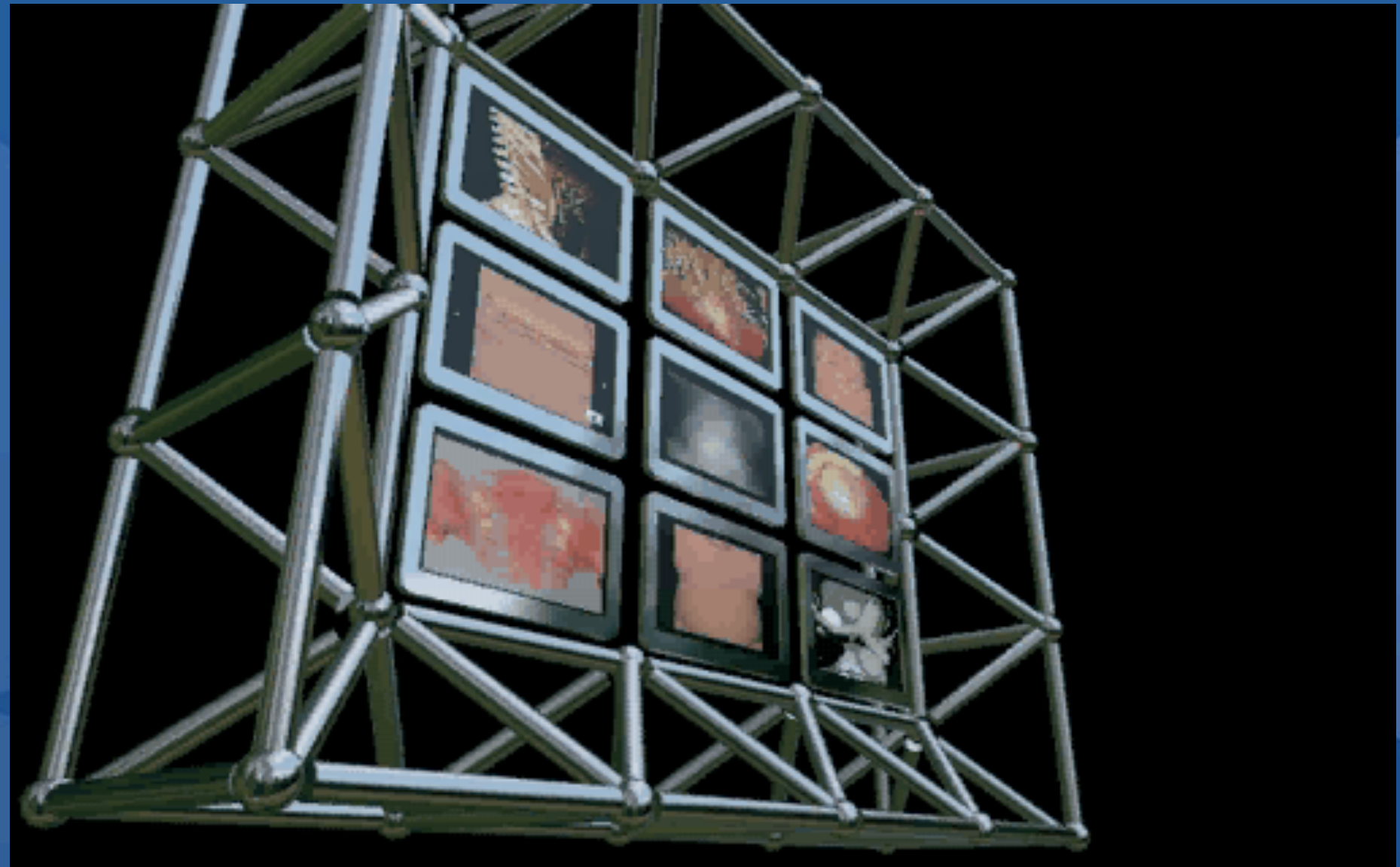


# Image Post-Processing, Workflow, & Interpretation

**Richard L. Hallett, MD**

*Chief, Cardiovascular Imaging  
Northwest Radiology Network  
Indianapolis, IN*

*Adjunct Assistant Professor  
Stanford University  
Stanford, CA*



**Disclosures: None**

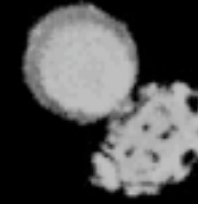
*Online Handouts from Lecture:*

[www.stanford.edu/~hallett](http://www.stanford.edu/~hallett)

# Outline

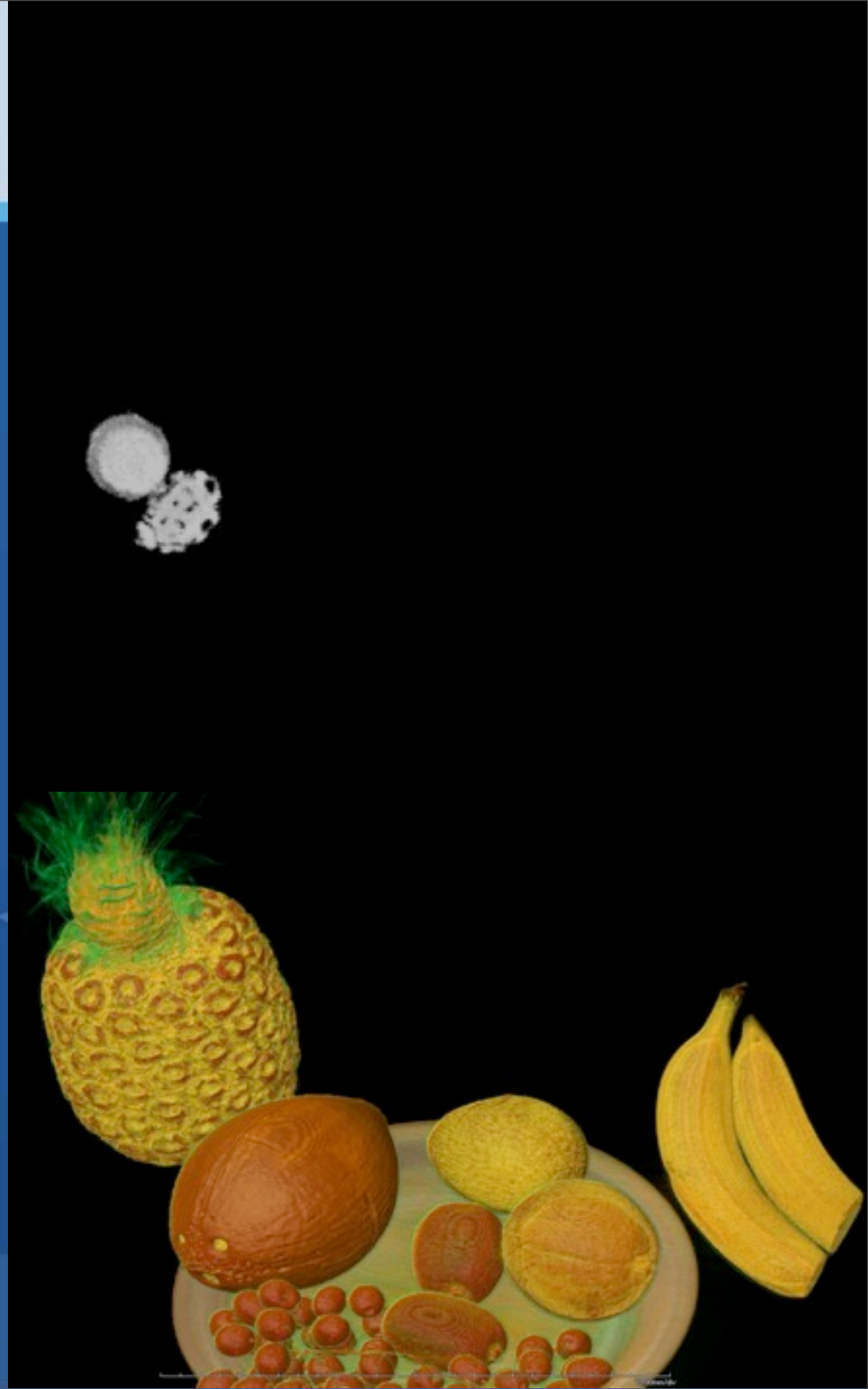
# Outline

- I. Image reconstruction and post-processing techniques**
- II. CTA Workflow and Interpretation Strategies**



# Outline

- I. Image reconstruction and post-processing techniques
- II. CTA Workflow and Interpretation Strategies



# *(Modifiable)* Image reconstruction techniques

1. Raw Data Reconstruction Mathematics
2. Individual Slice / Patient Characteristics
3. Field of View
4. Kernel

10.00mm/div

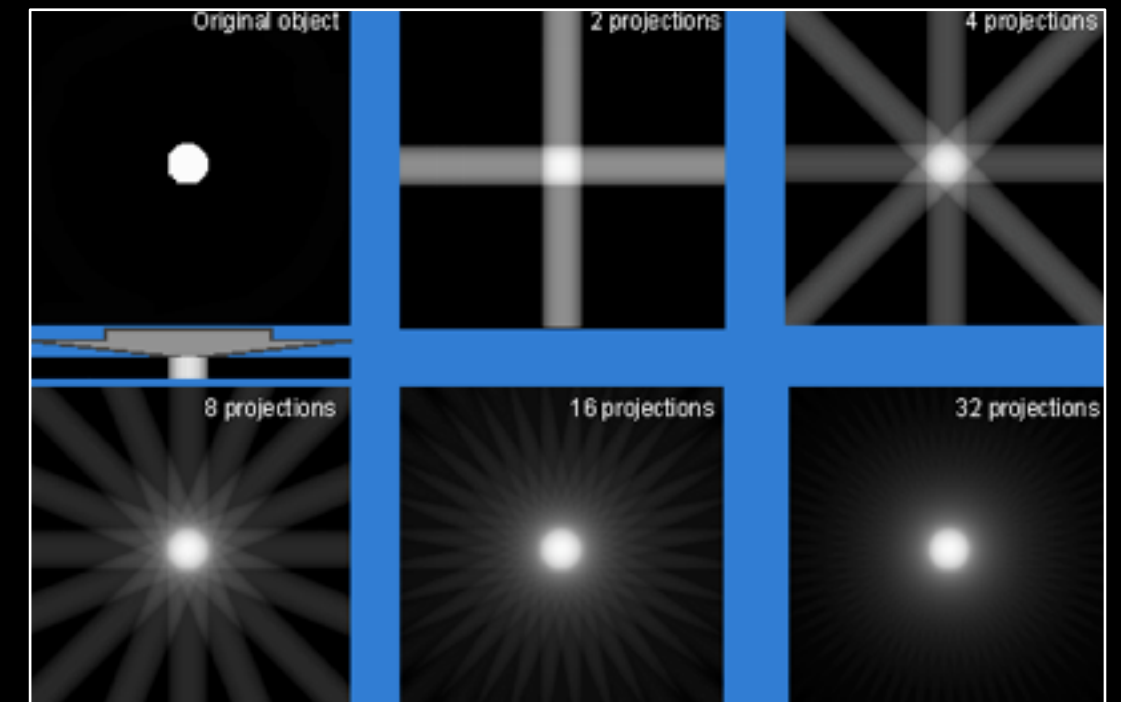
# *(Modifiable)* Image reconstruction techniques

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10.00mm/div

# TRADITIONAL Raw Data Reconstruction

- Traditionally reconstructed using Filtered Back Projection (FBP)
- Necessary ASSUMPTIONS:
  - Focal spot infinitely small
  - Detector is single point in center of detector cell
  - Reconstructed voxel - no shape or size
  - Measured signal has no error from photon statistics or image noise



10.00mm/div

# “New” Data Reconstruction Mathematics

- Iterative Reconstruction (IR)
  - Used in SPECT and PET years ago.....
  - Models CT system optics (geometric information) as well as statistics (noise)
    - ➔ Compare model to real raw data, correct, repeat
    - ➔ Model can be *iterated* over and over until image is essentially constant
  - Reduced noise, but computationally expensive

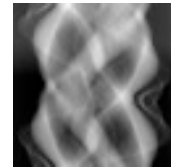
10.00mm/div

Hara AK, et al. Am J Roentgenol. 2009;193(9):764-771

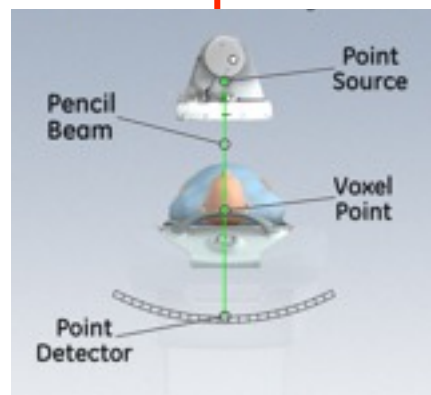
# Conventional FBP vs. Iterative Reconstruction

## FBP Reconstruction

Raw Data



Assume "ideal" system



Perfect Sample

Image

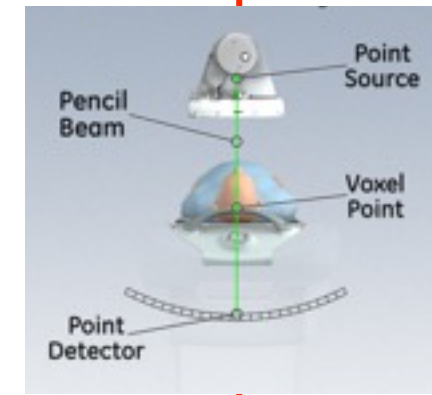
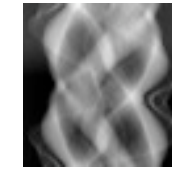


CPU required

Simple  
Fast reconstruction  
High Noise

## ASiR Reconstruction

Raw Data



Powerful  
Statistical Model

Image



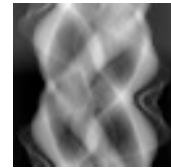
CPU required

Better Image Quality  
Low Noise  
Lower Dose

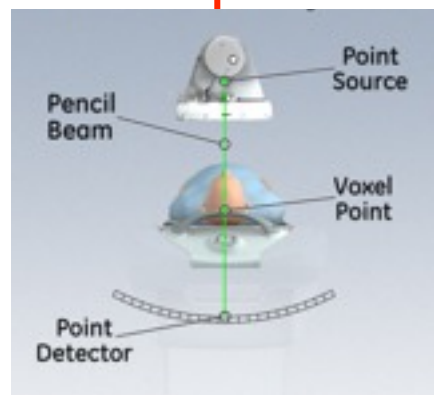
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## FBP Reconstruction

Raw Data



Assume "ideal" system



Perfect Sample

Image

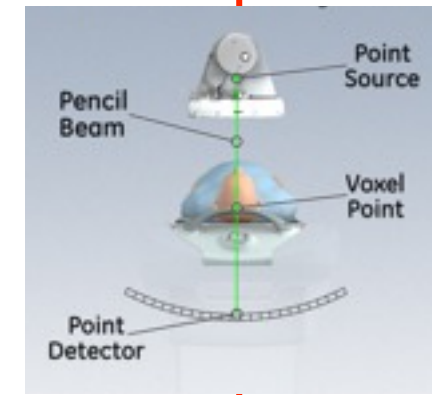
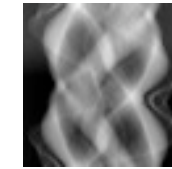


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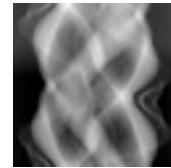
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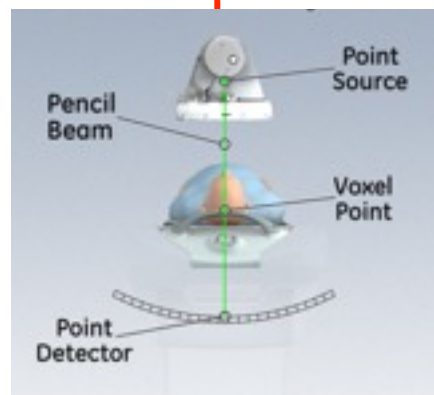
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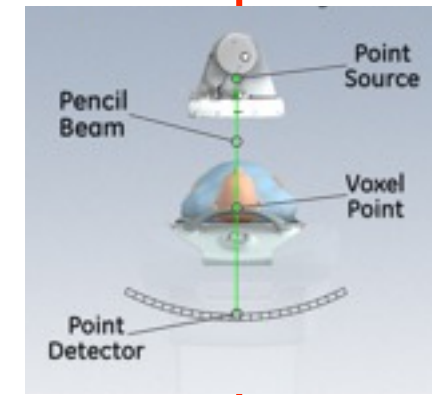
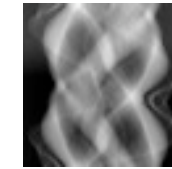


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Simple  
Fast reconstruction  
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## ASiR Reconstruction

Raw Data



Powerful  
Statistical Model

Image



CPU required

Better Image Quality  
Low Noise  
Lower Dose

# Iterative Reconstruction Benefits

- Up to 50% dose reduction is possible at same image noise
- **OR:** Improved image quality at same dose
- 40% improvement in low contrast detectability



2007, Std.  
Technique  
CTDI=19

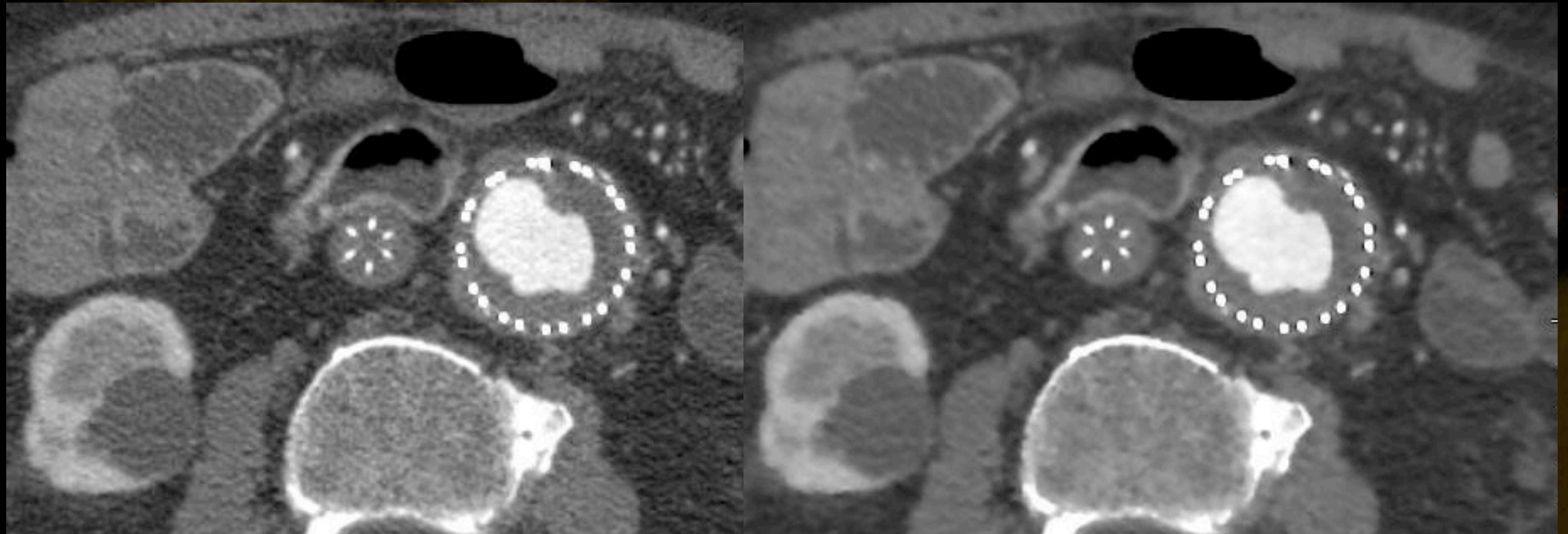


2008  
ASiR  
CTDI=9

10.00mm/div

Images courtesy of Mayo Clinic Arizona

# Iterative Reconstruction: Appearance

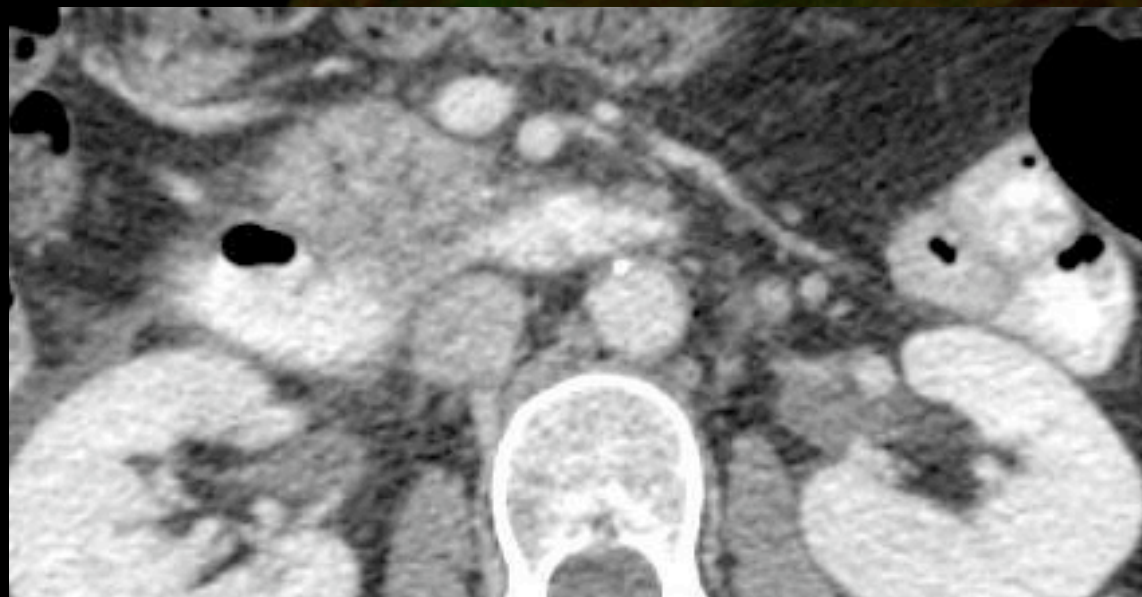


**0% IR**

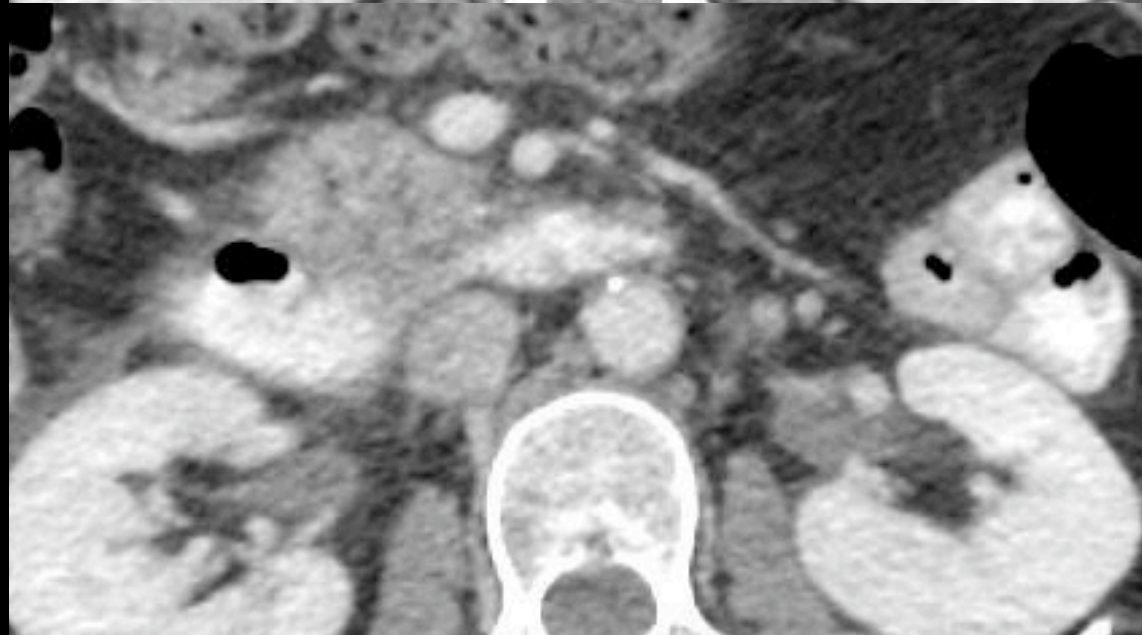
**100% IR**

10.00mm/div

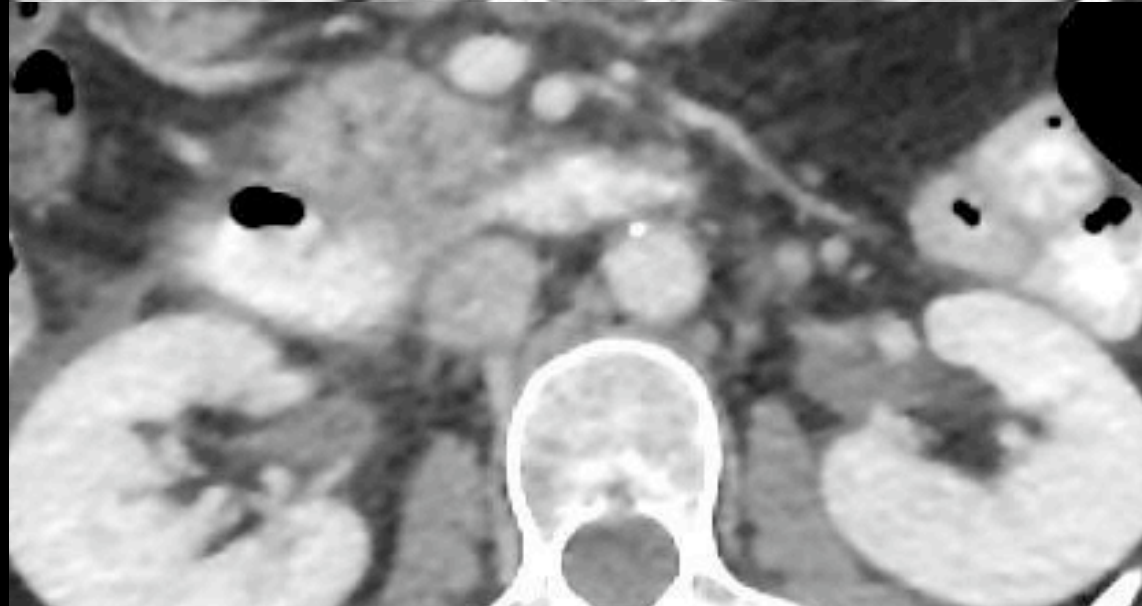
# Iterative Reconstruction Appearance



**0% IR**



**40% IR**

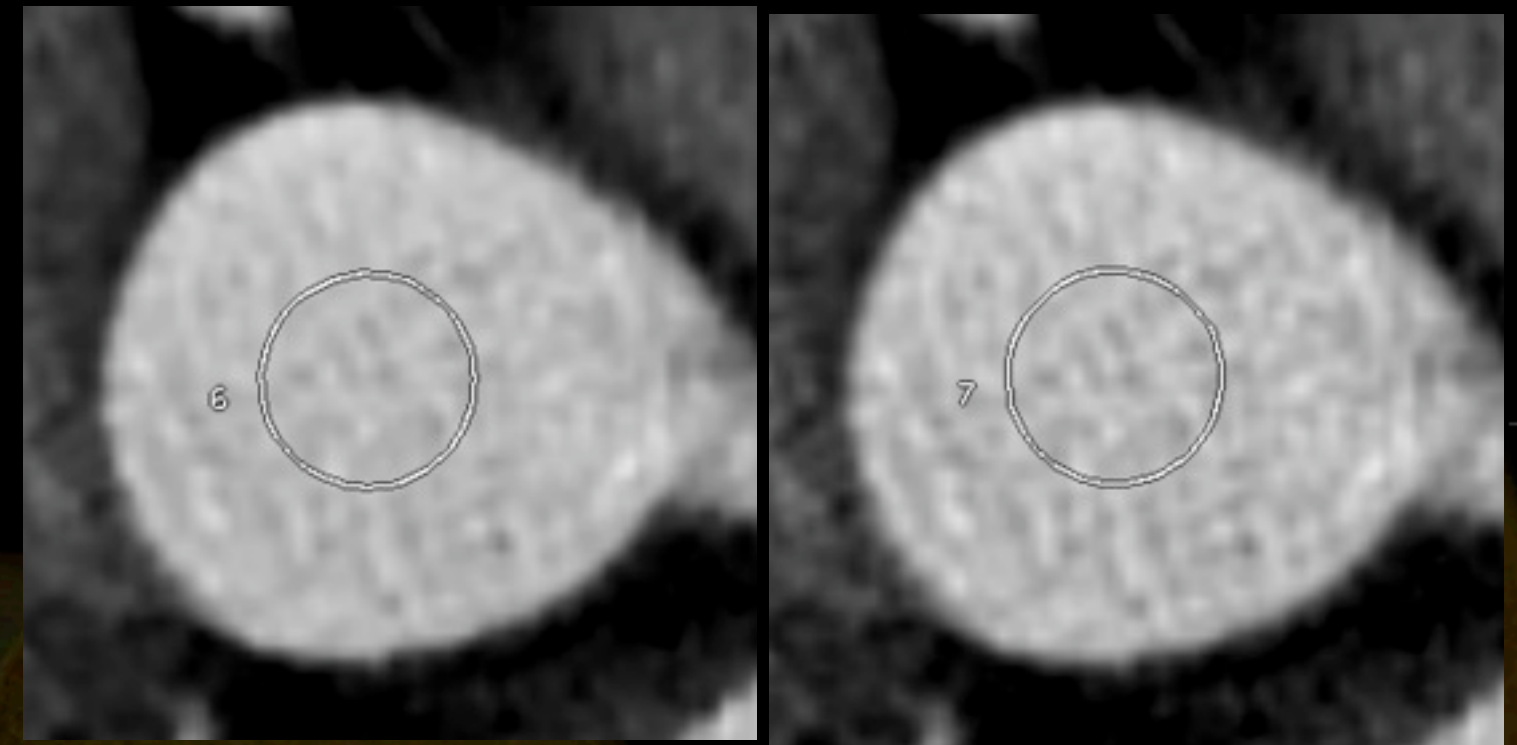


**100% IR**

10.00mm/div

# Iterative Recon for CCTA: the ERASIR STUDY

- 574 consecutive pts at 3 sites referred for CCTA: FBP vs. 40% ASiR blend
- **27% dose reduction from IR utilization**, without increased image noise or non-evaluable segments
- 45% total reduction including other scan parameters (100 kV, etc)



FBP

40% ASIR

Density, HU (signal)	718.6	719.3
SD (noise)	52.3	38.5

*Leipsic J, et al. AJR 2010; 195:655-660*

10.00mm/div

# *(Modifiable)* Image reconstruction techniques

1. Raw Data Reconstruction Mathematics
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3. Field of View
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10.00mm/div

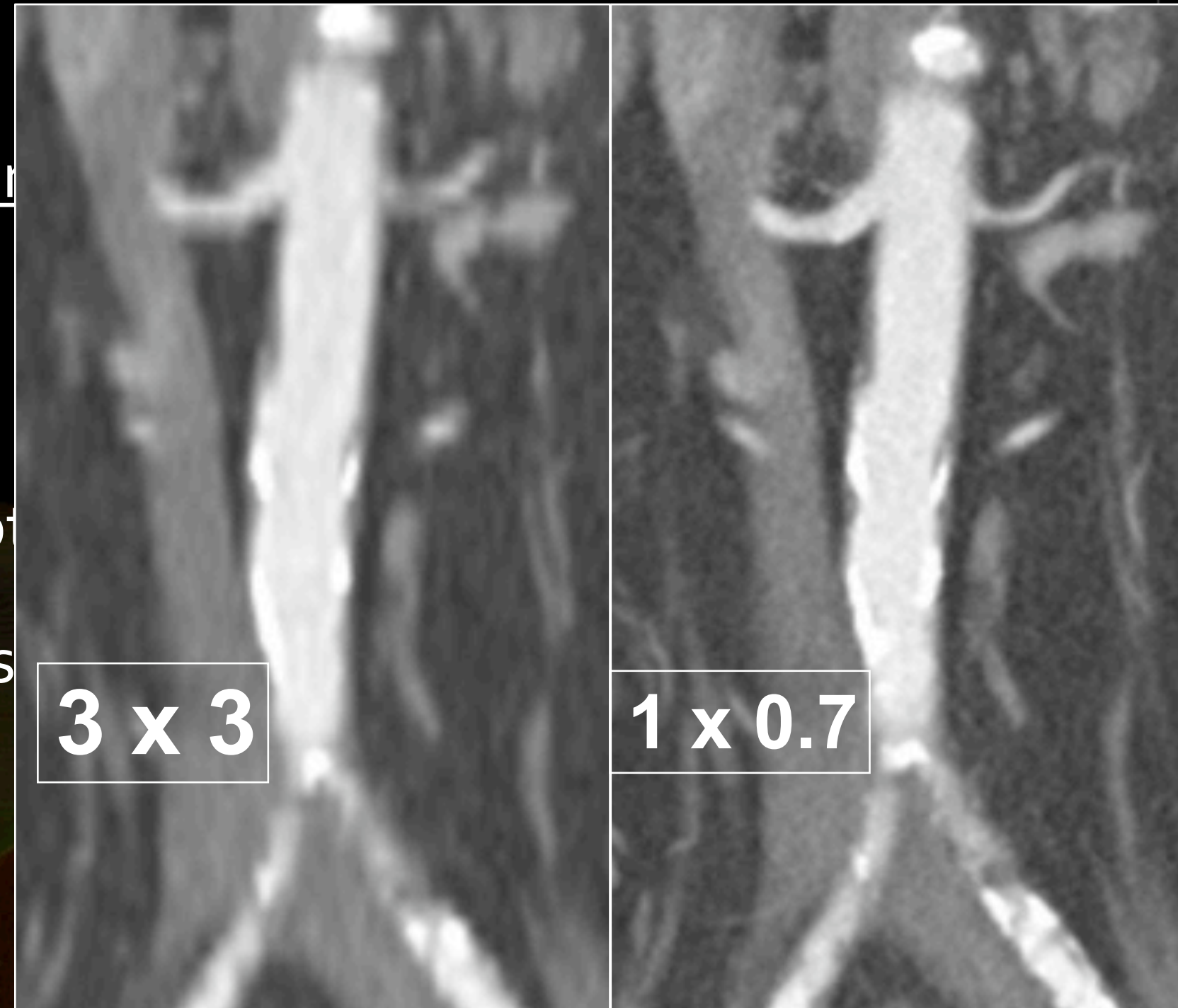
# Characteristics of the CT “slice”...

- “Effective” slice thickness
  - defined by the selection of collimator thickness during scan acquisition
- Thicker (but not thinner) recons
- Multi-planar reconstructions (MPR) obtained by *interpolation*
- MPR enhanced if your initial dataset is overlapped by ~ 30%
  - e.g. 1mm ST at 0.7 mm RI
  - Less “aliasing” (stairstep)

10.00mm/div

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  - e.g. 1mm ST at 0.7 mm RI
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# Tweaking / Help for Tough Datasets

- Large Patients:
  - Scan with thicker collimation (1.25 - 2.5 mm)
  - Use 140 kV
  - Slow down gantry rotation
- Small Patients:
  - Use 100 kV

10.00mm/div

# *(Modifiable)* Image reconstruction techniques

1. Raw Data Reconstruction Mathematics
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10.00mm/div

# Effect of changing FOV

- Standard CT image:
  - 512x512, FOV = 30 cm
  - Pixel size ~ 0.35 mm<sup>2</sup>
- Small FOV:
  - 512x512, FOV = 15 cm
  - Pixel size ~ 0.10 mm<sup>2</sup>
- BUT: "Isotropic" voxels easier to obtain at thicker slice / larger FOV

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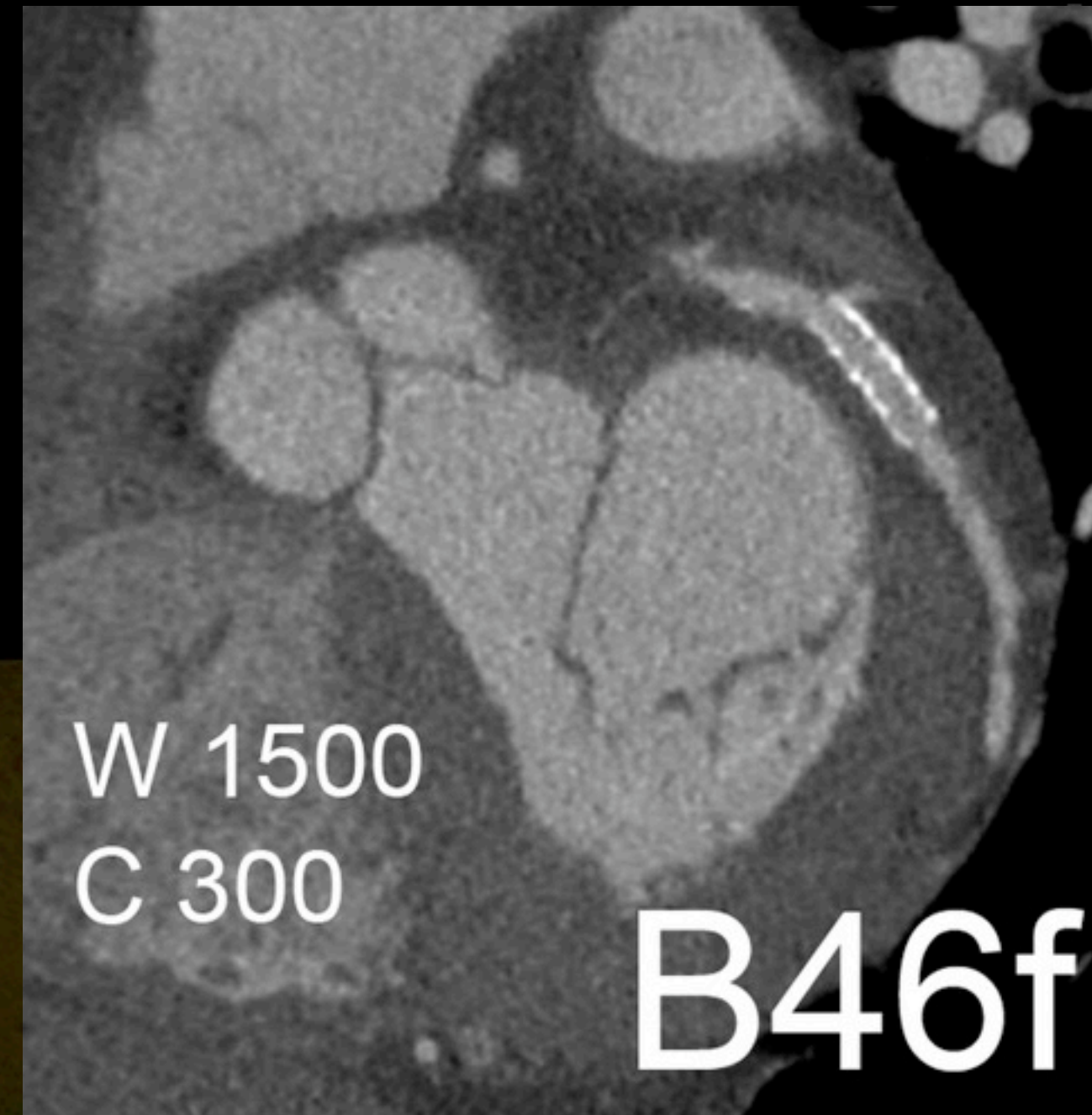
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10.00mm/div

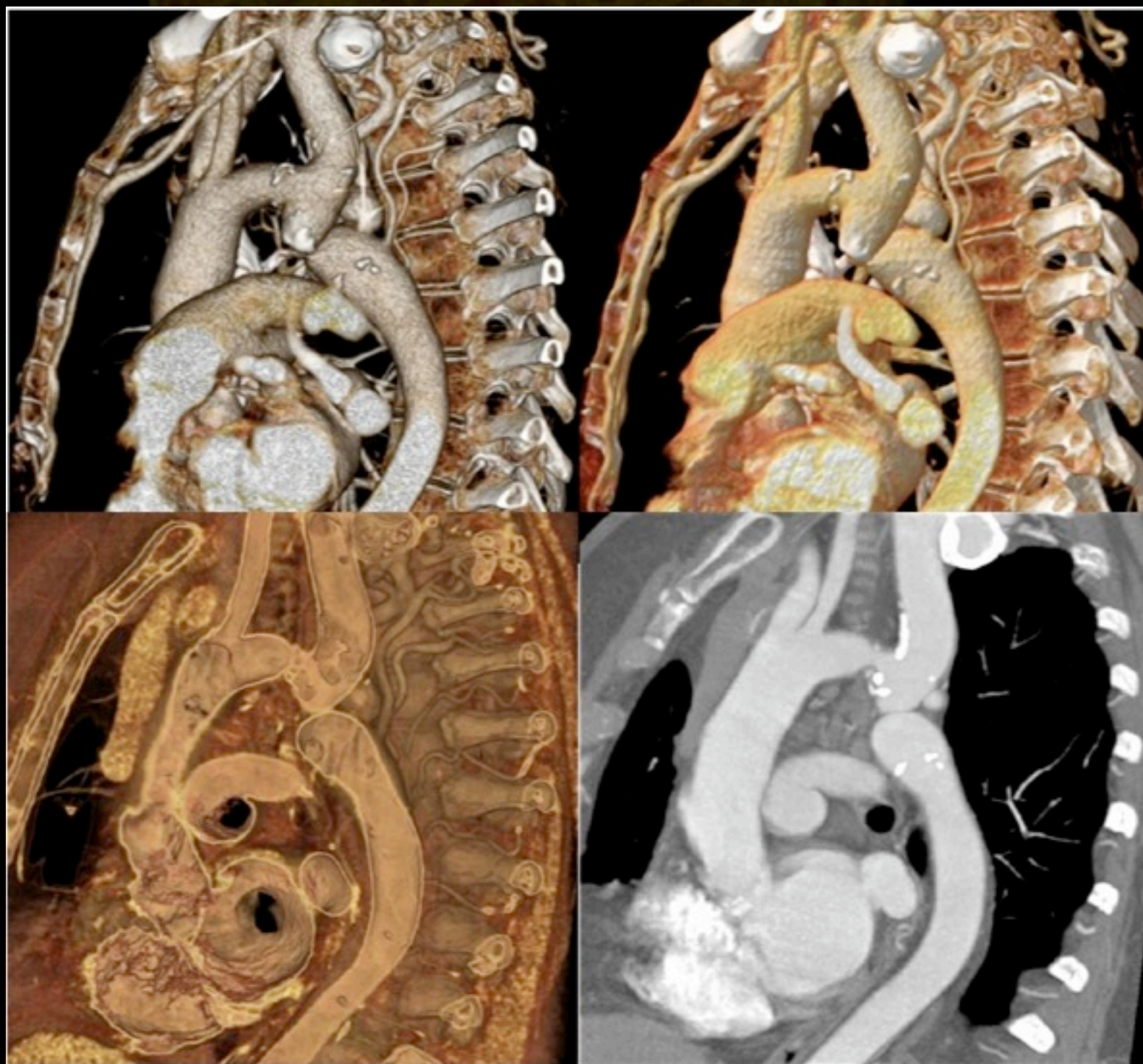
# Effect of Recon Kernel

- **Softer kernel: Less noise, less sharp**
  - Better 3D / Multiplanar recons
- **Sharper kernel: Higher detail, more noise**
  - **STENTS!!** (coronary, peripheral)
    - Less blooming artifact



Pugliese, F. et al. Radiographics 2006;26:887-904

# Image Post-Processing



Review of Image Types  
New Directions

10.00mm/div

# Reconstruction “Alphabet Soup”

- MPR
- MIP
- MINIP
- AIP (Raysum)
- CPR
- MP-CPR
- VR
- BPI-VR
- V-IVUS
- 4-D



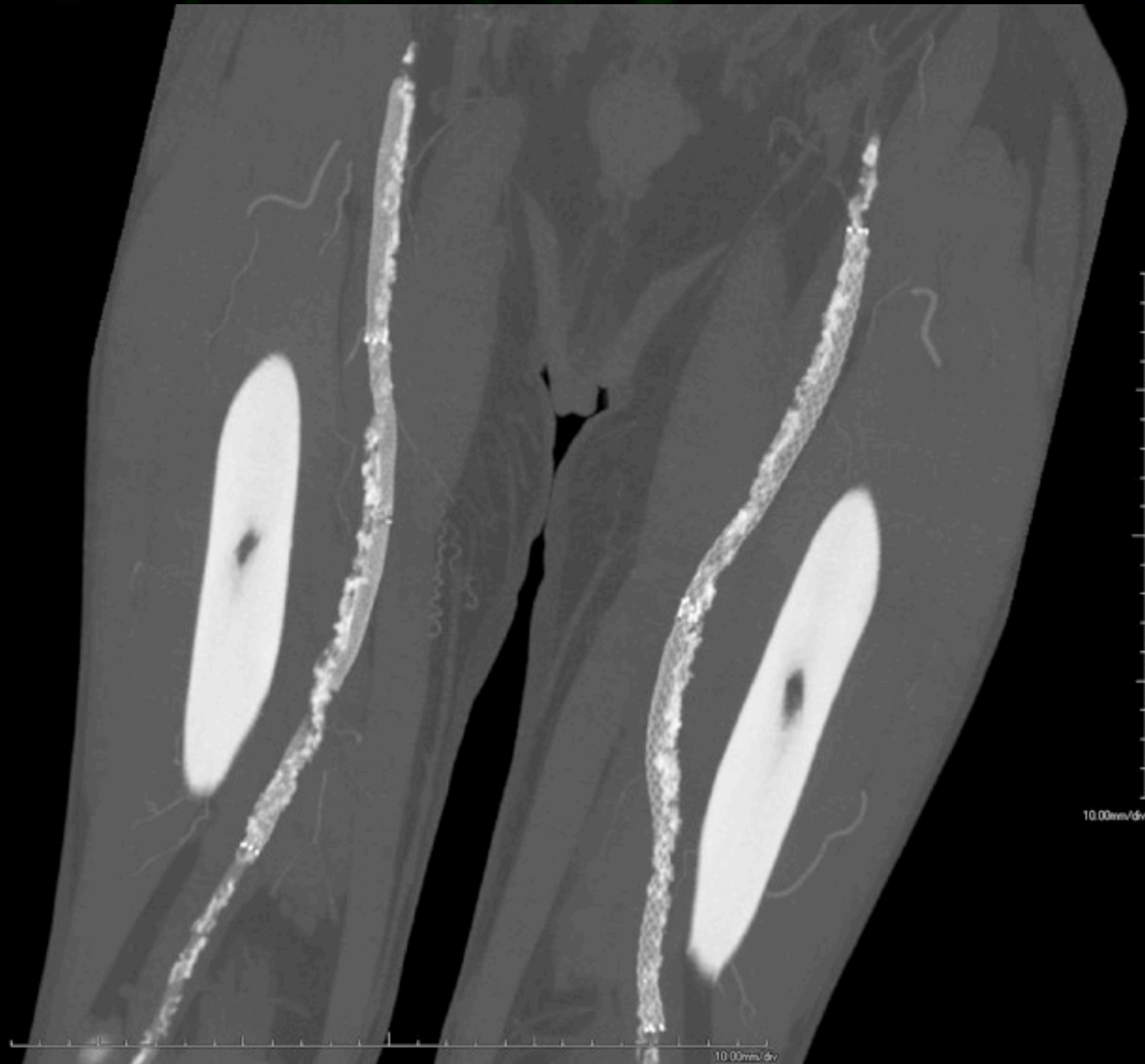
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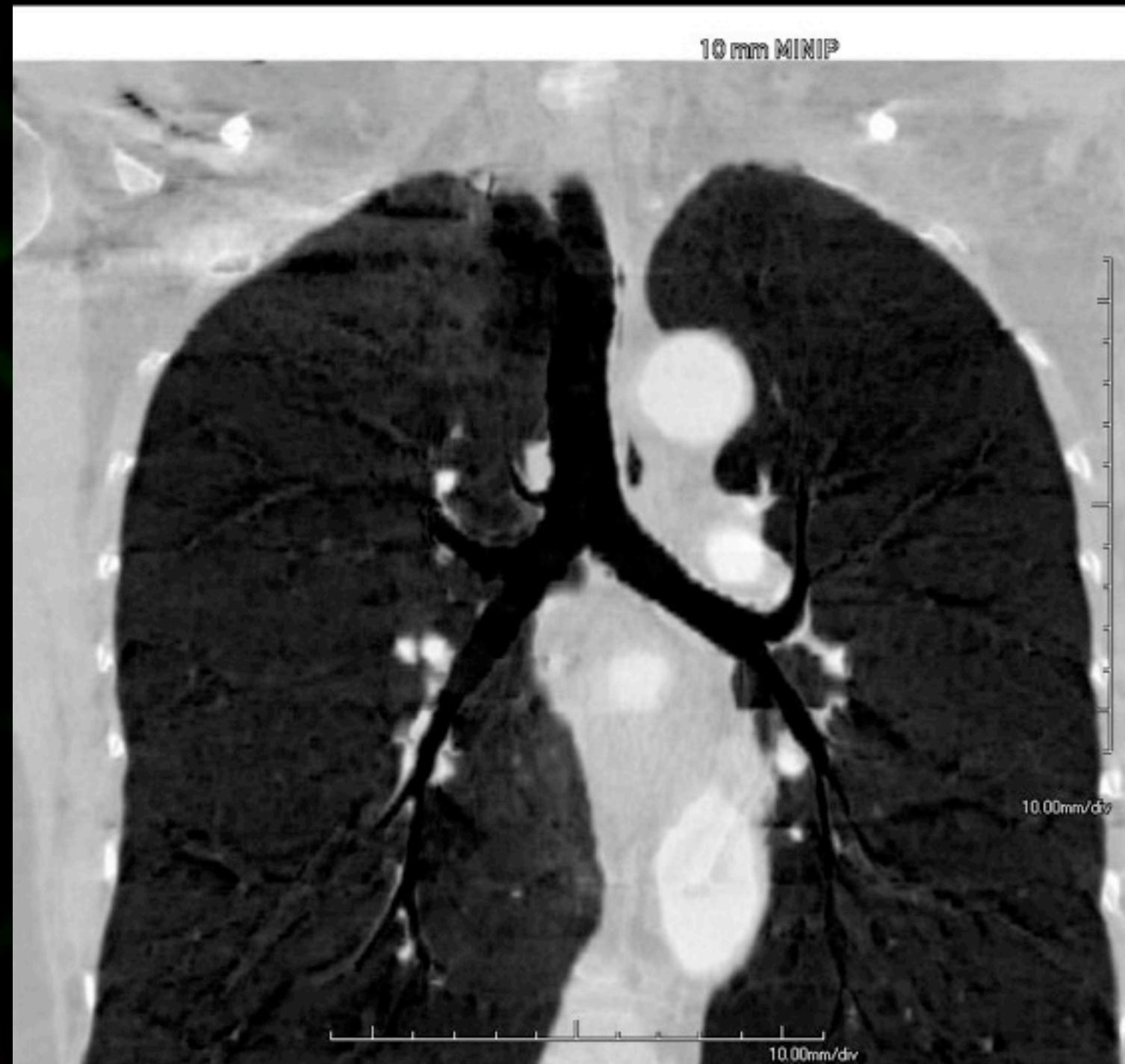
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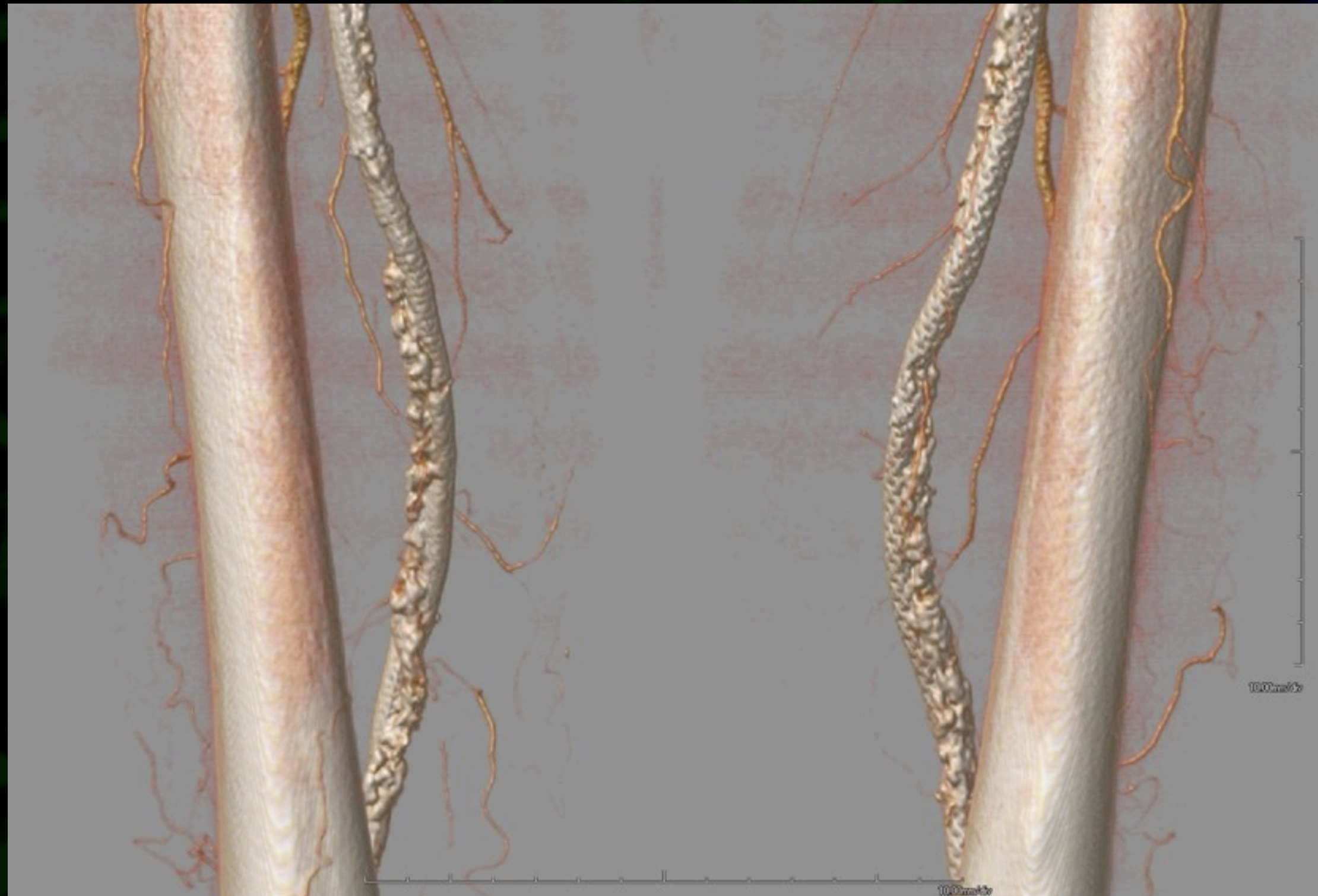
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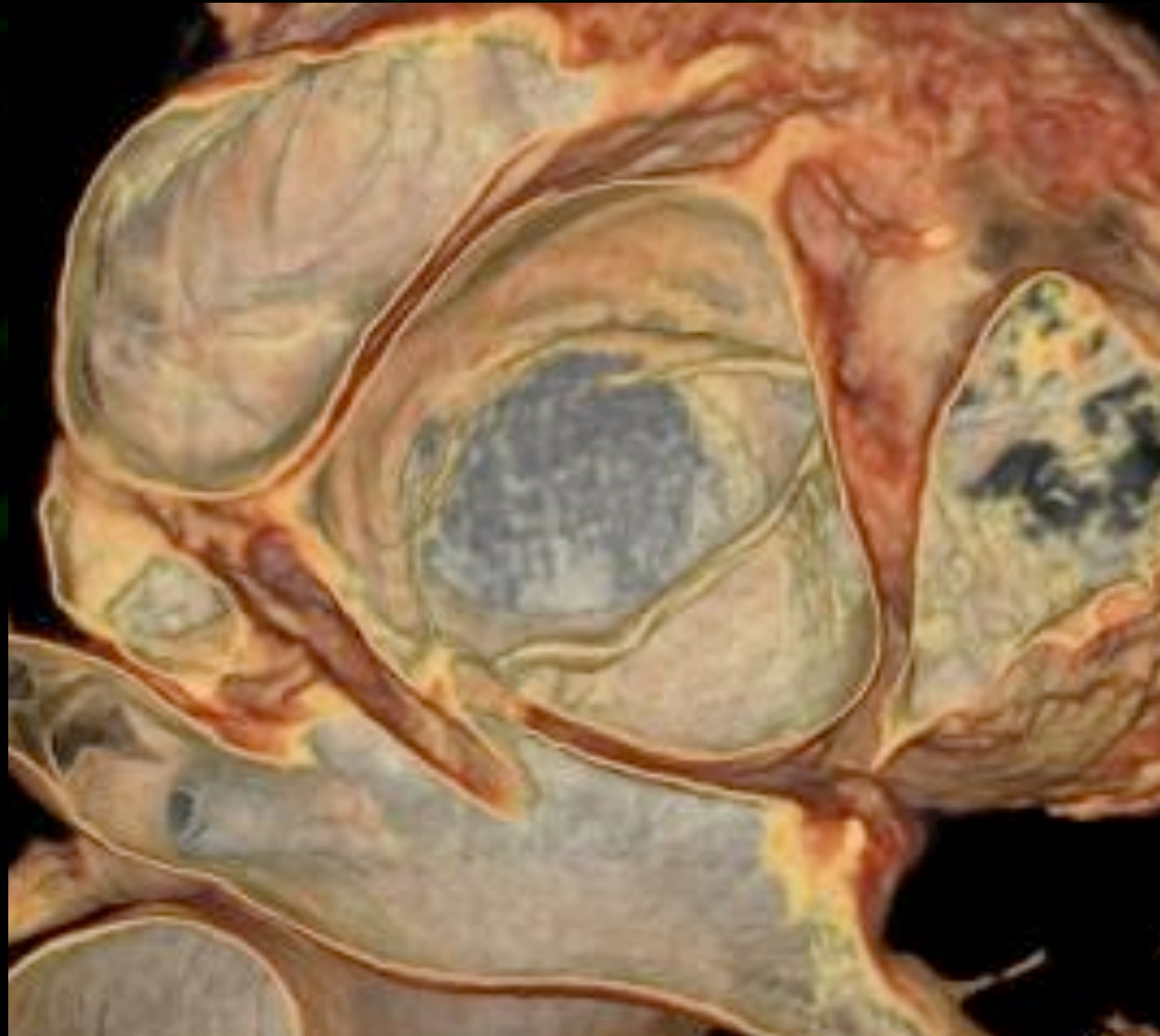
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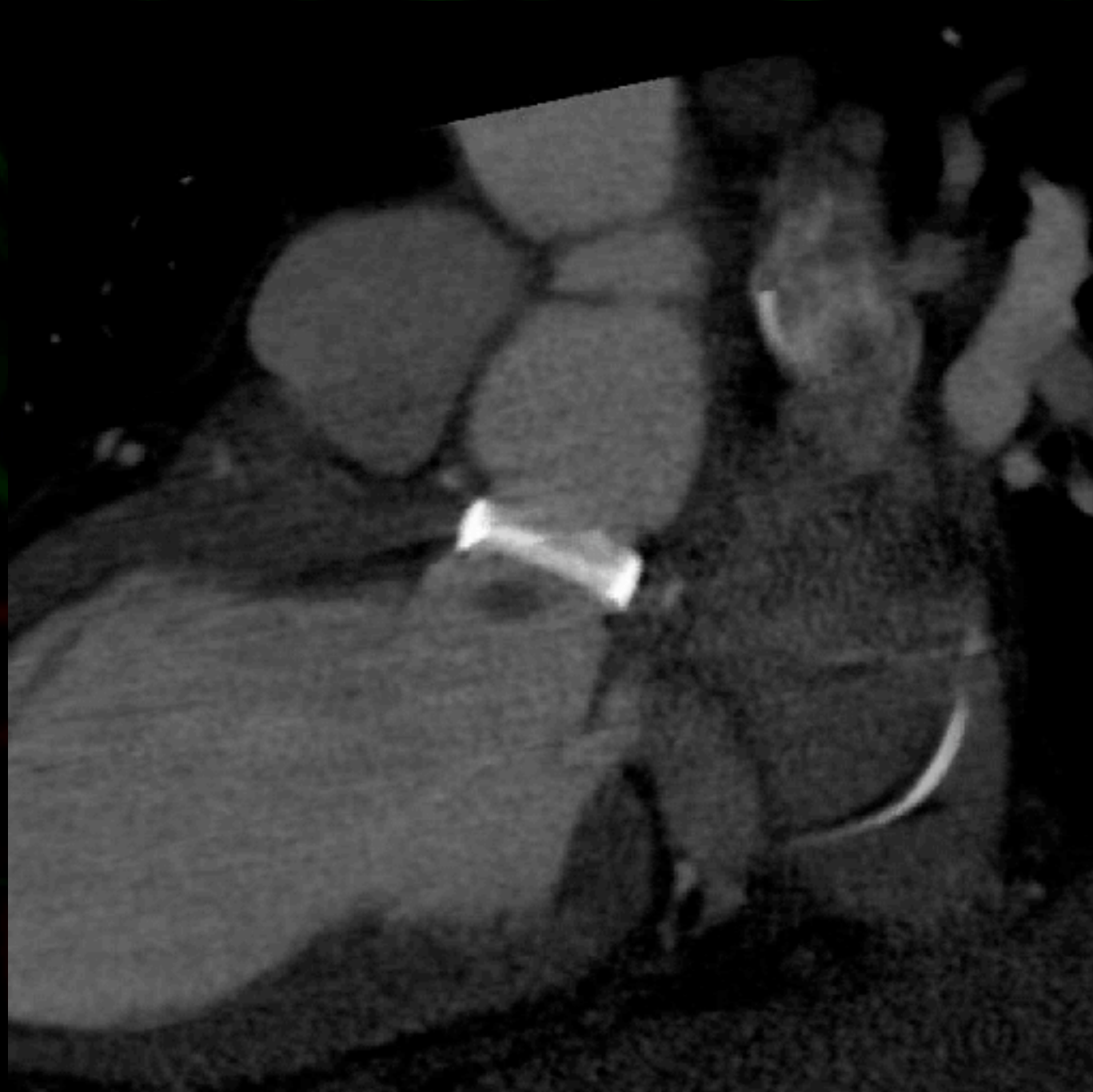
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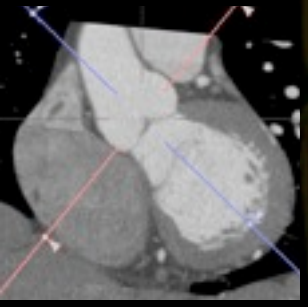
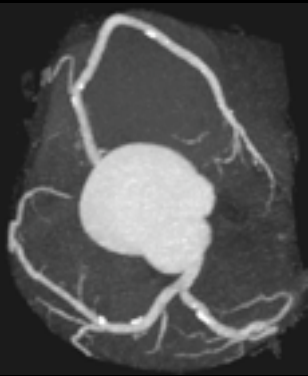
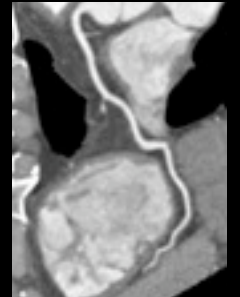

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# Reconstruction “Alphabet Soup”

- 4-D



	Major Uses	Advantages	Disadvantages
<b>MPR</b> 	Stenosis, vessel wall analysis Lung nodule measurement Orthogonal Measurements	Accurate for stenosis, nodule, orthogonal measurements Calcification, stent evaluation <b>"Thick MPR": salvage noisy datasets</b>	Limited spatial relationships <b>Limited display if curving vessel</b>
<b>MIP<sub>(MINIP)</sub></b> 	Angiographic overview, contextual with adjacent structures Lung nodule detection (coronal STS) Valves, Airways (MINIP)	Depicts course of small and/or poorly enhancing vessels Object - background contrast	Vessel, bone, visceral overlap Limited stent, calcium evaluation <b>Stenosis Overestimation</b> <b>NOISE IS ADDITIVE!!</b>
<b>CPR</b> 	Flow lumen, vessel wall analysis Curved Objects	Best for mural stenosis, occlusions, calcifications, stents Slice through display (perpendicular to CPR)	Distortion of extra-vascular structures <b>Dependent on accurate centerline (Needs Oversight)</b>
<b>VR</b> 	Angiographic overview, contextual with adjacent structures Pre-procedural planning	• Best for complex relationship display <b>WOW factor</b>	<b>Opacity transfer function and operator dependent</b> No accurate measurements <small>10.00mm/div</small>

# Recent Reconstruction Advances

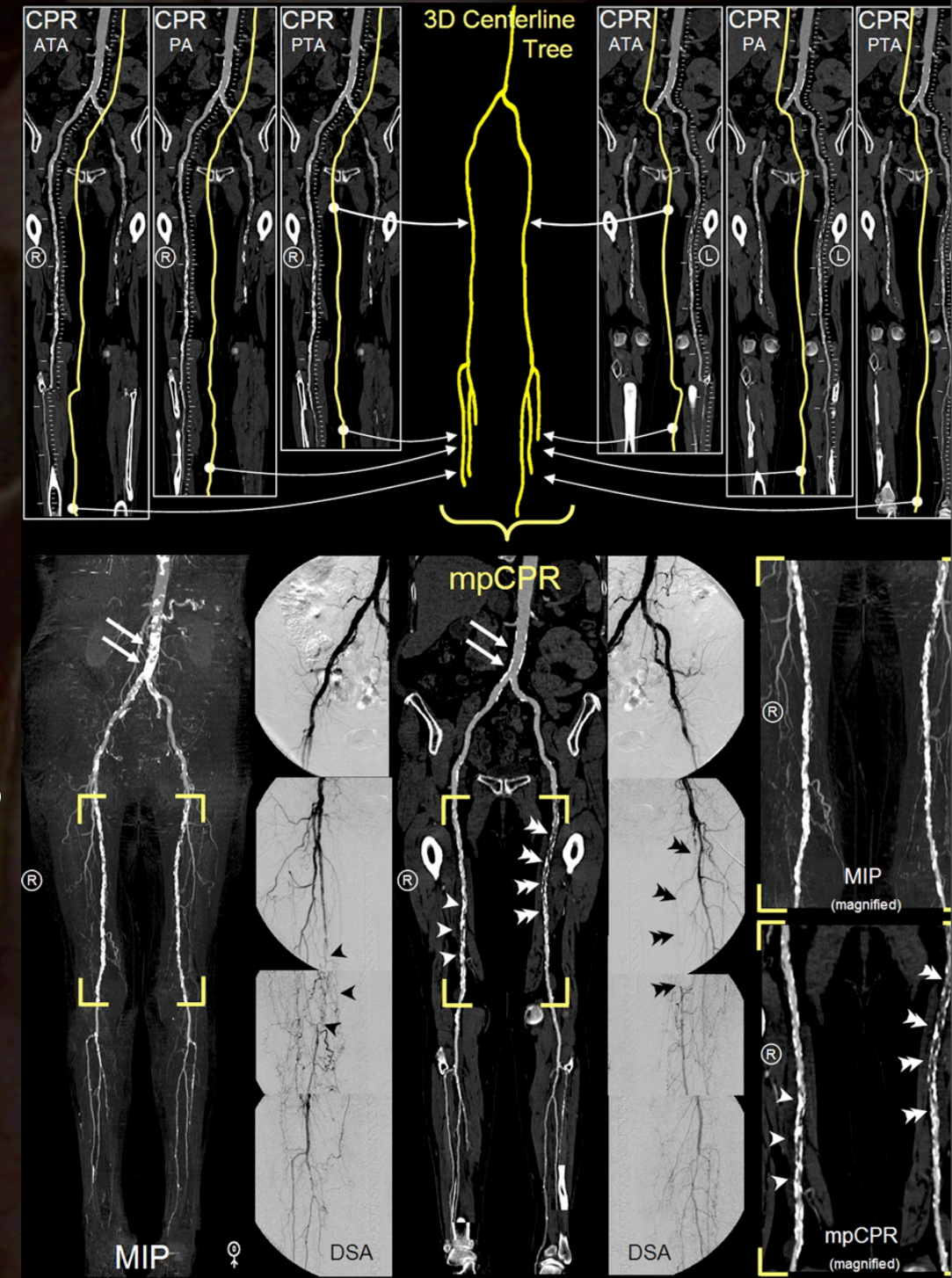
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  - ◆ Simultaneous CPRs of all LE vessels at once
- ◆ Partial Vector Shape Projection (PVSP)
  - ◆ Knowledge-based extrapolation of centerlines for SFA occlusions
  - ◆ Independent of local density and gradient information

*Roos JE, et al. Radiology 2007; 244 (1) 281-90.*

*Rakshe T, et al. Med Image Analysis 2007; 11 (2) 157-68.*

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# CTA Workflow and Interpretation

Online Handouts from Lecture:  
[www.stanford.edu/~hallett](http://www.stanford.edu/~hallett)

# Goals of CTA workflow

- Process studies efficiently
- Capture all appropriate charge codes
- Provide access to thin-slice datasets for radiologist interpretation and review
- Provide timely reports to referring clinicians / services

# Coordinated Efforts Yield Best Results

(think Ultrasound!)

---

- Physician-directed for primary interpretation
- Technologist-directed for measurement, documentation, reporting ~ 3D Laboratory

# Physician -Directed 3D

- Volumetric Interpretation via:
  - Workstation
  - Thin Client - Server
  - PACS
- Like Ultrasound, “Clarify” images obtained by 3D Lab / Techs
- Output:
  - Sent to PACS, emailed to referring MD
    - can also real-time “consult”

# Technologist (3D Lab) Tasks

- Template-Driven processing of cases:
  - Segmentation
  - Detailed measurements, volumes
  - Consistent output format
- Triage urgent exams
- Temporal tracking of measurements (AAA)
- Transfer of data to MDs, clinical reports, and PACS

# Interpretation Strategies:

- Good: Stack Review (PACS, workstations, etc)
- Better: Volumetric Review
  - “Volumetric datasets are best reviewed volumetrically”
- Interactive MPR review more accurate than review of pre-rendered CPR, MIP, etc

# What is An Effective Platform for Volumetric Interpretation?

- ◆ Fast and available
- ◆ Ergonomic
- ◆ Easy DICOM Import / Export
- ◆ email, JPG capabilities
- ◆ Allows Comparison to Prior / Loads multiple datasets

# *Platform Options*

- *Dedicated workstation*
  - + Most features, high performance
  - Local availability only, relatively expensive
- *Software running on standard PC*
  - + Distributed (depends on pricing), relatively inexpensive
  - Less features, lower performance
- *Server-Client*
  - + Distributed, lowest cost per “workstation”
  - Less features, performance is network dependent

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PRACTICE DEPENDENT !!

# How I do it.....

- RTs: process CPR, MIP, volumes
- Read from thin client whenever possible
- VR Overview then review axials
- Targeted interactive STS MIP and MPR evaluation of abnormalities
- My pertinent images - sent to PACS as a series
- VR images, stenosis evaluation emailed to referring MD
- Web-based “consult” feature: Use for intra-op

# How you should do it.....

- Find a workflow that works for you
- Review all the data
- Be efficient
- Communicate your results!

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# Conclusions

10.00mm/div

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- Image Reconstruction:
  - Remember how to improve and troubleshoot image reconstruction
  - Exploit new reconstruction techniques to save dose and/or improve quality

10.00mm/div

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- Image Reconstruction:
  - Remember how to improve and troubleshoot image reconstruction
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- Workflow:
  - View 3D like ultrasound- develop, train, trust techs
  - Thin client review – eases bandwidth and location requirements

10.00mm/div

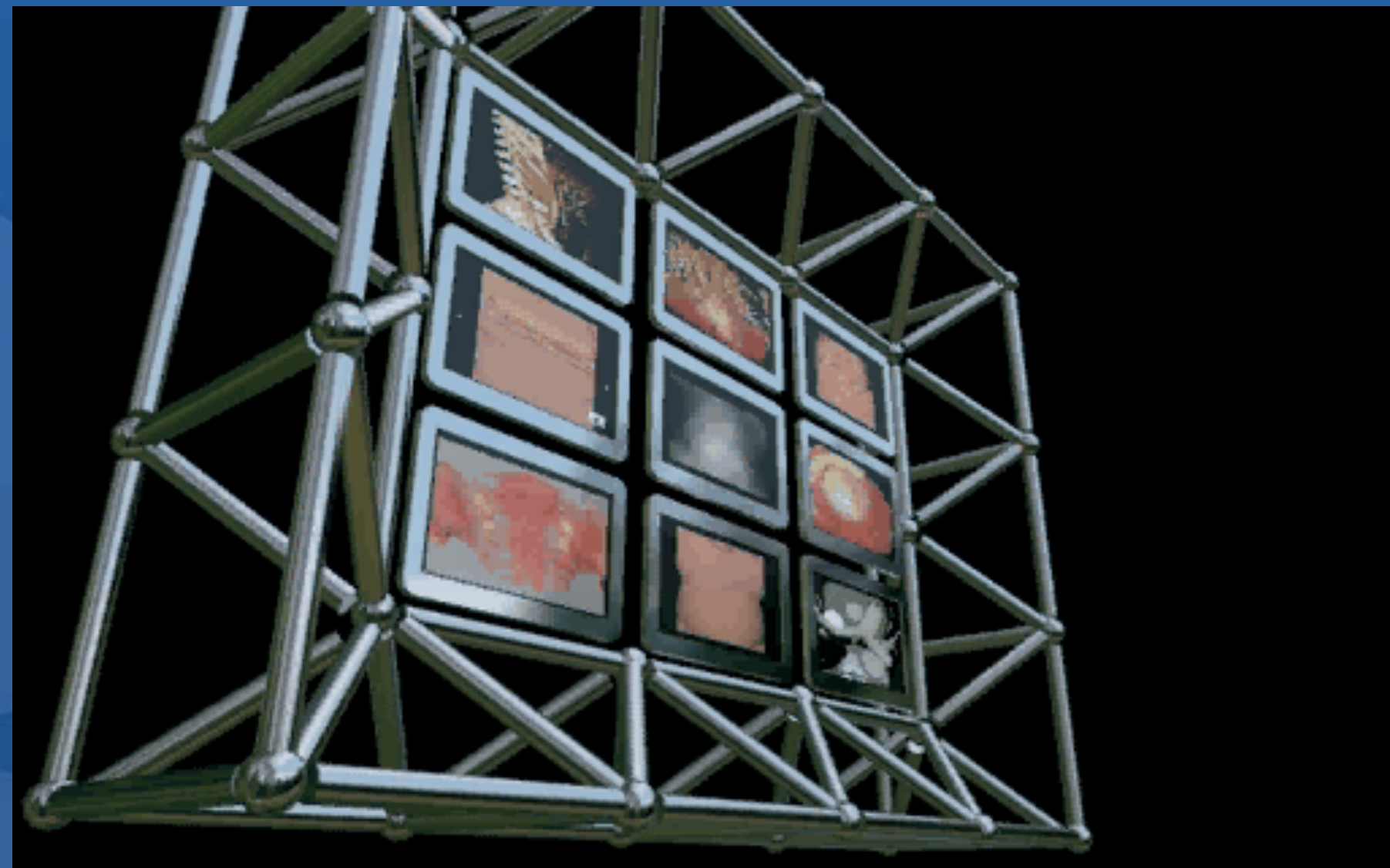
# Conclusions

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  - Exploit new reconstruction techniques to save dose and/or improve quality
- Workflow:
  - View 3D like ultrasound- develop, train, trust techs
  - Thin client review – eases bandwidth and location requirements
- Interpretation:
  - Develop a consistent reading algorithm, always have the source (thin) data available
  - Share your results!

10.00mm/div

# THANK YOU!!

Laura Pierce, MPA, RT (R) (CT) – Duke 3D Lab  
Mina Thakur, RT(R) (CT) – Riverview Hospital  
Jennifer Martin RT (R) (CT) - St. Vincent Indianapolis Hospital



Online Handouts from Lecture:  
[www.stanford.edu/~hallett](http://www.stanford.edu/~hallett)

# Further Reading:

- Image Reconstruction:

- Rubin GD, Sedat P, Wei JL: *Ch. 6. Postprocessing and Data Analysis*. In: Rubin GD and Rofsky N. CT and MR Angiography: Comprehensive Vascular Assessment Lippincott, Williams and Wilkins, 2008
- Barrett JF, *RadioGraphics* 2004;24:1679-1691
- Ch. 4: Image Reconstruction and Review. In: Lipson SA: MDCT and 3D Workstations. Springer, 2006.
- Luccichenti G, et al. *Eur Radiol* 2005; 15: 2146 - 2156
- Parrish FJ, *AJR* 2007; 189:528-534
- Dalrymple NC, *RadioGraphics* 2005;25:1409-1428
- Hara AK, et al. *Am J Roentgenol*. 2009;193(9):764-771
- Roos JE, et al. *Acad Radiol* 2009; 16 (6) 646-653.

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# Further Reading

- CTA Workflow:
  - [http://www.imagingcenterinstitute.com/RadInformatics/Volume1\\_No1/Radinformatics\\_CTA\\_0208.asp](http://www.imagingcenterinstitute.com/RadInformatics/Volume1_No1/Radinformatics_CTA_0208.asp)
  - [http://209.85.173.104/search?q=cache:4gicroz9cPMJ:images.ctisus.com/cta\\_web/12\\_06/%20AR\\_12\\_06\\_CTA\\_Jacobs.pdf+%22applied+radiology%22+jacobs+CTA&hl=en&ct=clnk%20&cd=4&gl=us&client=safari](http://209.85.173.104/search?q=cache:4gicroz9cPMJ:images.ctisus.com/cta_web/12_06/%20AR_12_06_CTA_Jacobs.pdf+%22applied+radiology%22+jacobs+CTA&hl=en&ct=clnk%20&cd=4&gl=us&client=safari)
  - [http://www.imagingeconomics.com/issues/articles/2004-07\\_10.asp](http://www.imagingeconomics.com/issues/articles/2004-07_10.asp)

10.00mm/div

# Further Reading

- CTA Interpretation Strategies:
  - Ferencik, M. Radiology 2007;243:696-702
  - Saba, et al. J Comput Assist Tomogr. 2007 Sep-Oct;31(5):712-6.
  - Maintz, D. et al. Am. J. Roentgenol. 2002;179:1319-1322
  - Pugliese, F. et al. Radiographics 2006;26:887-904
- OSIRIX:
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