

A Method for Cross-platform Comparison of Reconstruction Kernels in CT

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- 1 Introduction
 - Motivation
 - Aims
- 2 Kernel Extraction Approach
 - Assumptions
 - Overview
- 3 Evaluation
 - Methods
 - Results
 - Conclusions



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- Often need to standardize protocols across dissimilar scanners
- Of the parameters we control (kVp, tube output (mAs/CTDIvol), slice-thickness), *reconstruction kernel* remains problematic



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- Of the parameters we control (kVp, tube output (mAs/CTDIvol), slice-thickness), *reconstruction kernel* remains problematic



- Kernels obey some basic rules. If $G(u)$ is our kernel in the Fourier domain, and u is spatial frequency, then:
 - ▶ $G(u)$ should be real and even,
 - ▶ $G(u) = |u|$ for u near 0, and
 - ▶ $G(u)$ is smooth except at 0 and the Nyquist frequency.
- Kernels are otherwise somewhat “free-form”
 - ▶ Intra-manufacturer variations (name changes, scanner upgrades, etc.)
 - ▶ Inter-manufacturer variations (naming schemes, underlying kernel behavior, etc.)



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Examples

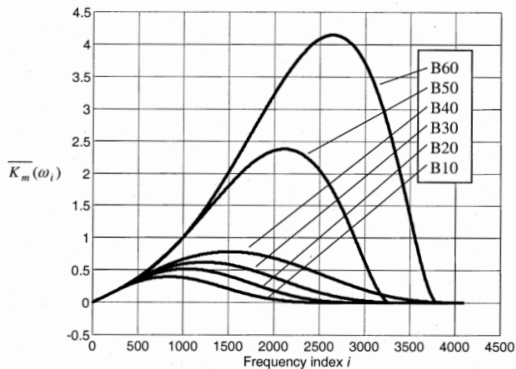


Fig. 2. Fourier transforms $\overline{K_m}(\omega_i)$ of the standard body kernels of the 4-slice CT scanner Siemens SOMATOM Volume Zoom.

Figure : Siemens CT Reconstruction kernel profiles from Volume Zoom. Source: Schaller et al. 2003



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- How can we probe beyond names like “body”, “lung”, “detail”, “B10s”, “H40f”, for some mathematical *information* about the reconstruction kernel?
- Let's develop a method to access reconstruction kernel structure.



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Assumptions

- We can perform **two sets of reconstructions from the same raw data**
 - ▶ a “test”
 - ▶ a “reference”
- Everything (algorithm, preprocessing, slice thickness, etc.) is the same *except* recon kernel.
- For scanner-independence, we **must know** the “reference” kernel profile in the Fourier domain.



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The Full Kernel Extraction Process

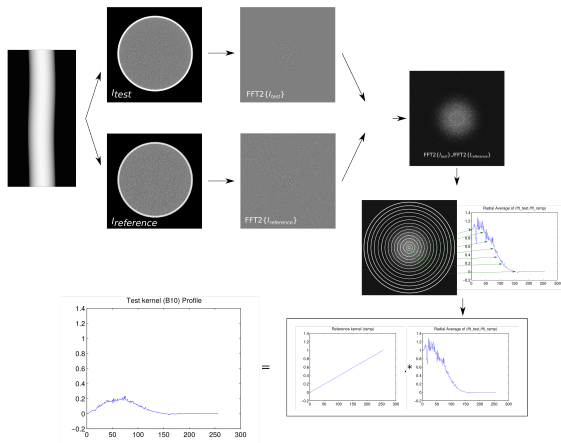


Figure : Flowchart of kernel extraction via proposed method



Step 1: Raw data to Fourier domain image data

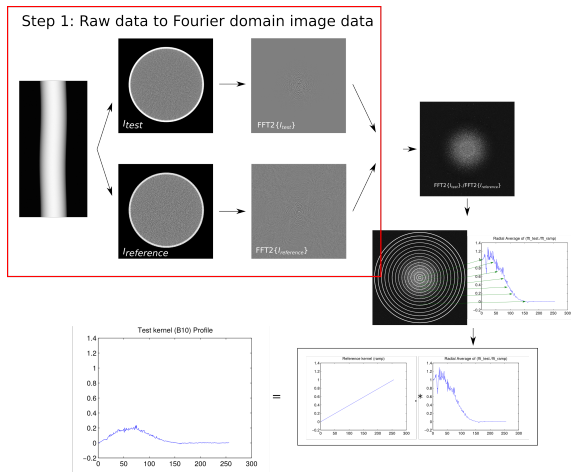
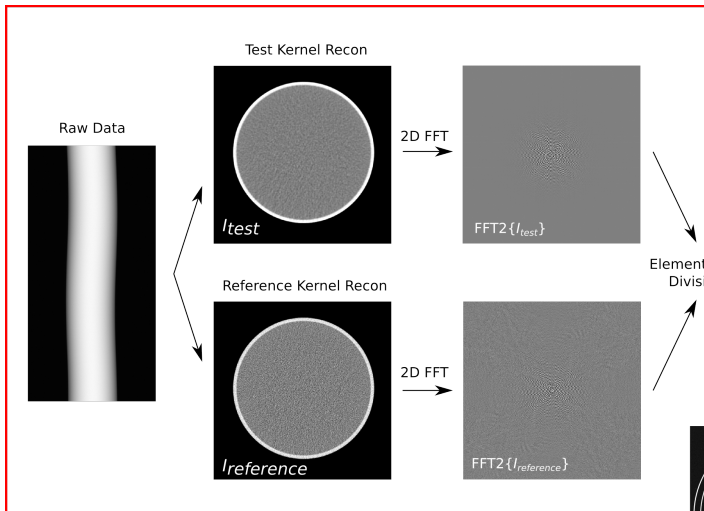


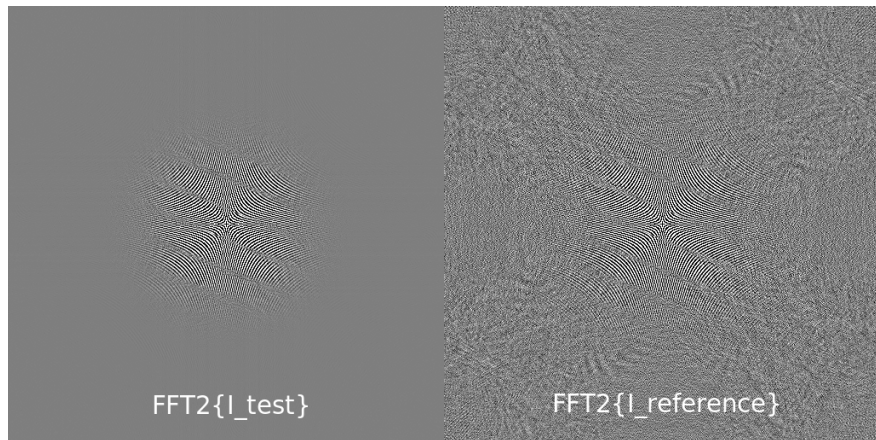
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Raw data to Fourier domain image data



Fourier domain image data - detail



Step 2: Ratio image and radial distribution

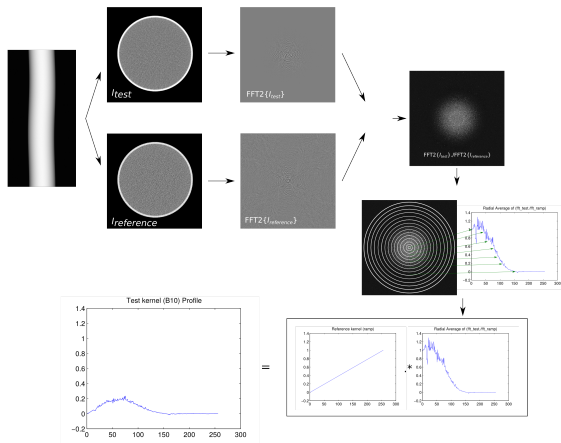


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Step 2: Ratio image and radial distribution

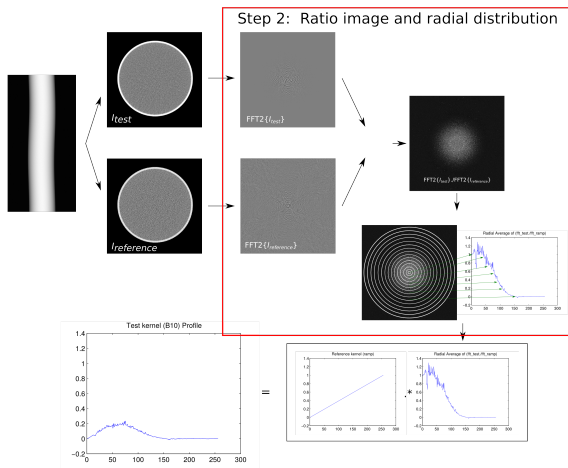
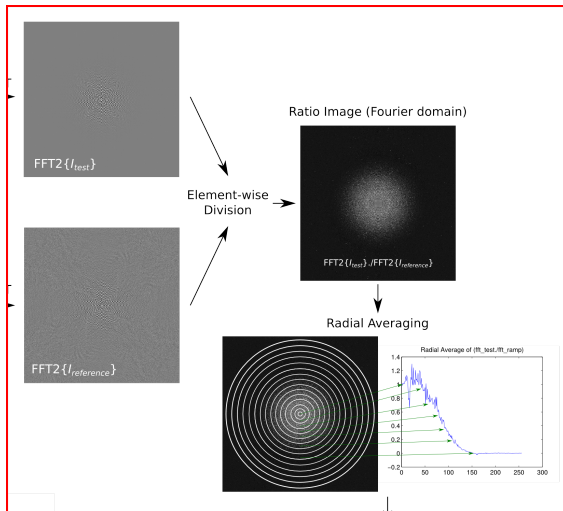


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Ratio image and radial distribution



Step 3: Multiply by reference kernel (if known)

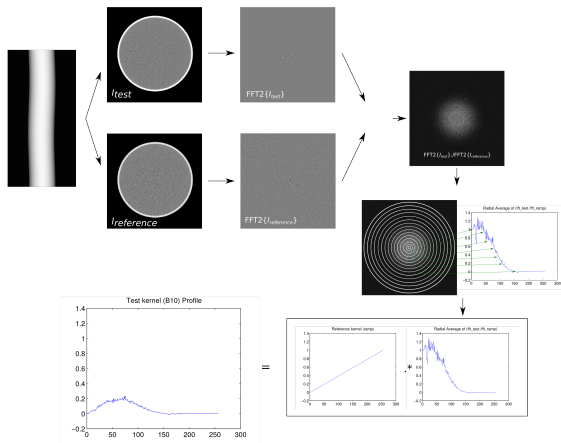


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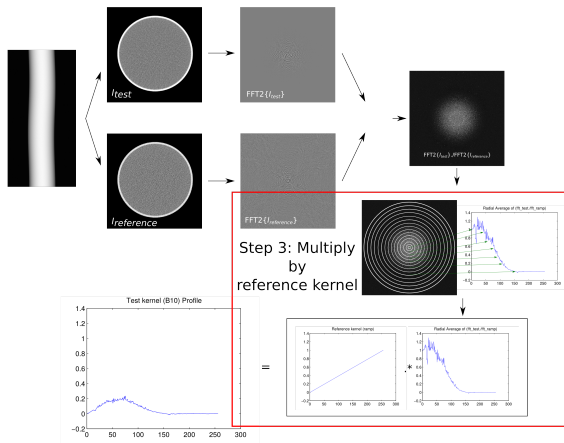
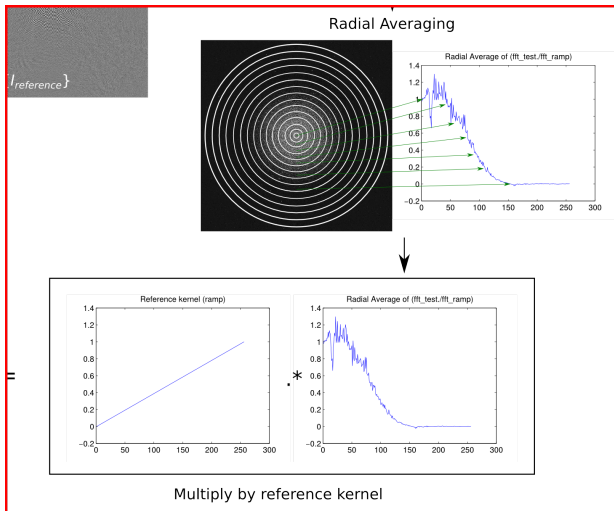


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Multiply by reference kernel



And finally....

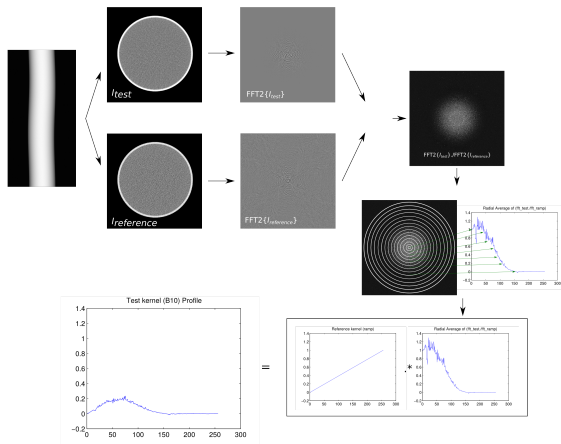


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Step 3 cont.: Arrive at final, absolute kernel profile

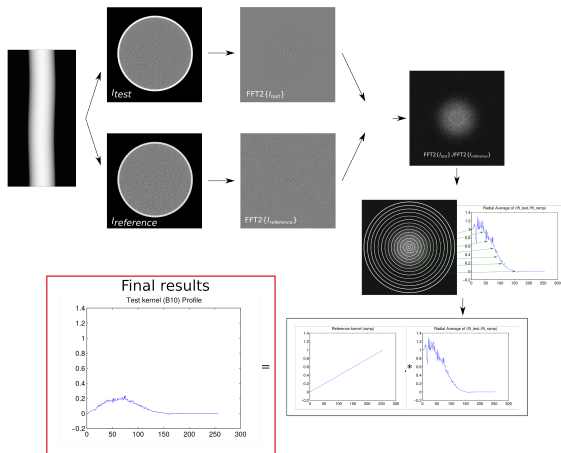
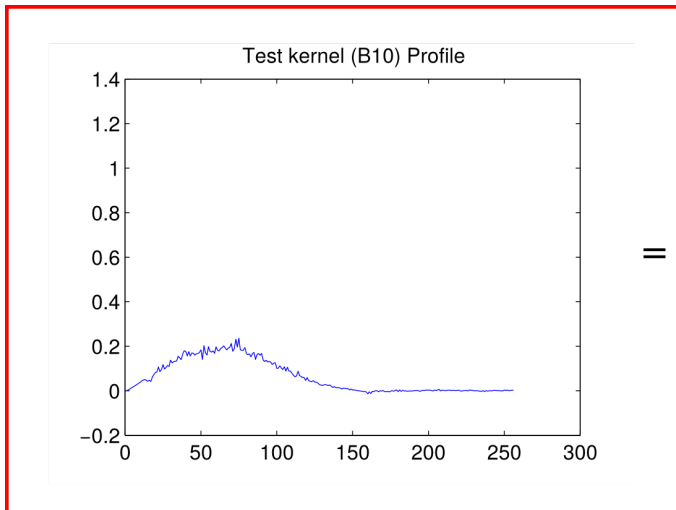


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Final kernel profile



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- Method was tested on data from **Sensation 64** and **Definition AS 64** (Siemens Healthcare, Forchheim, Germany)
- For each scanner, **5 scans** through a **16 cm, centered, water phantom** were acquired with
 - ▶ 1 second rotation time
 - ▶ 64x0.6mm collimation
 - ▶ Z + Phi flying focal spots



Reconstructions

- Each raw data file was reconstructed on the scanner with the following parameters:
 - ▶ Weighted filtered backprojection
 - ▶ Slice thickness and spacing: 0.6 mm
 - ▶ Reconstruction diameter (FOV): 250mm
 - ▶ Kernels: **B10, B20, B30, B40, B50, B60, B70, and B80** (“test” reconstructions)
- In addition, each raw data file was reconstructed using custom software, FreeCT_wFBP¹, using same parameters but with a **ramp kernel** (“reference” reconstructions).

¹http://github.com/FreeCT/FreeCT_wFBP, submitted to MedPhys



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- Each test image (scanner-reconstructed image) and corresponding reference (ramp image) were **analyzed using the outlined method**.
- All profiles for a given scanner and kernel were then averaged together for a final kernel profile.
- Sanity-check using scanner-specific B80 reconstruction as reference (instead of ramp kernel)



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Hypothesis: If the method works...

- We should see the **same kernel profiles between the two scanners.**



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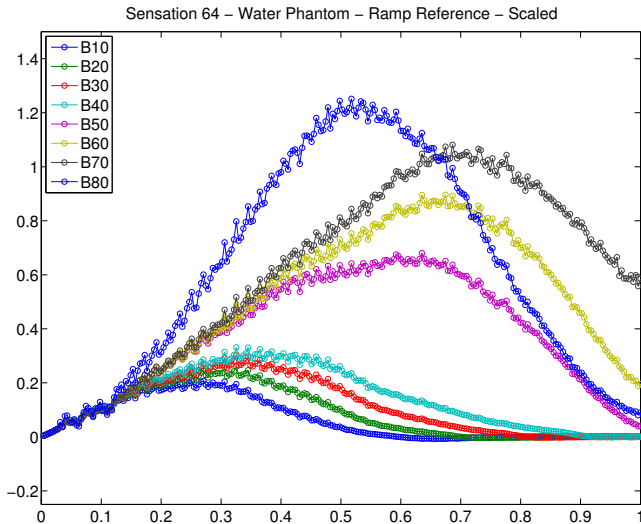
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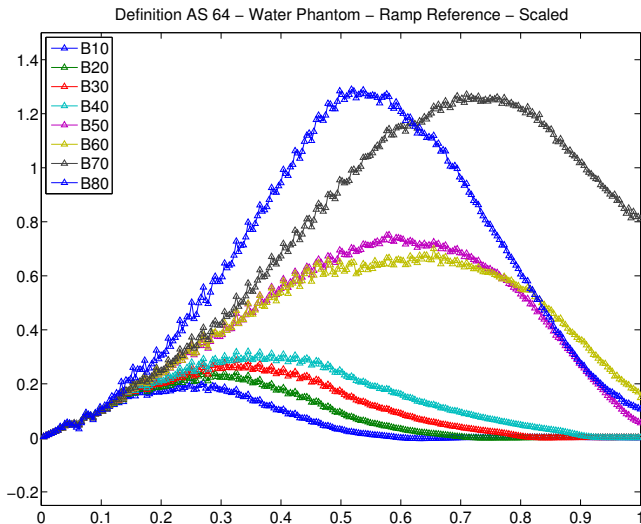
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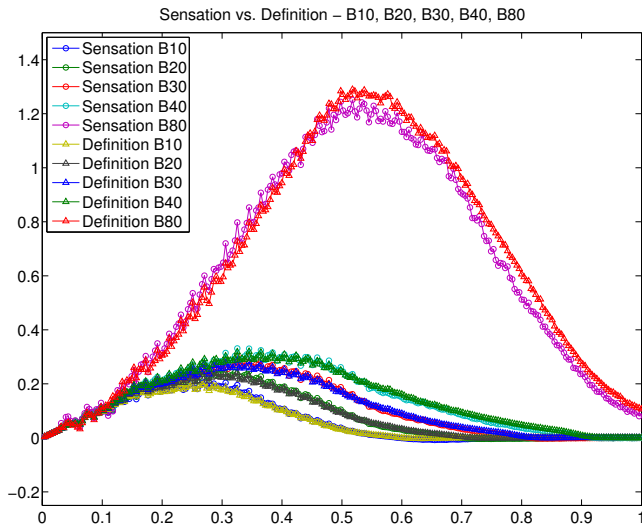
Results: Sensation 64



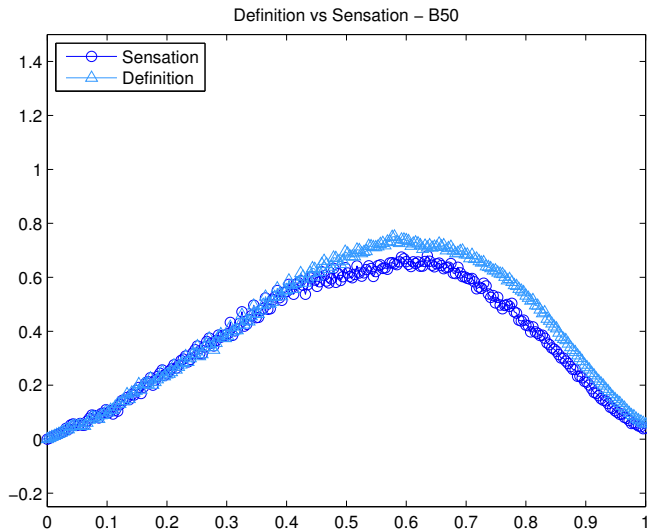
Results: Definition AS 64



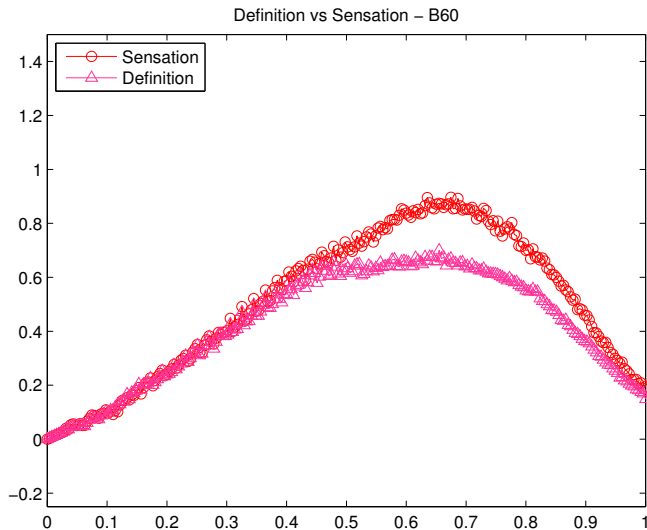
The Good



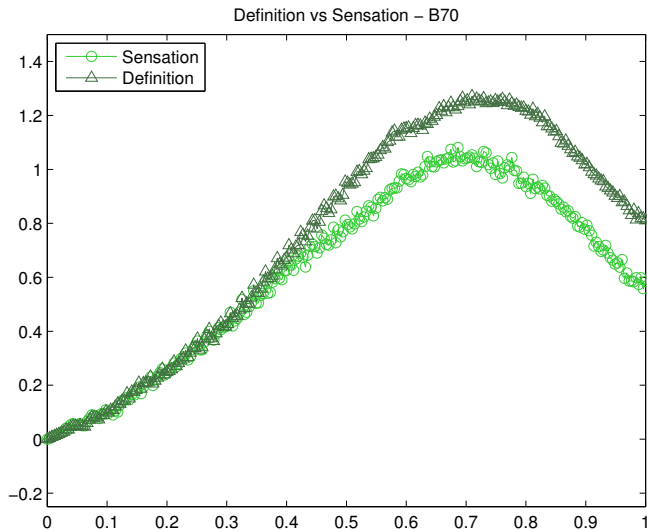
The Not-So-Good - B50



The Not-So-Good - B60



The Not-So-Good - B70

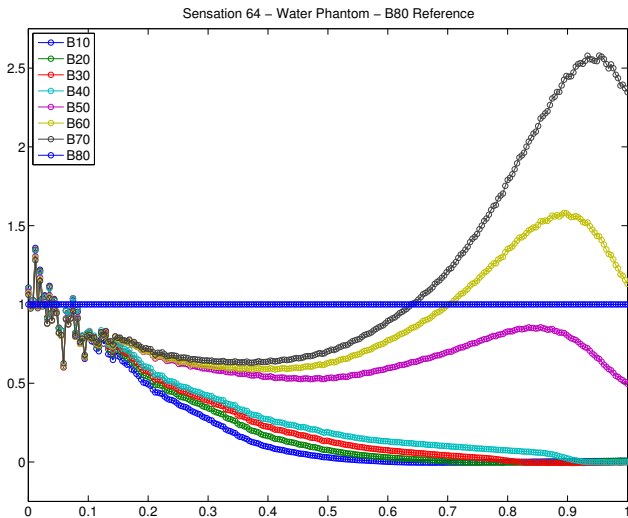


Possible causes

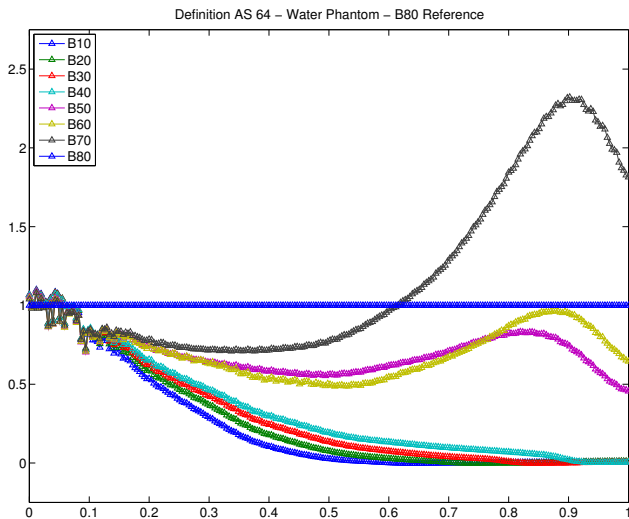
- Ramp kernel reconstruction
- Reconstruction kernels are different between scanners



Results: Sensation 64, B80 Reference



Results: Definition AS 64, B80 Reference



Possible causes

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- A method to extract reconstruction kernel profiles from image data has been presented
- Possible applications include:
 - ▶ Cross-platform protocol standardization (research, clinical trials, etc.)
 - ▶ Reverse engineering



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- Different phantoms
- Scanners from other manufacturers
- Effects of FOV, slice thickness, noise, etc.
- Does matching kernels necessarily match other image performance metrics (MTF, NPS, etc.)?
- Utilizing method for quantitative imaging



Finally...

Thank you for your interest and any questions!

