3.19E-10	4.48E-10
Fetal_brain	2.02E-09	2.35E-09	2.59E-09	2.37E-09	2.39E-09	2.58E-09	2.21E-09	2.43E-09	2.55E-09	2.54E-09	2.58E-09	2.7E-09	2.57E-09	2.95E-09
Fetal_skeleton	7.67E-09	9.72E-09	1.01E-08	9.92E-09	9.77E-09	9.94E-09	9.26E-09	9.44E-09	9.58E-09	9.37E-09	9.64E-09	9.93E-09	1E-08	1.03E-08
Fetal_soft_tiss.	2.68E-09	3.52E-09	3.71E-09	3.6E-09	3.48E-09	3.4E-09	3.36E-09	3.24E-09	3.32E-09	3.32E-09	3.43E-09	3.55E-09	3.52E-09	3.53E-09
Fetus_total	2.6E-09	3.38E-09	3.58E-09	3.45E-09	3.35E-09	3.3E-09	3.22E-09	3.15E-09	3.23E-09	3.22E-09	3.33E-09	3.45E-09	3.41E-09	3.46E-09
Gallbladder	2.84E-10	3.29E-10	3.16E-10	2.63E-10	3.42E-10	4.42E-10	3.89E-10	3.33E-10	4.58E-10	5.43E-10	4.27E-10	3.94E-10	5.77E-10	3.58E-10
Heart_wall	2E-10	2.76E-10	2.07E-10	3.02E-10	2.93E-10	2.45E-10	2.73E-10	2.95E-10	3.1E-10	3.73E-10	2.63E-10	3.96E-10	3.5E-10	3.62E-10
Kidneys	5.04E-10	5.325E-10	4.05E-10	5.44E-10	5.02E-10	4.825E-10	5.095E-10	5.655E-10	5.145E-10	5.005E-10	6.45E-10	6.85E-10	6.265E-10	7.025E-10
LI_ctons	1.45E-09	1.6E-09	1.69E-09	1.74E-09	1.82E-09	1.81E-09	1.76E-09	1.73E-09	1.81E-09	1.84E-09	1.99E-09	1.98E-09	2.06E-09	1.99E-09
LI_wall	1.45E-09	1.63E-09	1.7E-09	1.71E-09	1.76E-09	1.73E-09	1.78E-09	1.82E-09	1.89E-09	1.96E-09	1.98E-09	2.04E-09	2.04E-09	2.04E-09
Liver	3.03E-10	3.42E-10	3.16E-10	3.44E-10	3.09E-10	3.45E-10	3.48E-10	4.13E-10	3.39E-10	4.06E-10	4.72E-10	4.15E-10	4.69E-10	4.6E-10
Lungs	5.51E-10	5.77E-10	6E-10	5.85E-10	6.03E-10	6.31E-10	6.63E-10	6.57E-10	6.67E-10	7.5E-10	7.02E-10	7.31E-10	7.73E-10	8.03E-10
Ovaries	1.4E-09	1.64E-09	1.45E-09	1.57E-09	1.57E-09	1.24E-09	1.28E-09	1.1E-09	1.76E-09	1.58E-09	1.9E-09	2.02E-09	2.19E-09	2.01E-09
Pancreas	4E-10	4.44E-10	4.7E-10	5.79E-10	4E-10	3.77E-10	6.72E-10	5.42E-10	4.61E-10	4.36E-10	5.24E-10	6.64E-10	5.26E-10	6.64E-10
Placenta	6.89E-10	8.77E-10	8.7E-10	9.27E-10	8.55E-10	8.41E-10	8.81E-10	8.68E-10	8.52E-10	9.24E-10	9.24E-10	9.23E-10	1E-09	1E-09
Remainder	7.46E-09	4.5E-08	5.29E-08	4.37E-08	3.34E-08	2.68E-08	2.41E-08	2.17E-08	2.05E-08	2.25E-08	2.11E-08	2E-08	2.16E-08	2.36E-08
SI_wall_and...	6.99E-10	7.29E-10	7.17E-10	7.38E-10	7.95E-10	7.53E-10	7.93E-10	7.95E-10	8.18E-10	8.53E-10	9.39E-10	9E-10	1.01E-09	9.79E-10
Skeleton	2.58E-08	1.48E-07	3.23E-07	3.97E-07	4.24E-07	3.84E-07	3.03E-07	2.44E-07	2.05E-07	1.52E-07	1.6E-07	1.96E-07	1.72E-07	1.46E-07
Skin	8.77E-08	2.16E-07	1.24E-07	1.13E-07	9.85E-08	8.57E-08	7.24E-08	6.25E-08	5.42E-08	5.04E-08	5.04E-08	4.9E-08	4.86E-08	4.86E-08
Spleen	5.13E-10	4.43E-10	4.52E-10	4.05E-10	3.49E-10	4.61E-10	4E-10	5.12E-10	4.28E-10	4.26E-10	7.02E-10	4.74E-10	5.22E-10	5.01E-10
Stomach_wall	2.67E-10	3.28E-10	2.4E-10	2.88E-10	2.91E-10	3.06E-10	3.5E-10	3.47E-10	3.32E-10	4.39E-10	3.99E-10	4.33E-10	4.91E-10	6.01E-10
Thymus	1.46E-10	2.3E-10	2.32E-10	1.87E-10	2.15E-10	2.68E-10	2.2E-10	1.25E-10	1.24E-10	1.09E-10	7.54E-11	1.81E-10	2.23E-10	2.51E-10

Ding A, Gao Y, Liu H , Caracappa PF, Long DJ, Bolch WE, Liu Bob, Xu XG. VirtualDose: A New CT Dose Reporting Software for Adult and Pediatric Patients. *Phys. Med. Biol.* 60(14):5601-5625 (2015).

# CT Dose Reporting Software

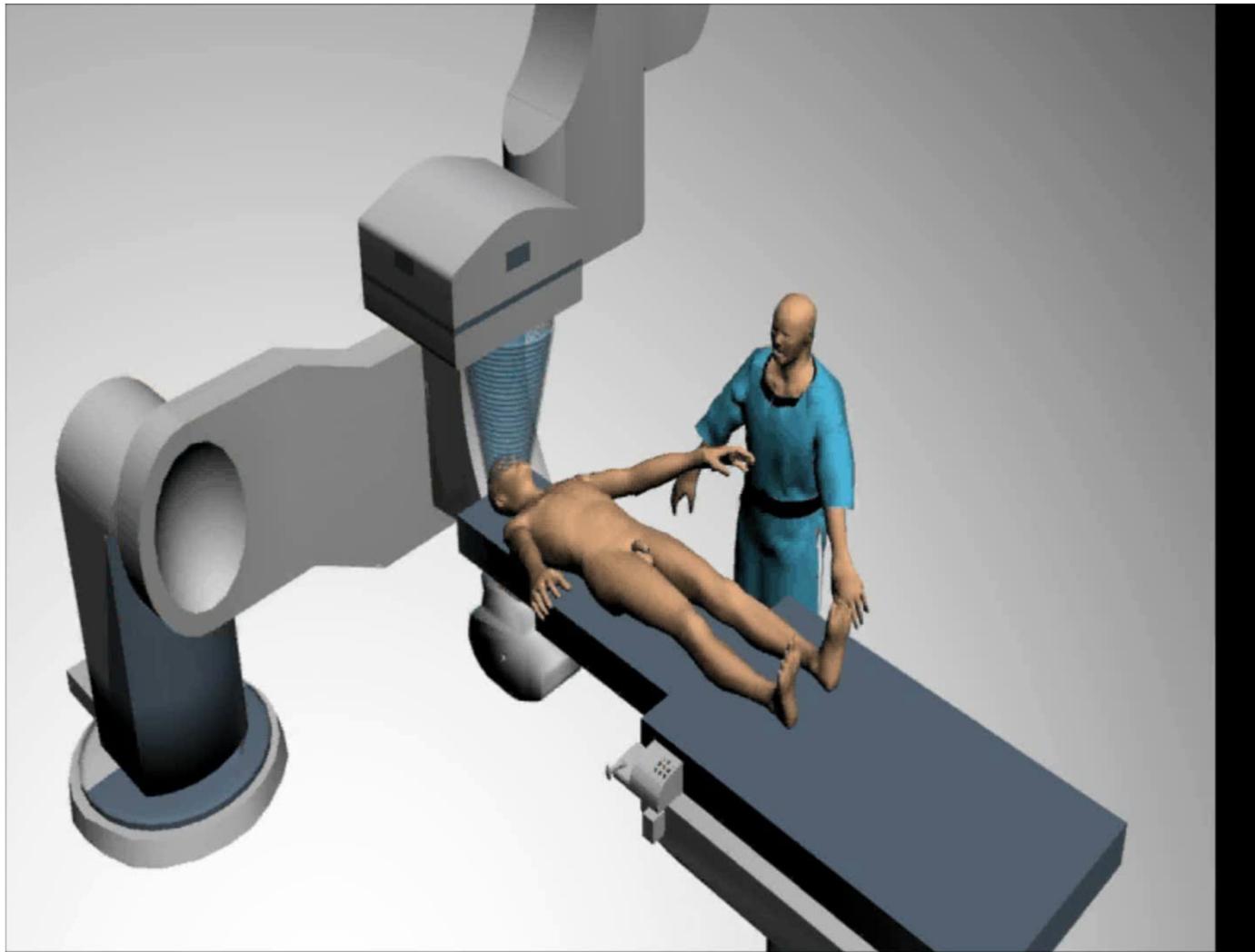
**VirtualDose™ CT**  
A product of Virtual Phantoms, Inc.

<http://www.virtual-dose.com>



In 2021, over 32-million dose calculation requests from more than 325 sites worldwide were processed

# VirtualDose for Interventional Radiology

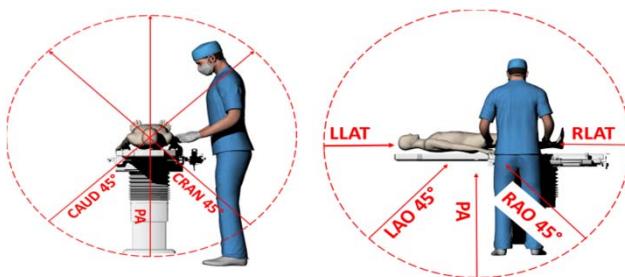
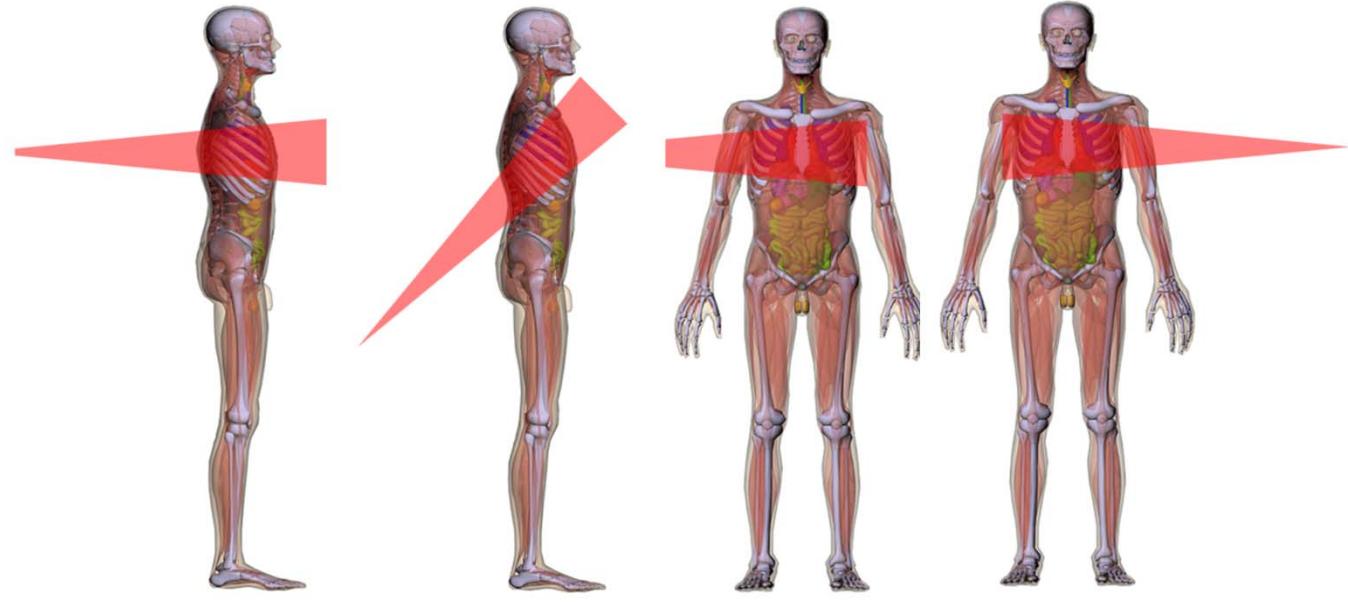


# VirtualDose-IR

## Dose from fluoroscopically guided intervention (FGI) - Patients

- Simulated beam directions:

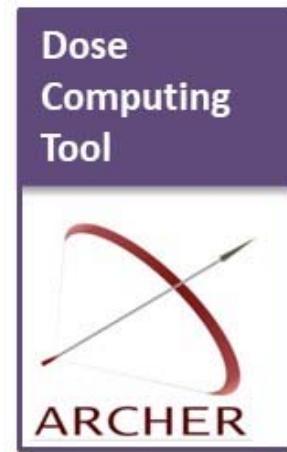
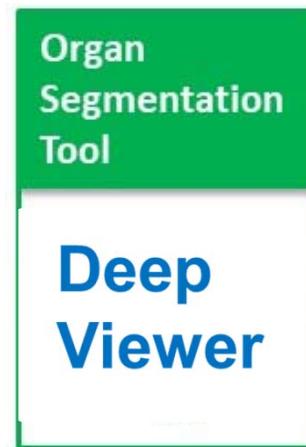
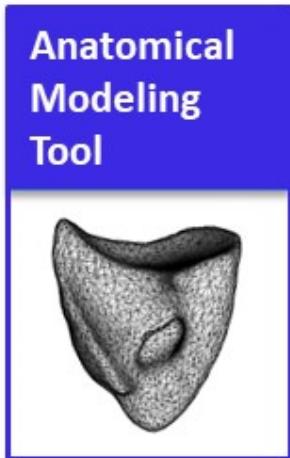
- Posterior-Anterior, Crani 45°
- Lateral: left & right
- Oblique: 45° left & right



# Comments

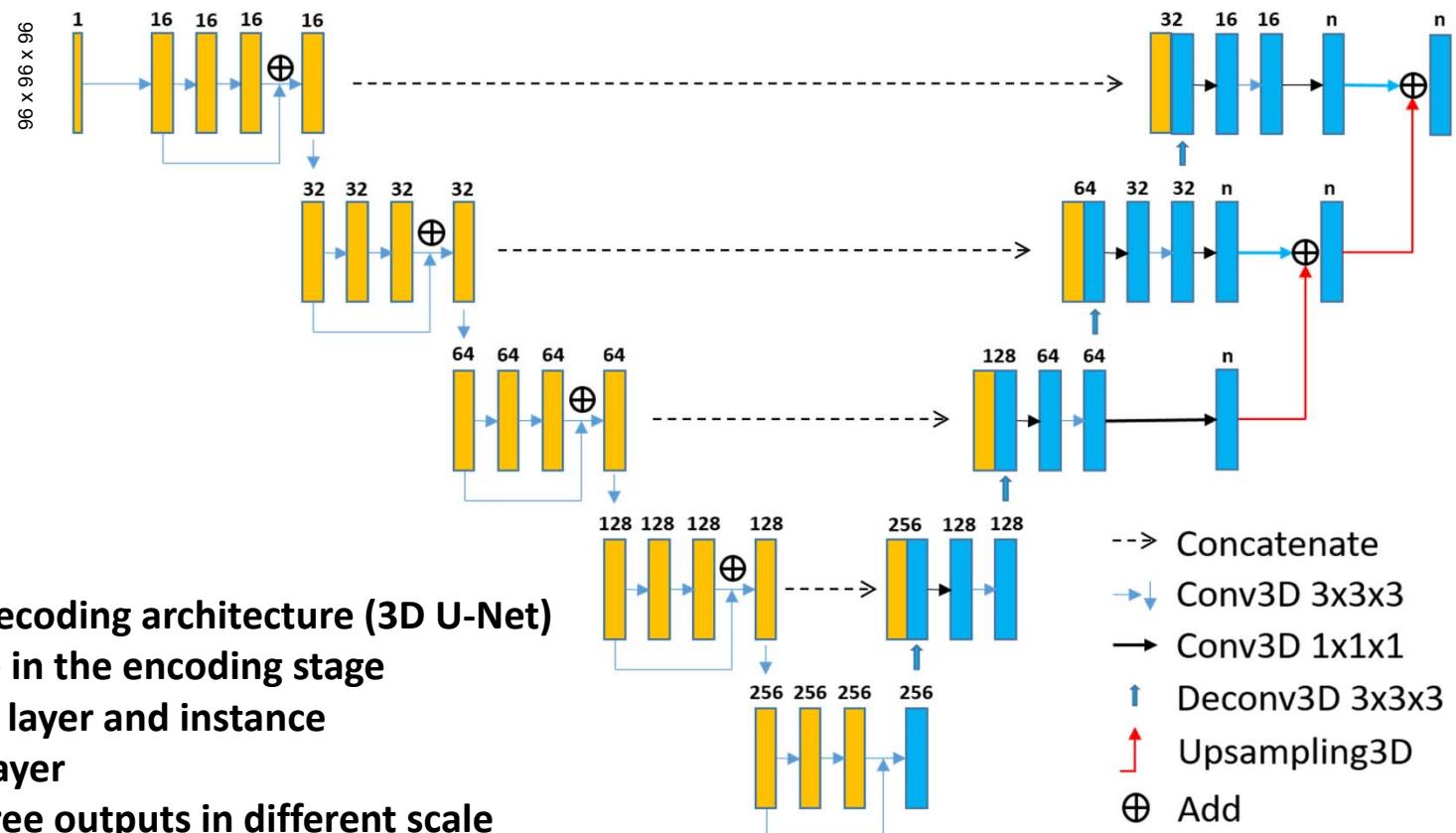
- For decades, in imaging procedures, all “organ doses” were assessed using ICRP “Reference Man” approach involving “phantom libraries”
- “Patient-specific organ doses” would require two newly available tools:
  1. automatic organ segmentation
  2. “near real-time” Monte Carlo simulations

# Outline of the presentation



# Auto Multi-organ Segmentation Made Possible Through Deep-Learning based Segmentation Methods

## Network architecture [1] :



### Features:

- Encoding and decoding architecture (3D U-Net)
- Res-net module in the encoding stage
- Spatial dropout layer and instance normalization layer
- Merging the three outputs in different scale and getting the final segmentation results

[1] O. Ronneberger, P. Fischer, and T. Brox, "U-net: Convolutional networks for biomedical image segmentation," in <sup>24</sup> *International Conference on Medical image computing and computer-assisted intervention*, 2015: Springer, pp. 234-241.

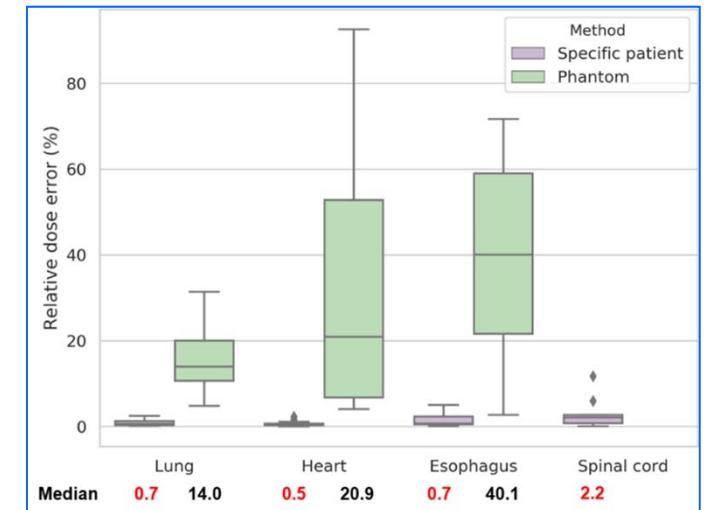
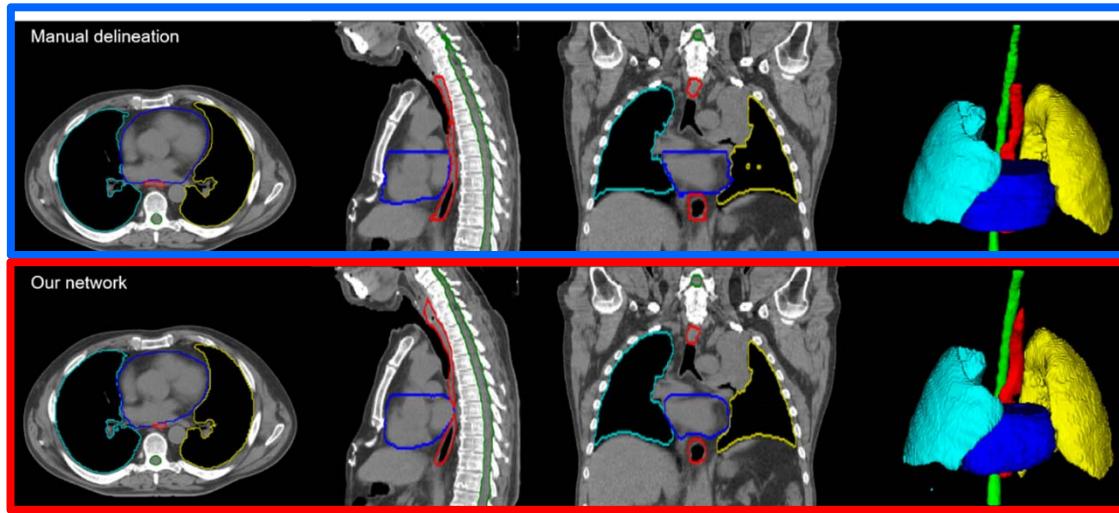
# Dataset 1: Lung CT Segmentation Challenge 2017 (LCTSC) [2]

[2] Yang Jinzhong, et al. Data from Lung CT Segmentation Challenge (2017). The Cancer Imaging Archive.

- **Training and validation set:** 48 patients
- **Testing set:** 12 patients
- **Loss function:** weighted dice similarity
- **5 segmented organs:** left lung (yellow), right lung (cyan), heart (blue), spinal cord (green), and esophagus (red)

Dice similarity coefficient (mean  $\pm$  standard deviation)

Methods	Left lung	Right lung	Heart	Esophagus	Spinal cord
Interrater variability	0.96 $\pm$ 0.02	0.96 $\pm$ 0.02	0.93 $\pm$ 0.02	0.82 $\pm$ 0.04	0.86 $\pm$ 0.04
1	0.97 $\pm$ 0.02	0.97 $\pm$ 0.02	0.93 $\pm$ 0.02	0.72 $\pm$ 0.10	0.88 $\pm$ 0.04
2	0.98 $\pm$ 0.01	0.97 $\pm$ 0.02	0.92 $\pm$ 0.02	0.64 $\pm$ 0.20	0.89 $\pm$ 0.04
3	0.98 $\pm$ 0.02	0.97 $\pm$ 0.02	0.91 $\pm$ 0.02	0.71 $\pm$ 0.12	0.87 $\pm$ 0.11
4	0.97 $\pm$ 0.01	0.97 $\pm$ 0.02	0.90 $\pm$ 0.03	0.64 $\pm$ 0.11	0.88 $\pm$ 0.05
5	0.96 $\pm$ 0.03	0.95 $\pm$ 0.05	0.92 $\pm$ 0.02	0.61 $\pm$ 0.11	0.85 $\pm$ 0.04
6	0.96 $\pm$ 0.01	0.96 $\pm$ 0.02	0.90 $\pm$ 0.02	0.58 $\pm$ 0.11	0.87 $\pm$ 0.02
7	0.95 $\pm$ 0.03	0.96 $\pm$ 0.02	0.85 $\pm$ 0.04	0.55 $\pm$ 0.20	0.83 $\pm$ 0.08
Ours	<b>0.96<math>\pm</math>0.01</b>	<b>0.96<math>\pm</math>0.02</b>	<b>0.93<math>\pm</math>0.02</b>	<b>0.73<math>\pm</math>0.10</b>	<b>0.88<math>\pm</math>0.04</b>



Peng Z, Fang X, Yan P, Shan H, Liu T, Pei X, Wang G, Liu B, Kalra MK, Xu XG. A Method of Rapid Quantification of Patient-specific Organ Doses for CT Using Deep-learning-based Multi-organ Segmentation and GPU-accelerated Monte Carlo Dose Computing. *Med Phys*, (47)6: 2526–2536 (2020) 25

## Dataset 2: The Cancer Image Archive (TCIA) Pancreas-CT [3]

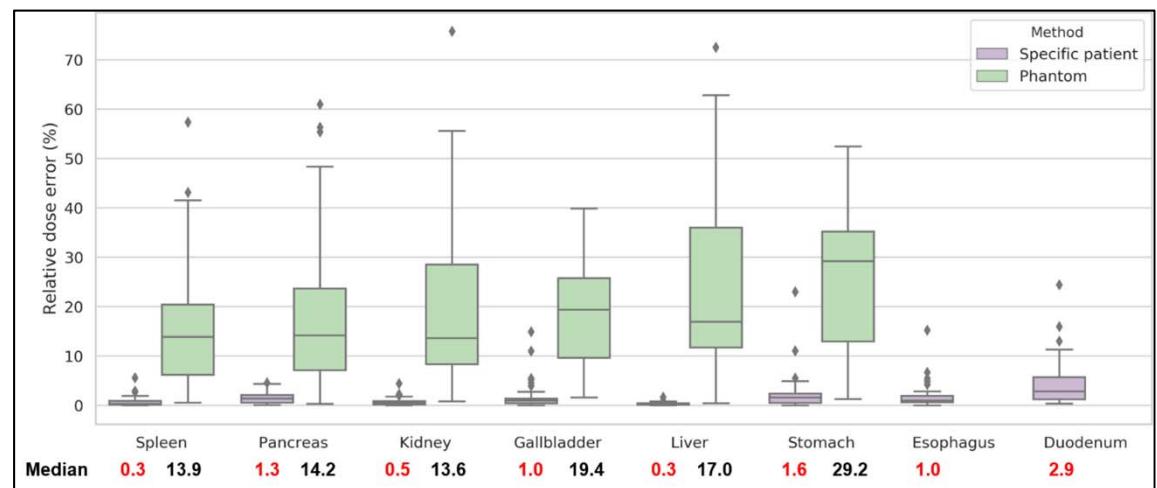
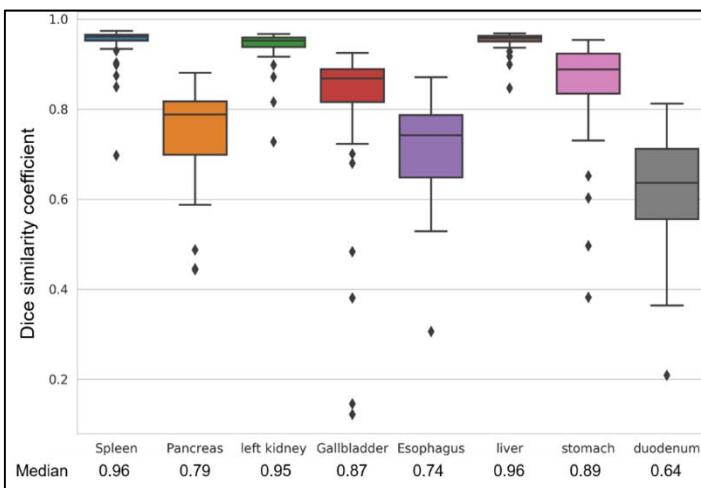
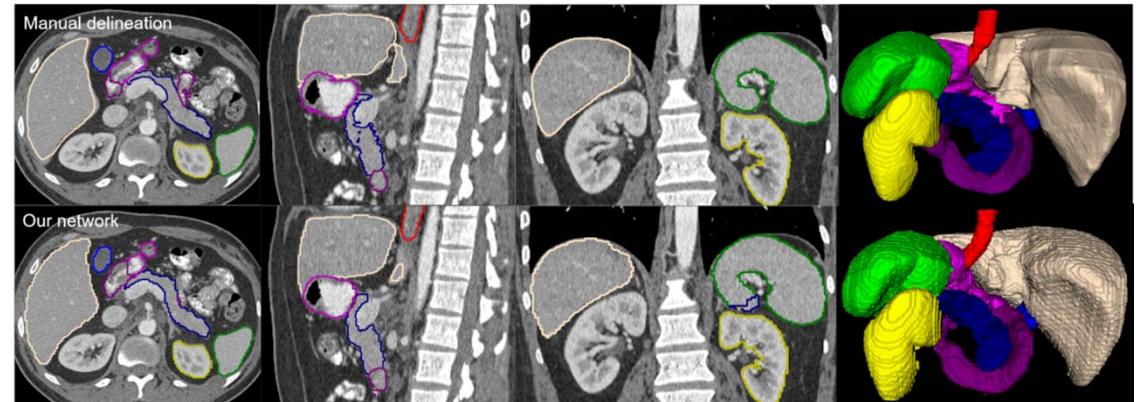
[3] Gibson E, et al. Automatic multi-organ segmentation on abdominal CT with dense v-networks. IEEE Transactions on Medical Imaging, 2018.

**Total patients:** 43

**5 cross-validation:** 8, 8, 9, 9, 9

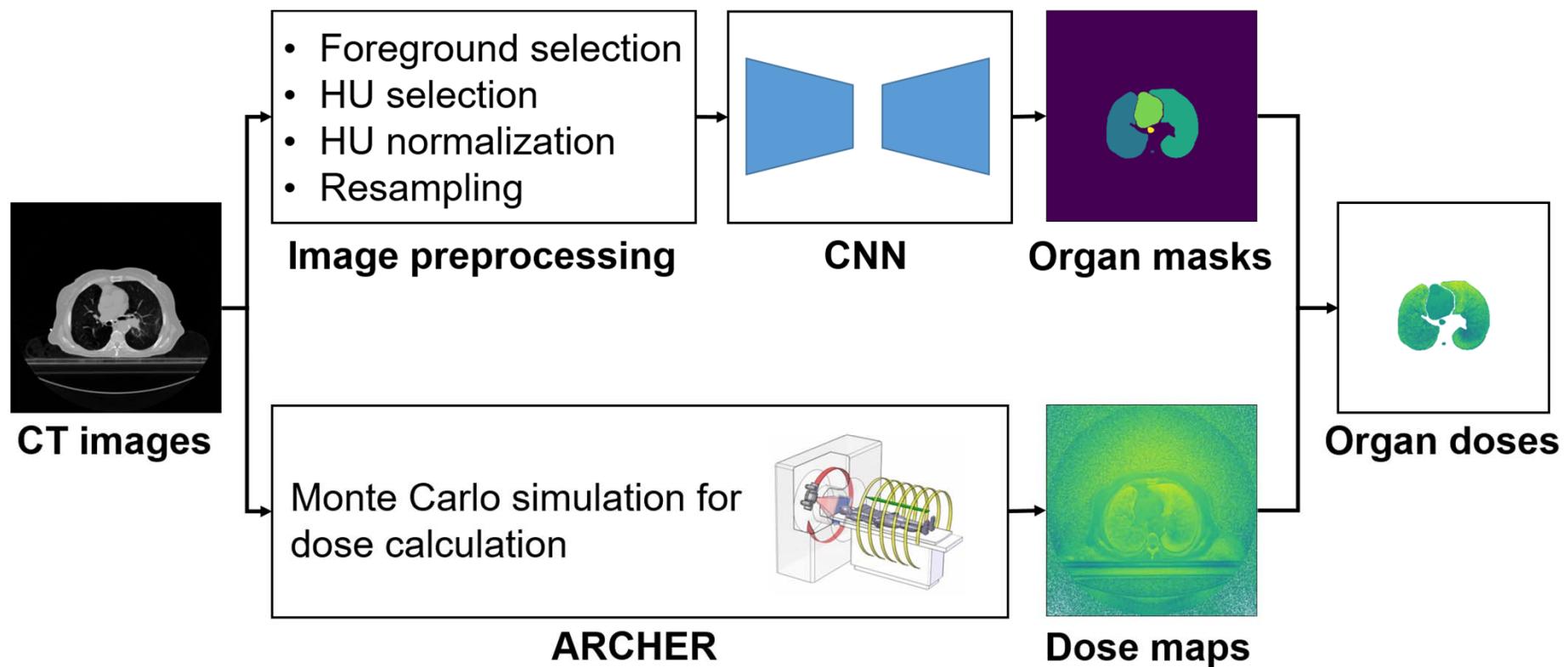
**Loss function:** weighted dice similarity coefficient

**8 segmented organs:** spleen (green), pancreas (navy), left kidney (yellow), gallbladder (blue), esophagus (red), liver (bisque), stomach (magenta), and duodenum (purple)



Peng Z, Fang X, Yan P, Shan H, Liu T, Pei X, Wang G, Liu B, Kalra MK, Xu XG. A Method of Rapid Quantification of Patient-specific Organ Doses for CT Using Deep-learning-based Multi-organ Segmentation and GPU-accelerated Monte Carlo Dose Computing. *Med Phys*, (47)6: 2526–2536 (2020) 26

# Method: Organ doses for patients undergoing CT exams



Peng Z, Fang X, Yan P, Shan H, Liu T, Pei X, Wang G, Liu B, Kalra MK, Xu XG. A Method of Rapid Quantification of Patient-specific Organ Doses for CT Using Deep-learning-based Multi-organ Segmentation and GPU-accelerated Monte Carlo Dose Computing. *Med Phys*, (47)6: 2526–2536 (2020)

# Segmentation Testing for the DeepViewer Tool

- Can perform more than 40 organs and tissues
- Whole body in 3-5 minutes

Organs	Dice	Organs	Dice	Organs	Dice	Organs	Dice
Skin	0.99	Brainstem	0.94	Stomach	0.88	Larynx	0.85
Brain	0.98	Femoral Head	0.93	Temporal Lobe	0.88	Rectum	0.84
Lung	0.98	Pelvis	0.93	Parotid	0.86	Oral Cavity	0.83
Liver	0.97	Heart	0.93	Esophagus	0.86	Breast	0.81
Spleen	0.95	Mandible	0.93	Pancreas	0.86	Optic Nerve	0.78
Bladder	0.95	Trachea	0.92	Lens	0.85	Pituitary	0.76
Cerebellum	0.95	Spinal Cord	0.92	Thyroid	0.85	Optic Chiasm	0.75
Kidney	0.95	Eye Ball	0.9	Bowel	0.85	Cochlea	0.75

# *DeepViewer*, for Automatic Target and OAR Contouring in Radiotherapy

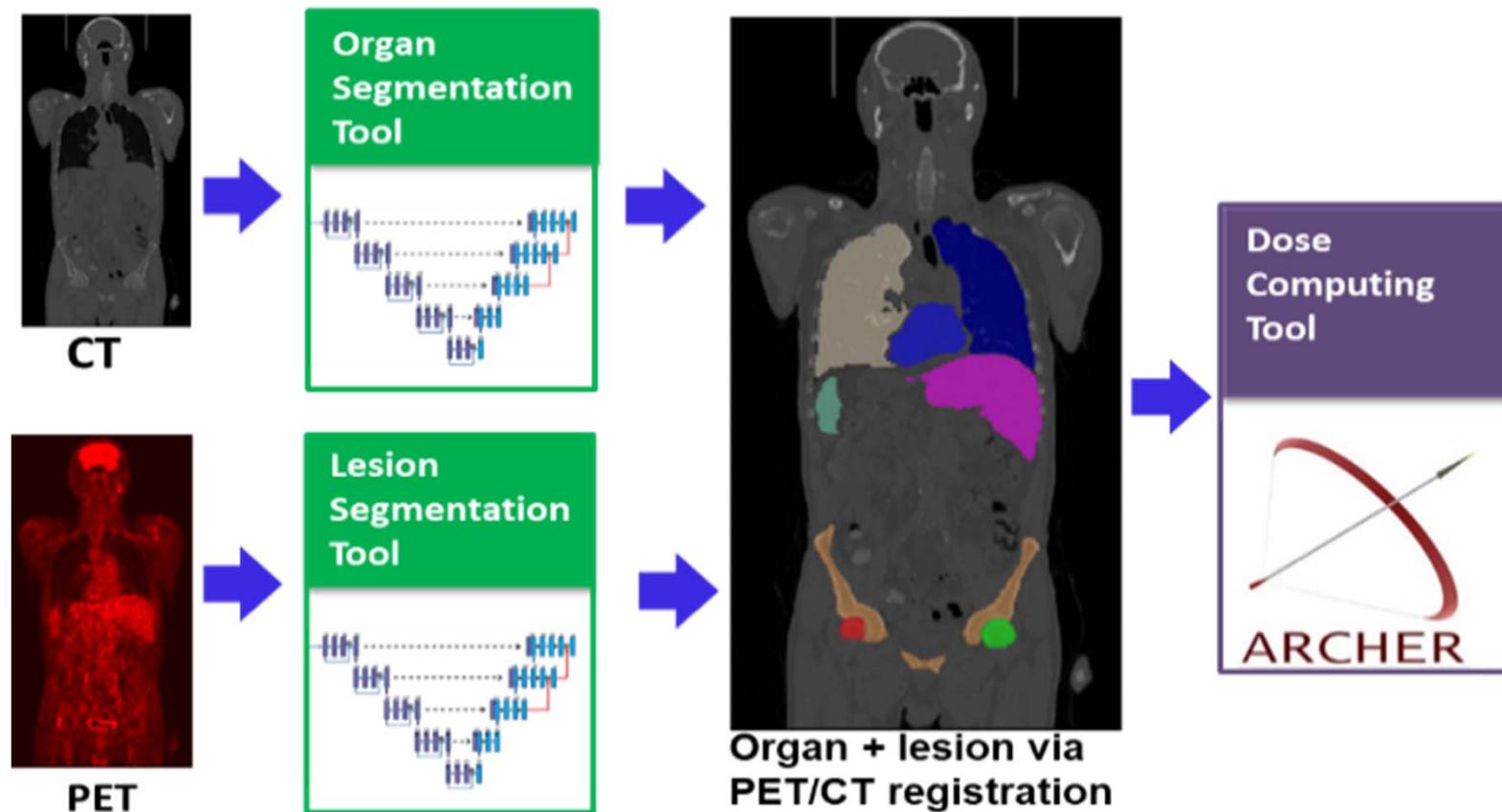
<http://www.wisdom-tech.online/>



3D visualization

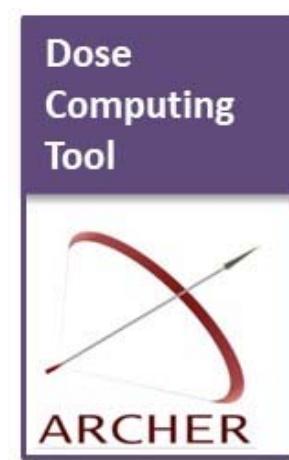
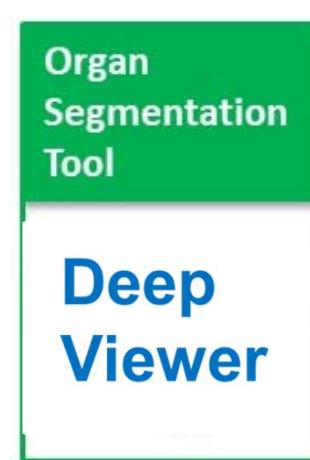
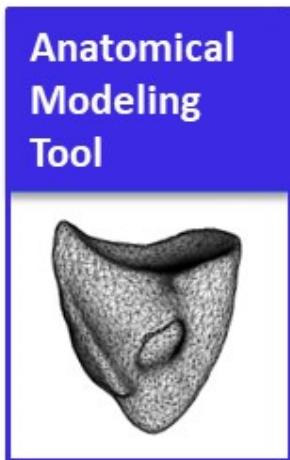
- Head/neck, chest, abdominal treatment plans
- 40+ OARs
- 3-5 minutes to complete
- Acceptance rate 95% or better

# Archer-NM, Organ doses for patients undergoing PET/CT exams (Nuclear Medicine)



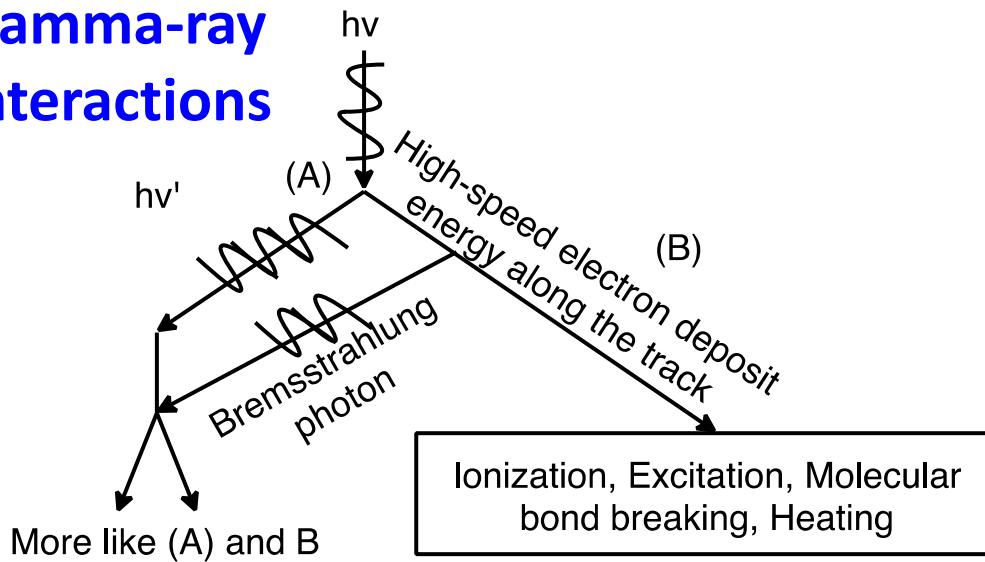
Peng Z, Li Y, Xu Y, Lu Y, Li M, Pei X, Xu XG. Development of a GPU-accelerated Monte Carlo dose calculation module for nuclear medicine, ARCHER-NM: Application for PET/CT imaging procedure. *Phys Med Biol*, 67, 06NT02 (2022)

# Outline of the presentation



# Introduction: Dose Calculation and Monte Carlo Methods

## Gamma-ray interactions



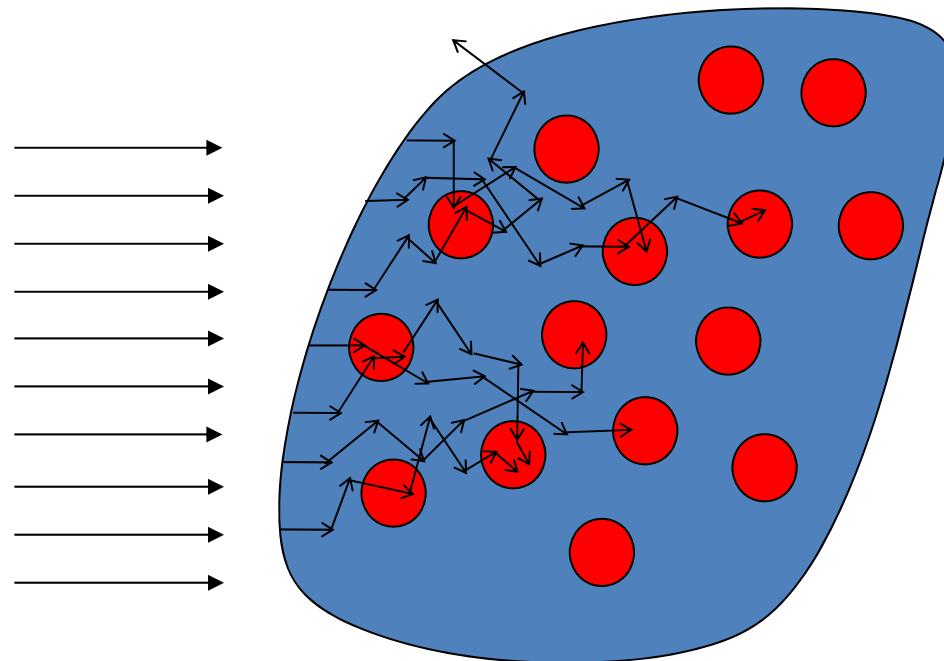
- Dose: energy imparted to matter via ionization and excitation per unit mass
- Deterministic methods subject to non-trivial errors by approximating electron transport in water
- Monte Carlo particle transport is the ***gold standard***
- But lengthy computation time is the main bottleneck of MC applications for clinical settings

# Monte Carlo Methods Ideal for Radiation Transport Simulations, but Used to be Extremely Slow

Boltzmann Transport Equation can be solved by MC methods

$$\begin{aligned} & \frac{1}{v} \frac{\partial}{\partial t} \psi(\vec{r}, \hat{\Omega}, E, t) + \hat{\Omega} \cdot \vec{\nabla} \psi(\vec{r}, \hat{\Omega}, E, t) + \Sigma_t(\vec{r}, E) \psi(\vec{r}, \hat{\Omega}, E, t) \\ &= \int dE' \int d\Omega' \Sigma_s(\vec{r}, E' \rightarrow E, \hat{\Omega}' \cdot \hat{\Omega}) \psi(\vec{r}, \hat{\Omega}', E', t) + S(\vec{r}, \hat{\Omega}, E, t) \end{aligned}$$

A Statistical Experiment



# The ARCHER Project

**-towards “real-time” Monte Carlo**



ARCHER (Accelerated RComputations in Heterogeneous EnviRonments)

- Initiated in 2009
- Goals:

1. To understand heterogeneous computing architecture and programming models
2. To test code performance on GPUs and MICs
3. To develop functional Monte Carlo codes that take full advantage of CPUs, GPUs and MICs

# Solution: ARCHER – A GPU-based Monte Carlo Code

## POINT/COUNTERPOINT

Suggestions for topics suitable for these Point/Counterpoint debates should be addressed to Colin G. Orton, Professor Emeritus, Wayne State University, Detroit; [orton@comcast.net](mailto:orton@comcast.net). Persons participating in Point/Counterpoint discussions are selected for their knowledge and communicative skill. Their positions for or against a proposition may or may not reflect their personal opinions or the positions of their employers.

GPU technology is the hone for near real-time Monte Carlo dose calculations

Xun Jia, Ph.D.  
Department of Radiation Oncology  
(Tel: 214-648-3224; E-mail: [xia@utsouthwestern.edu](mailto:xia@utsouthwestern.edu))

X. George Xu, Ph.D.  
Nuclear Engineering Program  
(Tel: 518-276-4014; E-mail: [xug@rpi.edu](mailto:xug@rpi.edu))

Colin G. Orton, Ph.D., Moderator

(Received 15 November 2014; accepted for publication 20 November 2014; published 11 March 2015)

[<http://dx.doi.org/10.1118/1.4903901>]

## Real-time MC?

### OVERVIEW

Monte Carlo (MC) dose calculations are recognized as being the most accurate modality for radiotherapy treatment planning but, because of the excessive computational time required, they cannot presently be used for near real-time dose calculations. Currently, the most common way to accelerate MC dose calculations is to use clusters of central processing units (CPUs), but some believe that the future of near real-time MC dose calculations lies not with clusters of CPUs but with the use of graphics processing unit (GPU) technology. This is the claim debated in this month's Point/Counterpoint.



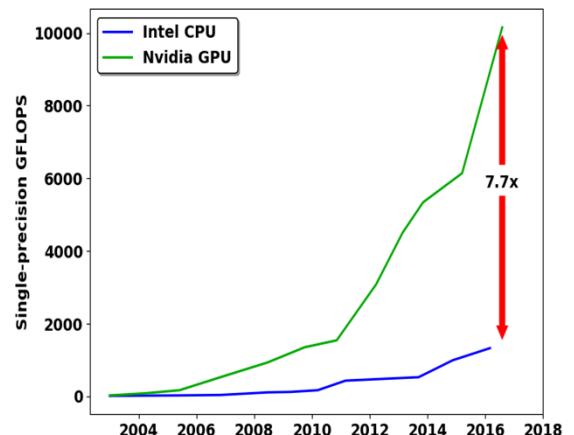
Arguing against the Proposition is X. George Xu, Ph.D. Dr. Xu obtained his Ph.D. in Nuclear Engineering from Texas A&M University, College Station, TX and, for the past 20 years, he has been on the faculty of Rensselaer Polytechnic Institute, Troy, NY, where he currently holds the Edward E. Hood Endowed Chair of Engineering. Dr. Xu's research has centered around applications of

Monte Carlo methods to problems in radiation protection, imaging, and radiation therapy. He has been continuously funded by the NIH over the past ten years, including an R01 grant to develop a new Monte Carlo code, ARCHER, for heterogeneous computing involving GPUs and coprocessors. He is the author of more than 150 journal papers and book chapters, and 270 conference abstracts. Dr. Xu is a Fellow of the American Association of Physicists in Medicine, the Health Physics Society, and the American Nuclear Society. In 2014, he was re-elected to a 6-year term as a council member of the National Council on Radiation Protection and Measurements.

Arguing for the Proposition is Xun Jia, Ph.D. Dr. Jia received his Masters degree in Applied Mathematics and Ph.D. degree in Physics, both from UCLA. He is currently an Assistant Professor in the Department of Radiation Oncology, University of Texas Southwestern Medical Center. Dr. Jia's research focuses on GPU-based high-performance computing for medical physics and medical

imaging. He has developed several Monte Carlo packages to improve efficiency for photon, electron, and proton transport. Dr. Jia's research has been supported by government and industrial grants and he has published 60 peer-reviewed papers. He is currently a section editor of the *Journal of Applied Clinical Medical Physics*.

## GPU is more powerful than CPU



### Application

- CT imaging
- Radiotherapy
- Shielding design
- Reactor analysis

### ARCHER



### Hardware

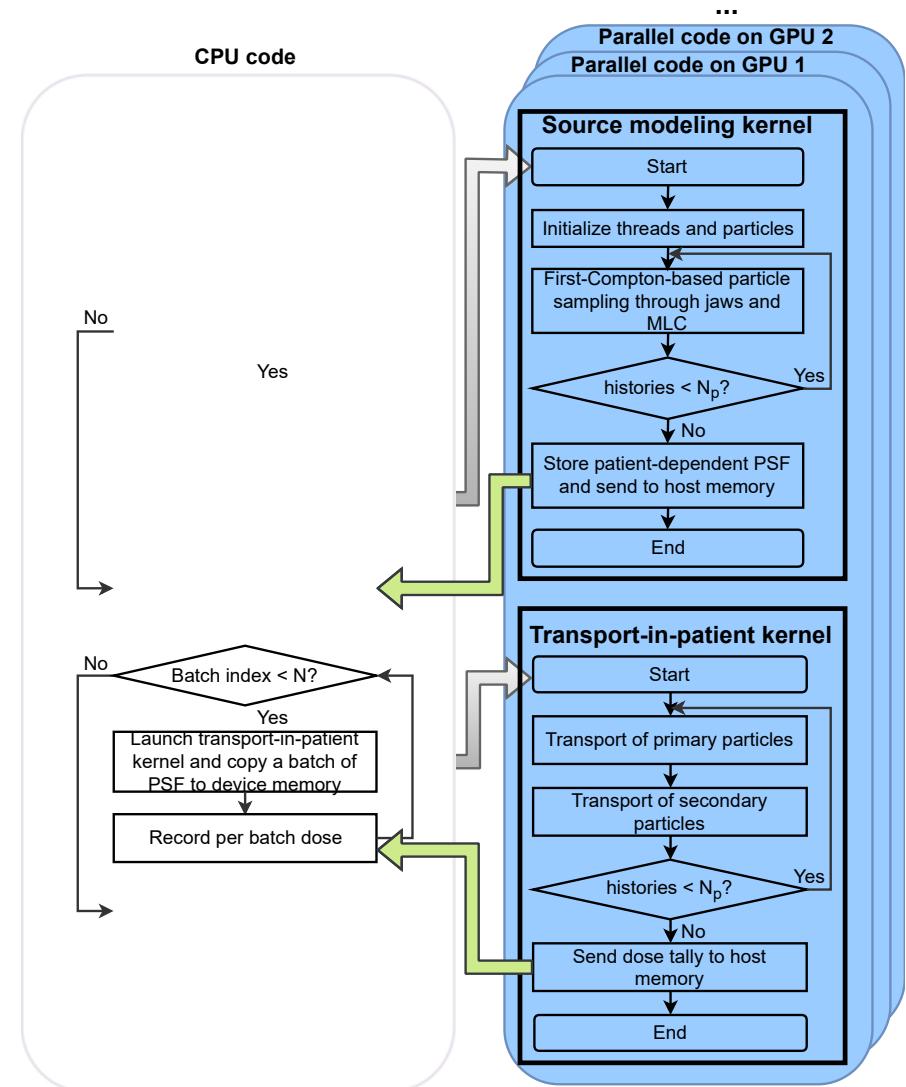
- Intel/AMD multi-core CPU
- NVIDIA Fermi/Kepler GPU
- AMD GCN GPU
- Intel MIC coprocessor

### Software

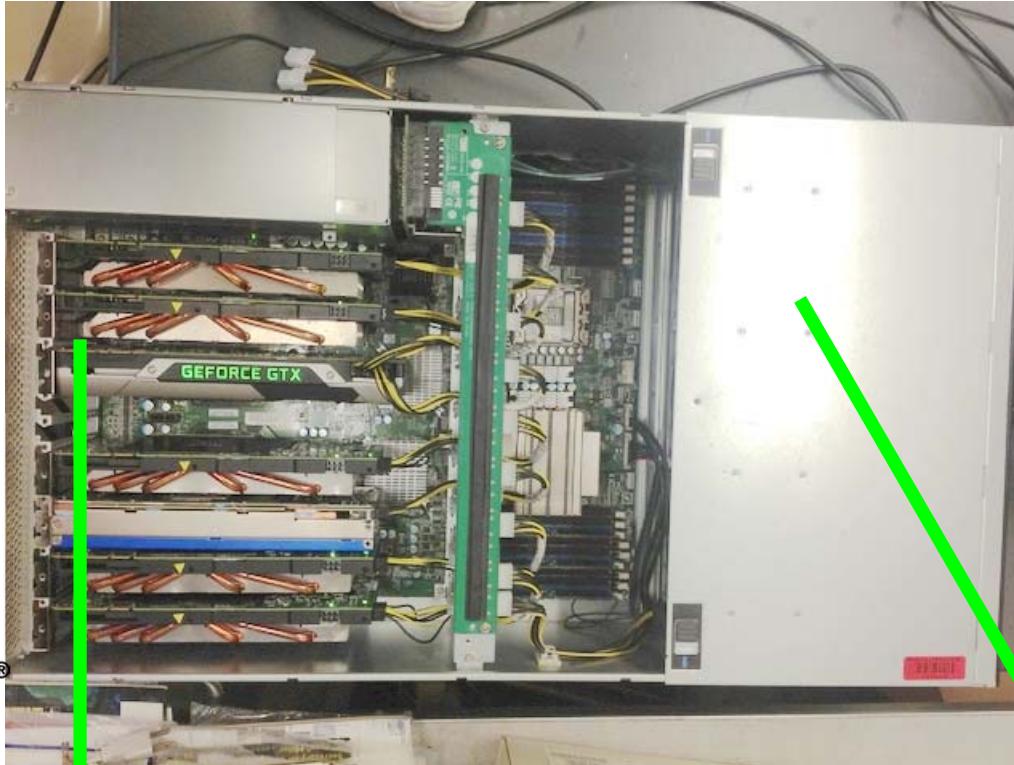
- MPI
- OpenMP
- Pthreads
- CUDA
- OpenACC
- OpenCL
- Cilk

# ARCHER Software Development

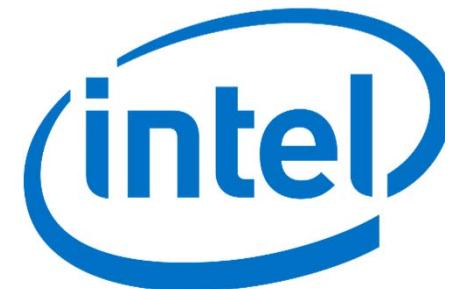
- Two code variants
  - ✓ ARCHER<sub>CPU</sub>: multi-threaded parallelization
  - ✓ ARCHER<sub>GPU</sub>: multi-GPUs parallelization
- GPU/CUDA code optimization challenging
- 8 Ph.D. dissertations



# Desk-Top CPU/GPU Hardware @ \$10k



Multiple Nvidia GPU cards

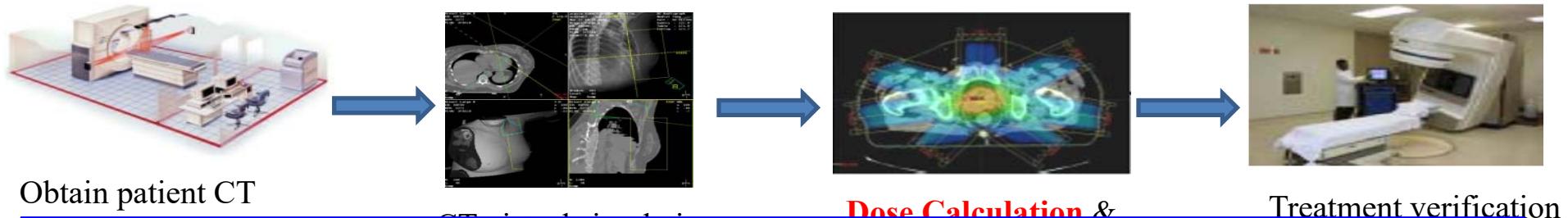


Intel CPU



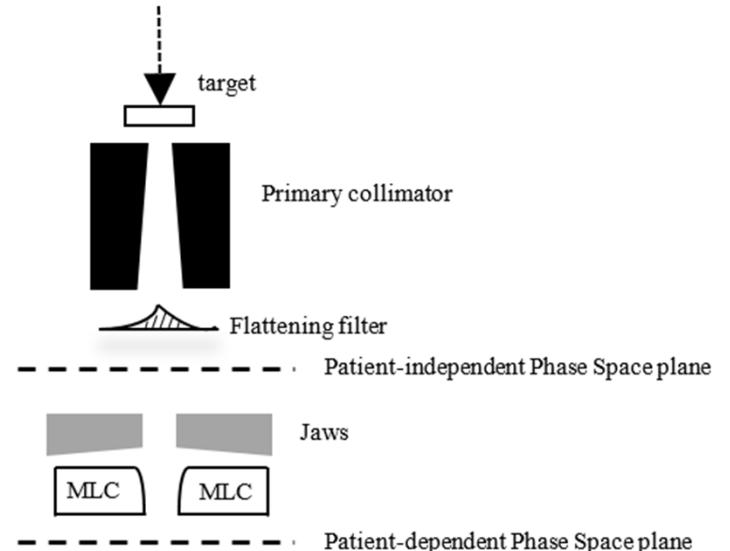
[1]

# Introduction: External-Beam Radiotherapy



- ***Dose verification systems provide an opportunity for complementing the measurement-based approach***
  - ***Far less time consuming***
  - ***3D and patient-specific***
- Report of AAPM Task Group 219 on independent calculation-based dose/MU verification for IMRT (2021)

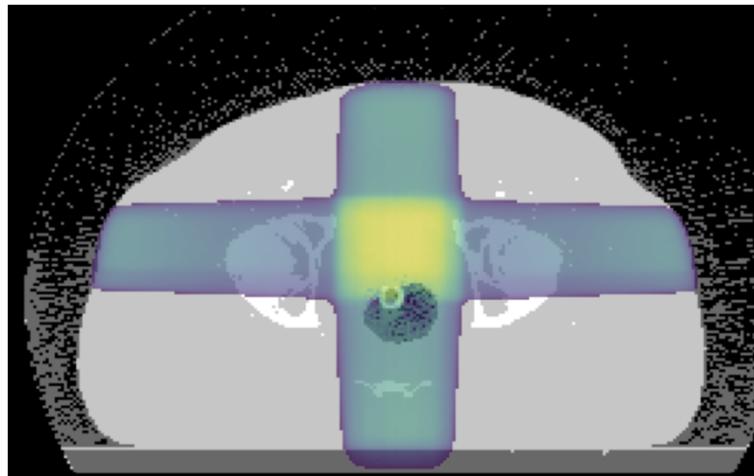
# Virtual Source Modeling



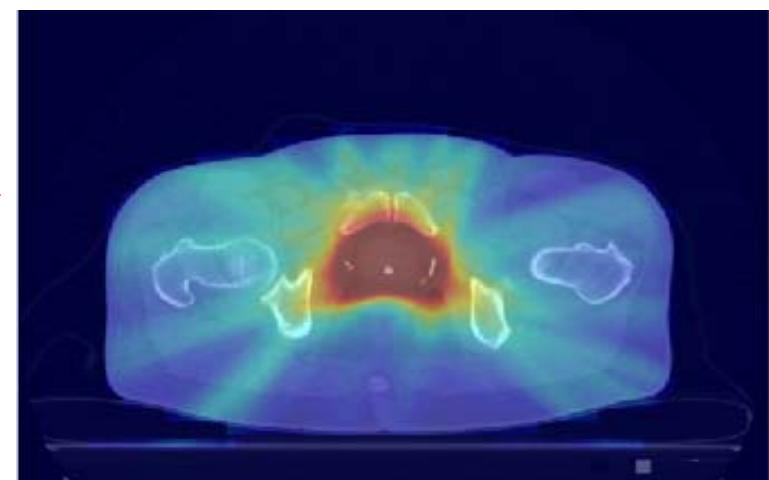
- Phase space source
  - Phase space particles: a collection of representative pseudo-particles emerging from a radiotherapy source along with their properties ( $x, y, z, u, v, w, E, wt$ )
  - can be categorized as
    - Patient-independent phase space: could be utilized repeatedly
    - Patient-dependent phase space: specific to each plan and beam

# Results: ARCHER for External-Beam Therapy (Tomotherapy, IMRT and VMAT) [1-5]

Step and Shoot IMRT



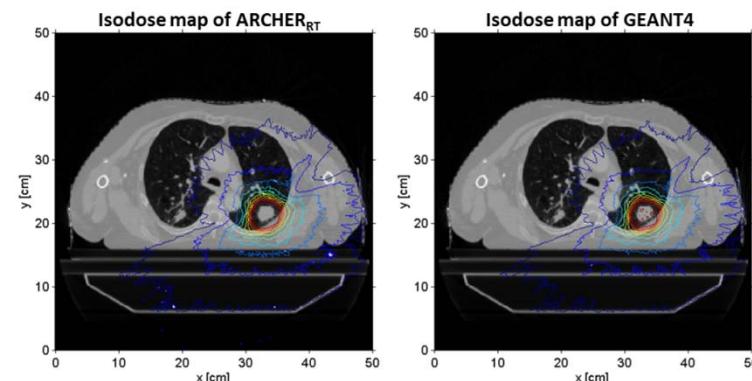
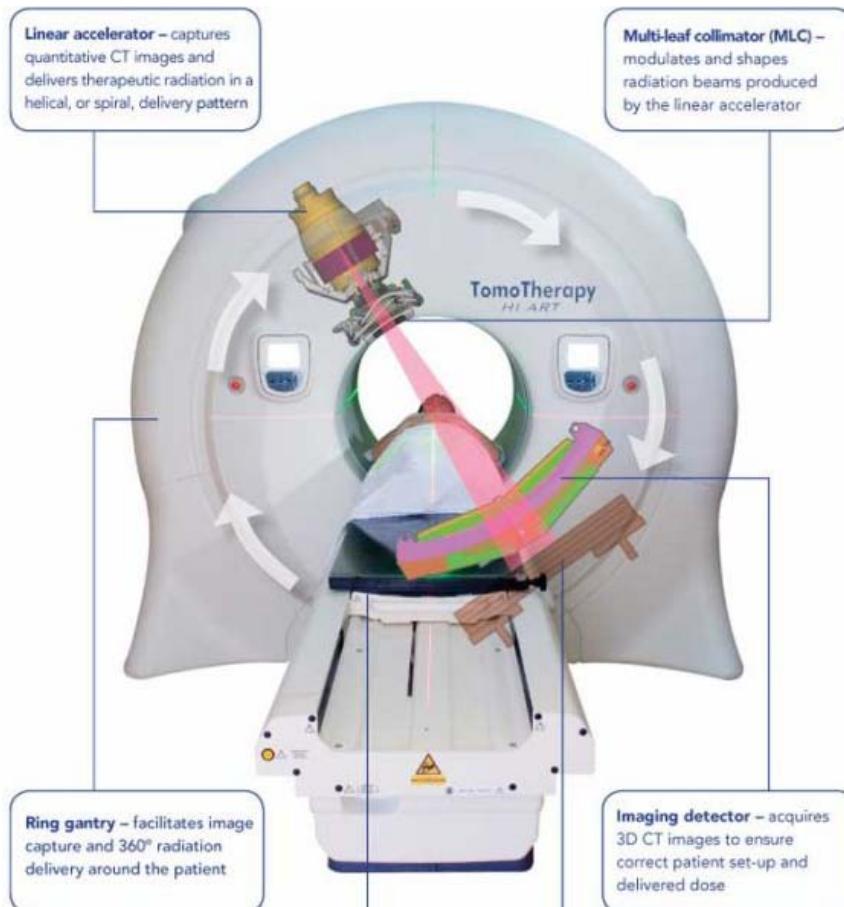
VMAT



- (1) Liu T. Development of ARCHER - a parallel Monte Carlo radiation transport code -- for X-ray CT dose calculations using GPU and coprocessor technologies, Ph.D. Dissertation, Rensselaer Polytechnic Institute (2014).
- (2) Su L. Development and Application of a GPU-Based Fast Electron-Photon Coupled Monte Carlo Code for Radiation Therapy, Ph.D. Dissertation, Rensselaer Polytechnic Institute (2014).
- (3) Lin H. GPU-based Monte Carlo Source Modeling and Simulation for Radiation Therapy involving Varian Truebeam Linac. , Ph.D. Dissertation, Rensselaer Polytechnic Institute (2018).
- (4) Su L, Yang YM, Bednarz B, Sterpin E, Du X, Liu T, Ji W, Xu XG. ARCHER<sub>RT</sub> — A Photon-Electron Coupled Monte Carlo Dose Computing Engine for GPU: Software Development of and Application to Helical Tomotherapy. *Med Phys.* 41:071709 (2014).
- (5) Adam DP\*, Liu T, Caracappa PF, Bednarz BP, Xu XG. New capabilities of the Monte Carlo dose engine ARCHER-RT: clinical validation of the Varian TrueBeam machine for VMAT external beam radiotherapy. *Med Phys* 2537-2549 (2020).

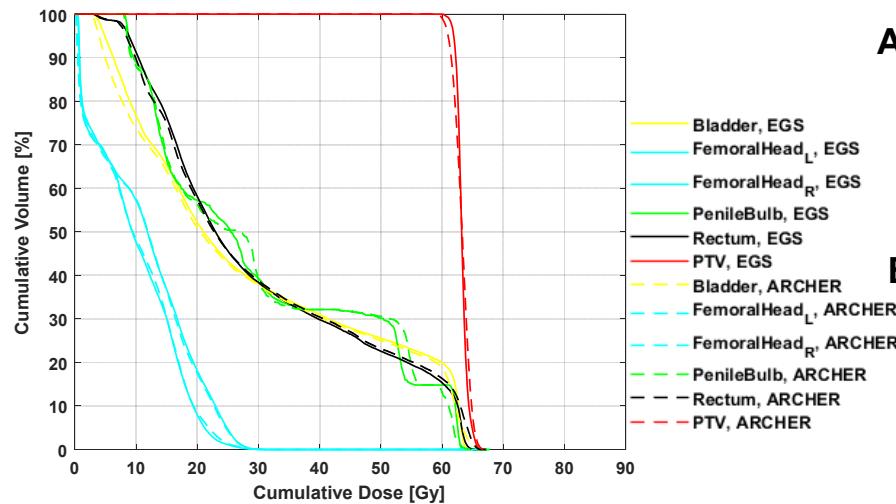
# Results: Tomotherapy

Su L, Yang YM, Bednarz, Edmond Sterpin, Du X, Liu T, Ji W, Xu XG. ARCHER<sub>RT</sub> — A Photon-Electron Coupled Monte Carlo Dose Computing Engine for GPU: Software Development of and Application to Helical Tomotherapy. Med Phys (2014).



***GPU computing time ~ 1 sec,***  
***GEANT4 ~ 500 CPU hours***

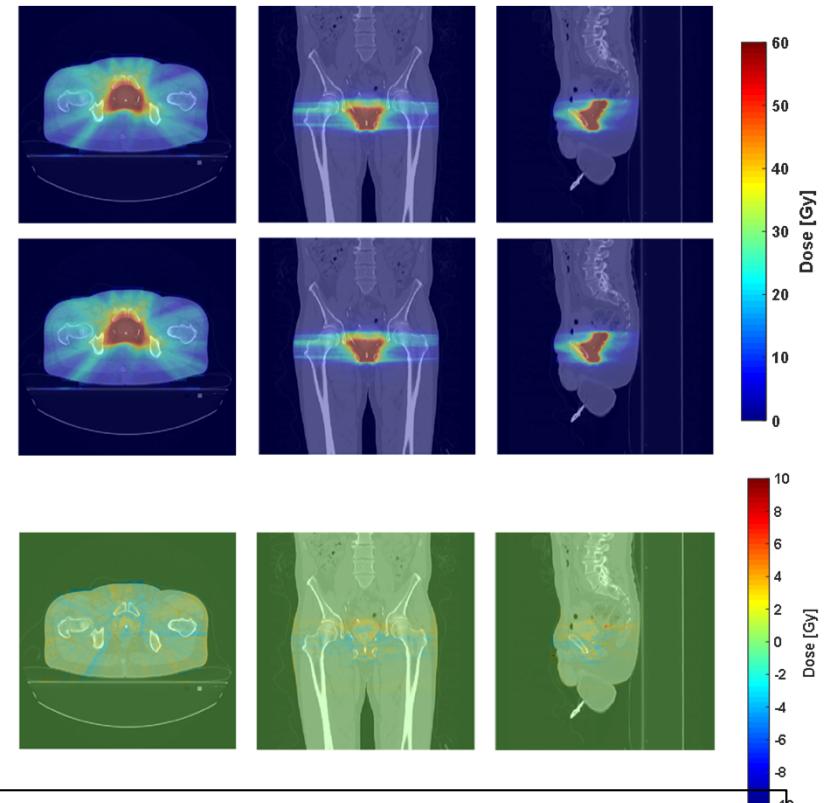
# Results: Clinical VMAT (prostate case)



ARCHER

EGSnrc

- Gamma pass rates:
  - 3%/3mm: 99.71%



<input type="checkbox"/> ARCHER (NVIDIA 1080Ti GPU) →	48 seconds
<input type="checkbox"/> Patient transport	
<input type="checkbox"/> Source modeling (linac)	
<input type="checkbox"/> EGSnrc (120-node cluster Intel Xeon E5335 CPUs) →	24 hours

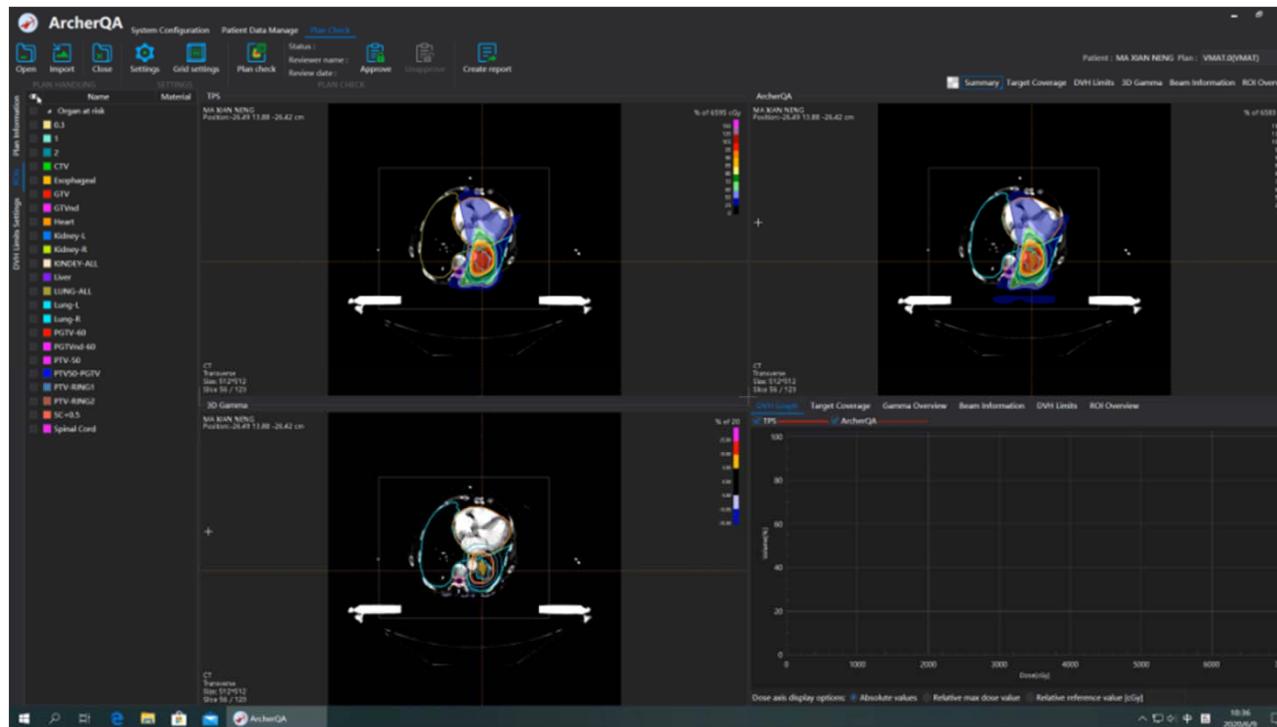
Adam DP\*, Liu T, Caracappa PF, Bednarz BP, Xu XG. New capabilities of the Monte Carlo dose engine ARCHER-RT: clinical validation of the Varian TrueBeam machine for VMAT external beam radiotherapy. Med Phys 2537-2549 (2020).

# **ArcherQA, for 3D Independent Dose-check (verification of TPS results)**



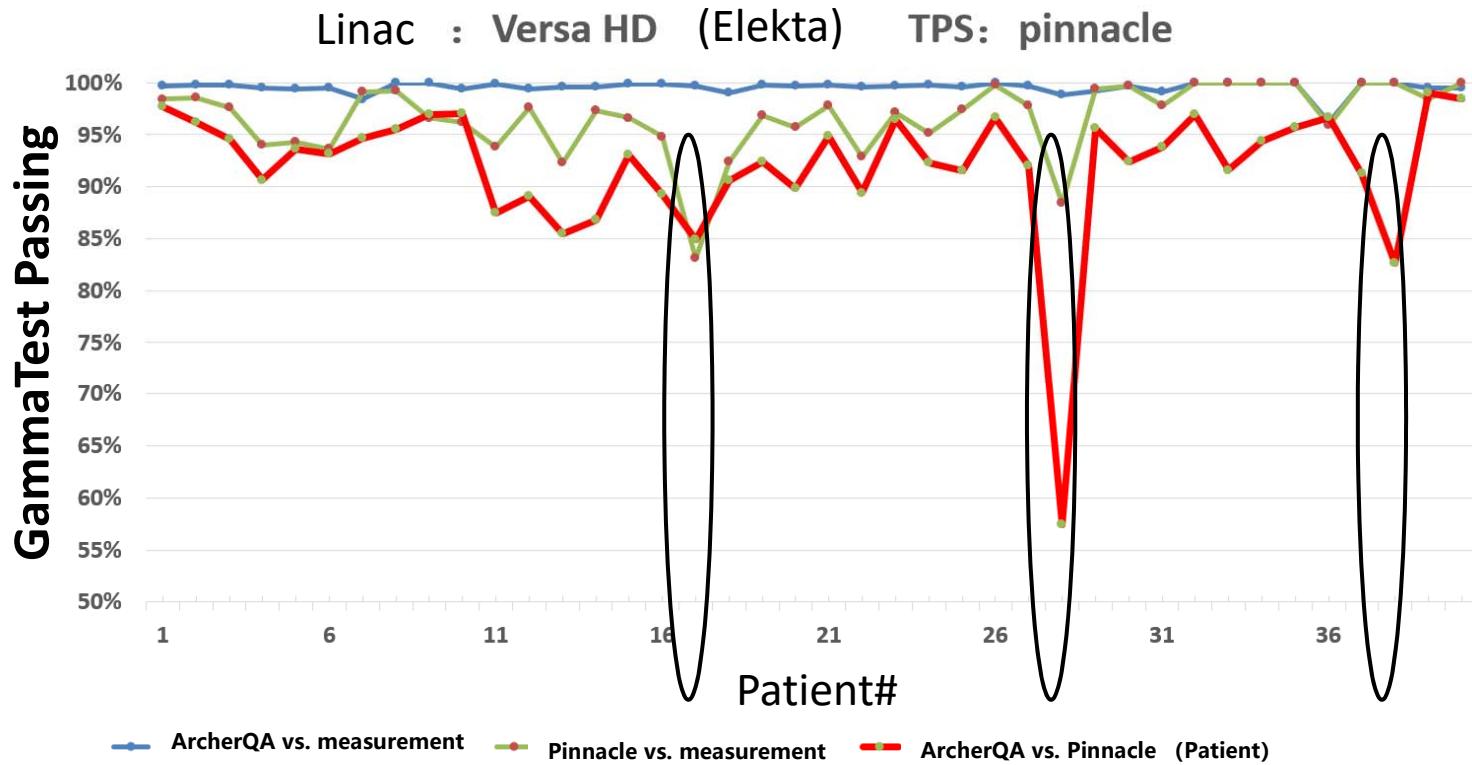
<http://www.wisdom-tech.online/>

- 1. GPU-accelerated Monte Carlo dose engine**
- 2. Data analyze tools, such as 2D and 3D Gamma Test and DVHs**
- 3. TPS and Machine Check**
- 4. Friendly user interfaces, easy to commission and setup**



# ArcherQA Testing Results

## User report #1 Linac old (~6 years)



- **Gamma Test: 3%/3mm, >10%**
- **ArcherQA vs. measurement: GammaTest average ~99.56%**
- **Pinnacle vs. measurement: GammaTest average ~ 96.64%**
- **ArcherQA vs. Pinnacle: gammaTest average ~ 91.94%**

## Clinical applications

- Installed in more than 30 hospitals in China and U.S.
- Checked more than 6400 patient treatment plans
- Linac types include:

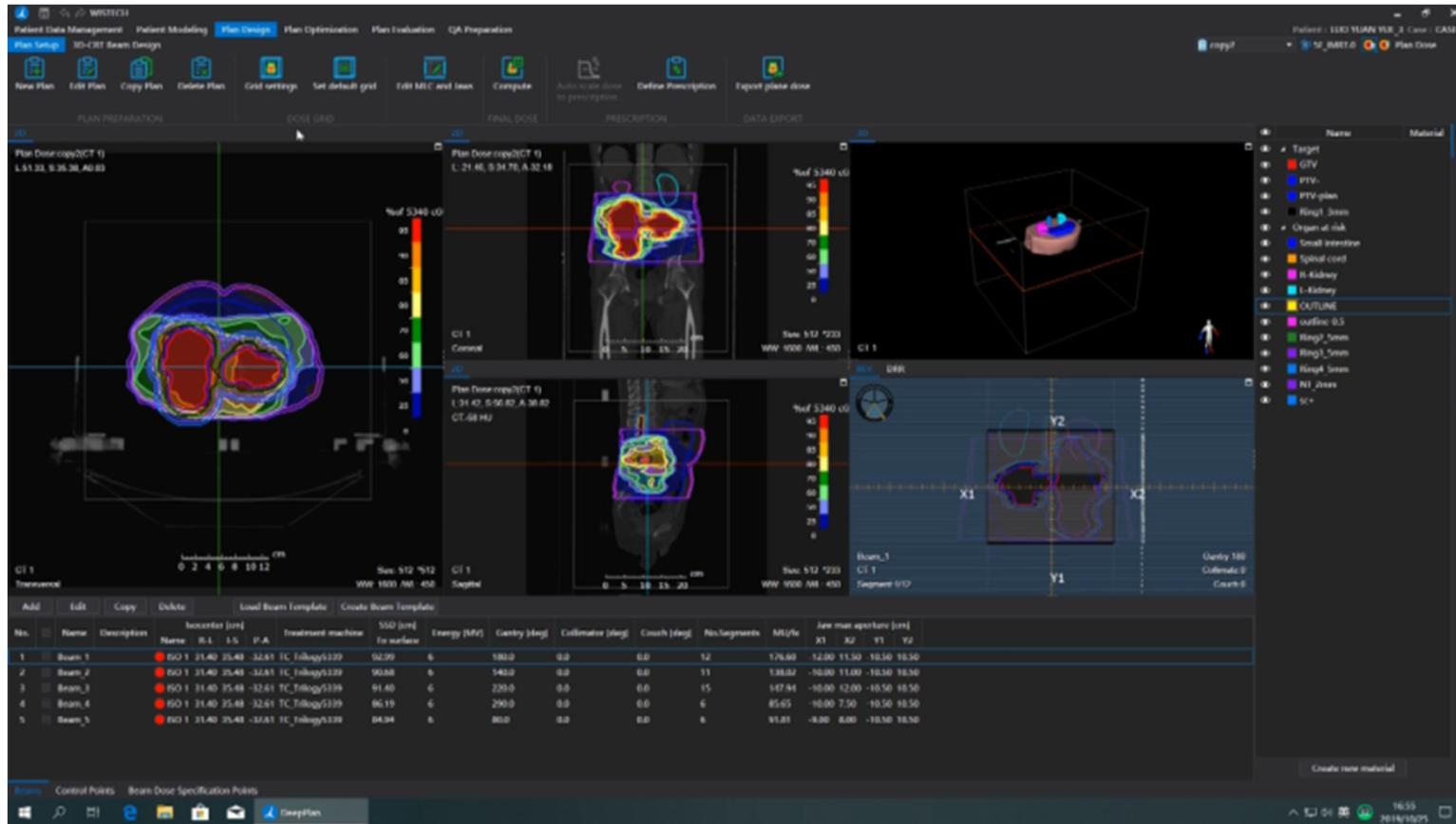
Vendor	Model
Varian	VitalBeam, Truebeam , Halcyon, EDGE, Trilogy
Elekta	Unity, Synergy, Versa HD, Axesse, Infinity, Precise
Accuray	Tomotherapy, CyberKnife

Chinese vendor	Model
United Imaging	506c
ZhongNeng	OMX6i

On-going Clinical Testing

## DeepPlan, Treatment Planning System (TPS)

- Photons, electrons, protons
- Different dose algorithms including GPU-based Monte Carlo
- Integrated with auto contouring and registration



# Summary

1. Computational phantoms grew exponentially towards “personalization” and “multi-scale”
2. AI facilitates clinical medical physics workflow (segmentation tools are well tested)
3. Real-time MC computing provides RT QA solution:

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DeepViewer

