

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

J. Hoffman¹ M. McNitt-Gray¹

¹Biomedical Physics Interdepartmental Program
University of California Los Angeles

Radiological Society of North America Annual Meeting, 2015

Disclosures

- John Hoffman:
 - ▶ Part-time intern, Toshiba Medical Research Institute, USA, Inc.
- Michal McNitt-Gray:
 - ▶ Institutional research agreement, Siemens Healthcare
 - ▶ Past recipient, research grant support, Siemens Healthcare
 - ▶ Consultant, Toshiba America Medical Systems
 - ▶ Consultant, Samsung Electronics



Outline

1 Introduction

- Motivation
- Aims

2 Kernel Extraction Approach

- Assumptions
- Overview

3 Evaluation

- Methods
- Results
- Conclusions



Outline

1 Introduction

- Motivation
- Aims

2 Kernel Extraction Approach

- Assumptions
- Overview

3 Evaluation

- Methods
- Results
- Conclusions



Motivation

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

Motivation

- Often need to standardize protocols across dissimilar scanners
- Of the parameters we control (kVp, tube output ($\text{mAs}/\text{CTDI}_{\text{vol}}$), slice-thickness), *reconstruction kernel* remains problematic

Kernels

- 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.)



Kernels

- 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.)



- 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.)



- 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.)



- 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.)



Examples

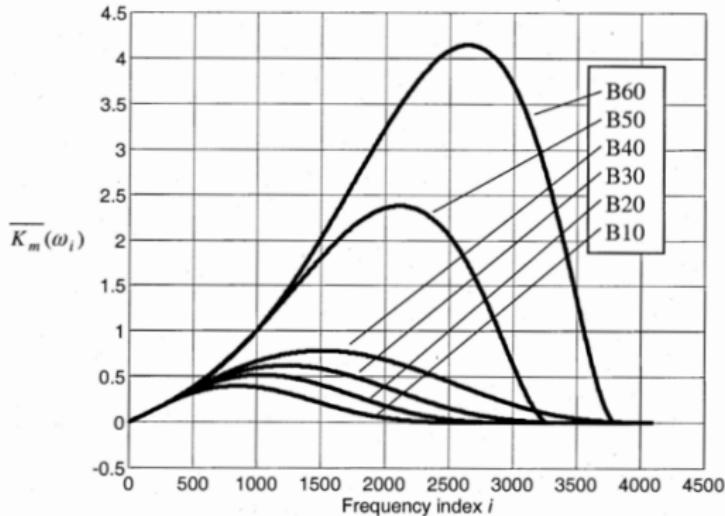


Fig. 2. Fourier transforms $\bar{K}_m(\omega_t)$ of the standard body kernels of the 4-slice CT scanner Siemens SOMATOM Volume Zoom.

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



Outline

1 Introduction

- Motivation
- Aims

2 Kernel Extraction Approach

- Assumptions
- Overview

3 Evaluation

- Methods
- Results
- Conclusions



- 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.

- 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.

Outline

1 Introduction

- Motivation
- Aims

2 Kernel Extraction Approach

- Assumptions
- Overview

3 Evaluation

- Methods
- Results
- Conclusions



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.**



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**.



Outline

1 Introduction

- Motivation
- Aims

2 Kernel Extraction Approach

- Assumptions
- Overview

3 Evaluation

- Methods
- Results
- Conclusions



The Full Kernel Extraction Process

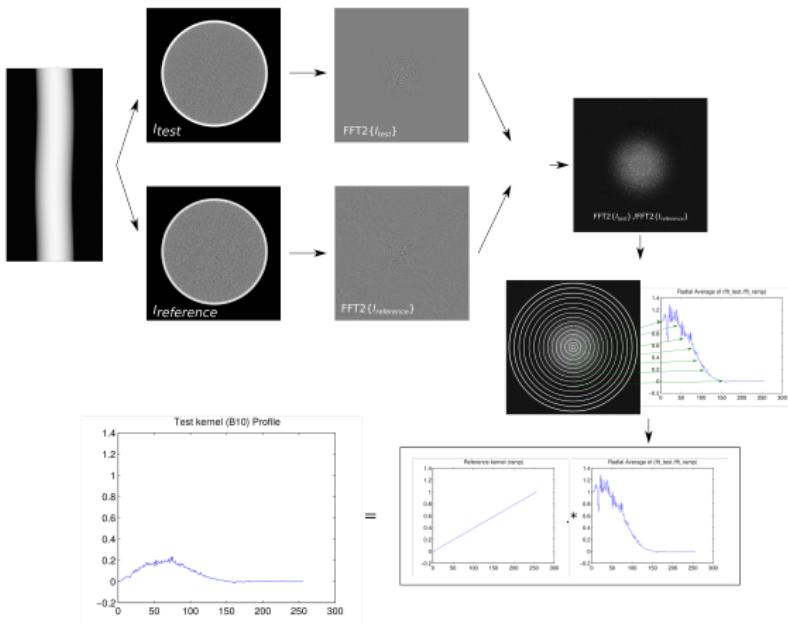


Figure : Flowchart of kernel extraction via proposed method



Step 1: Raw data to Fourier domain image data

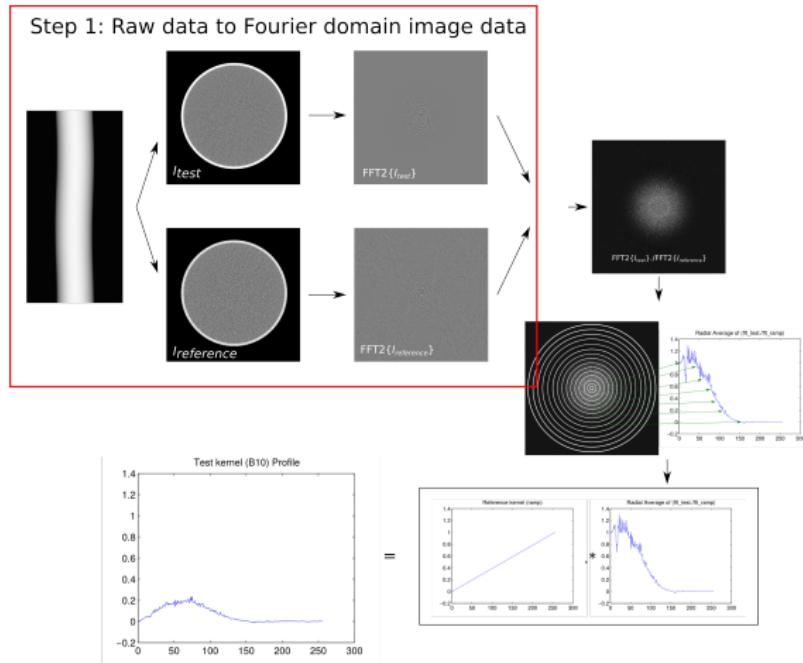
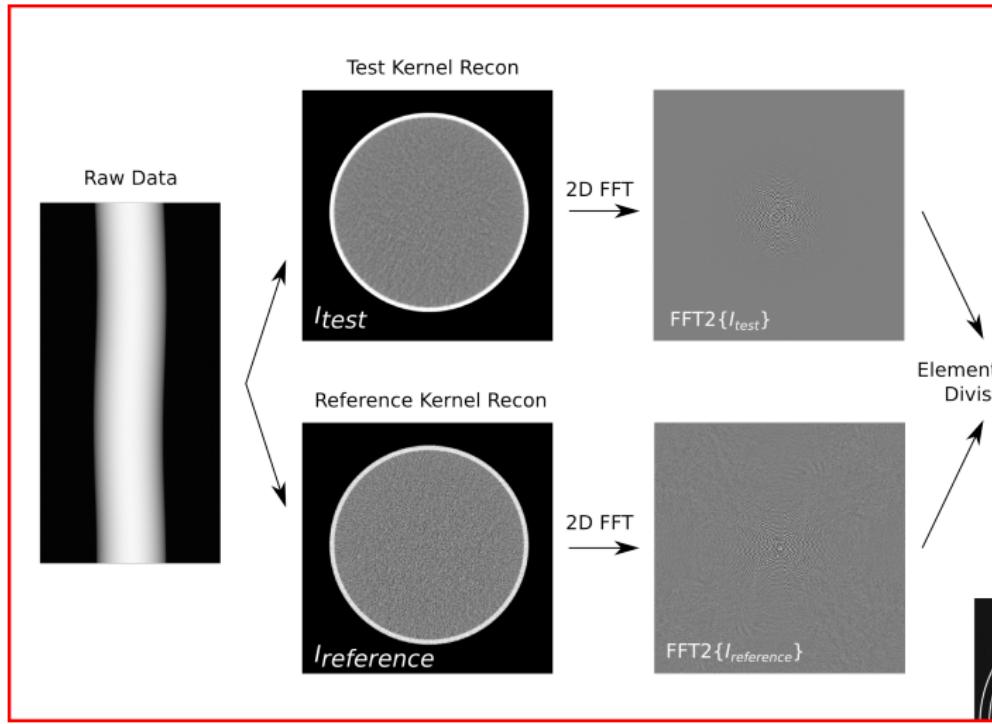


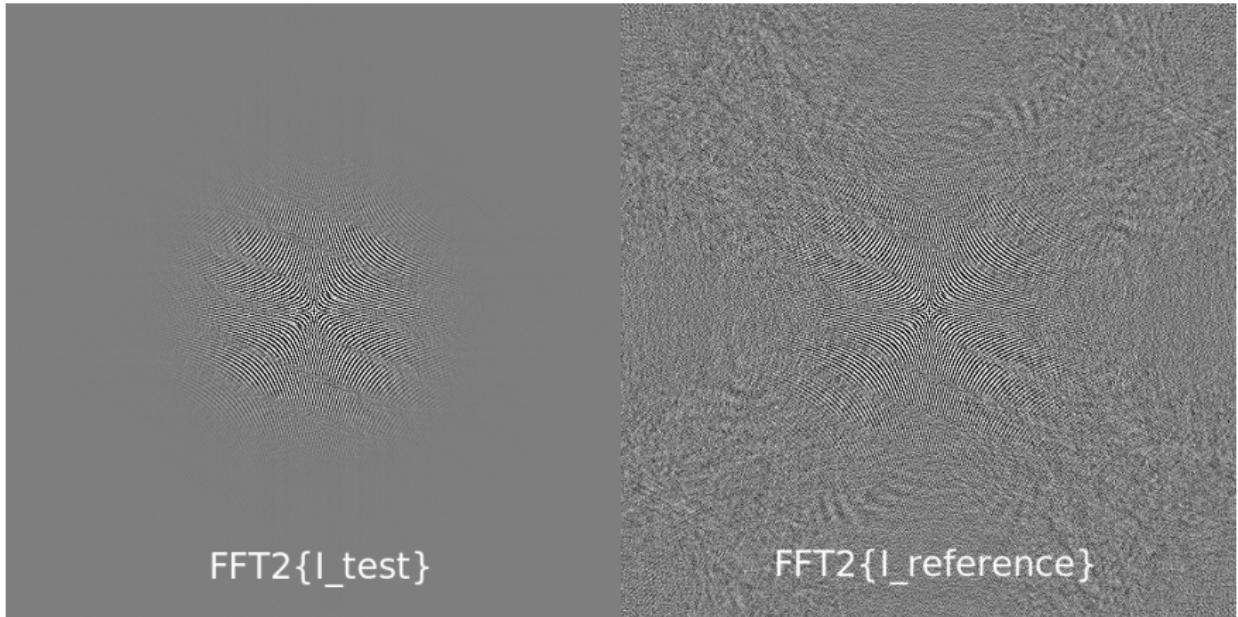
Figure : Flowchart of kernel extraction via proposed method



Raw data to Fourier domain image data



Fourier domain image data - detail



Step 2: Ratio image and radial distribution

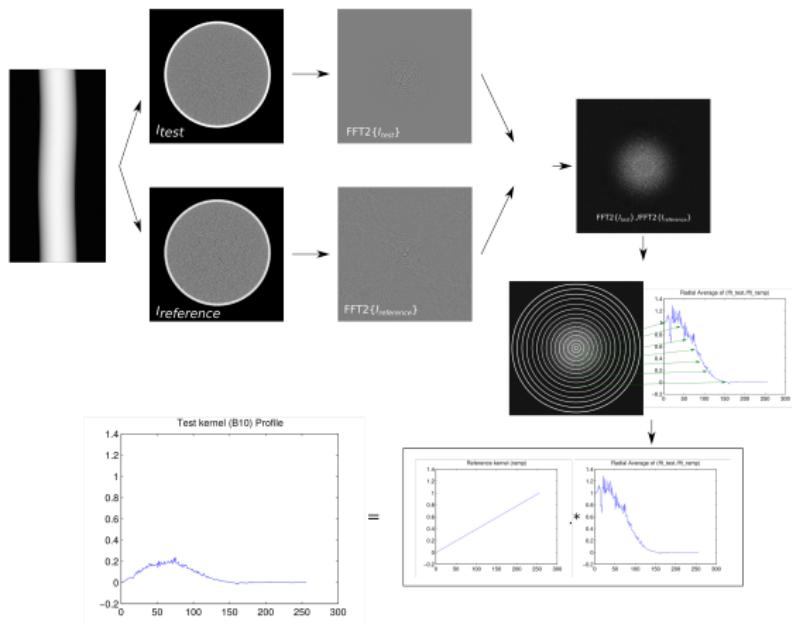


Figure : Flowchart of kernel extraction via proposed method



Step 2: Ratio image and radial distribution

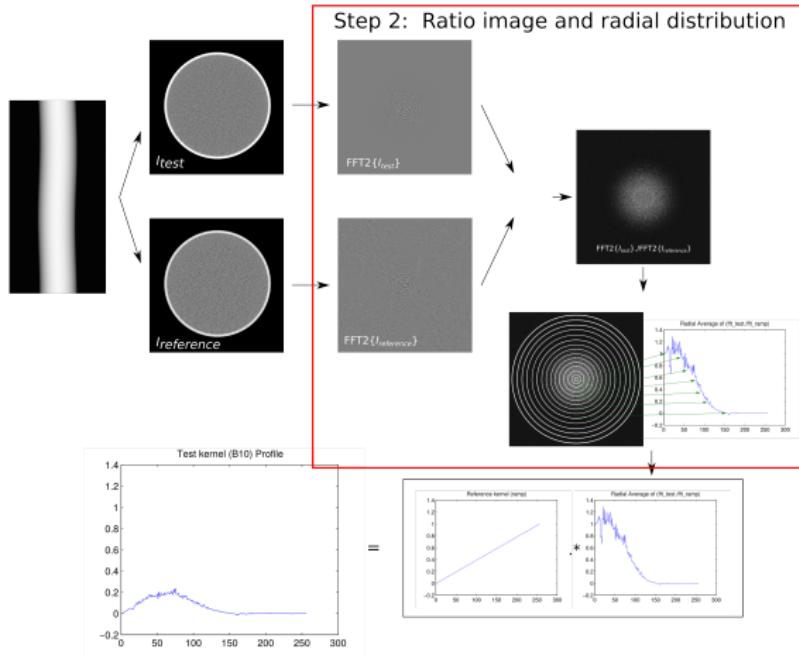
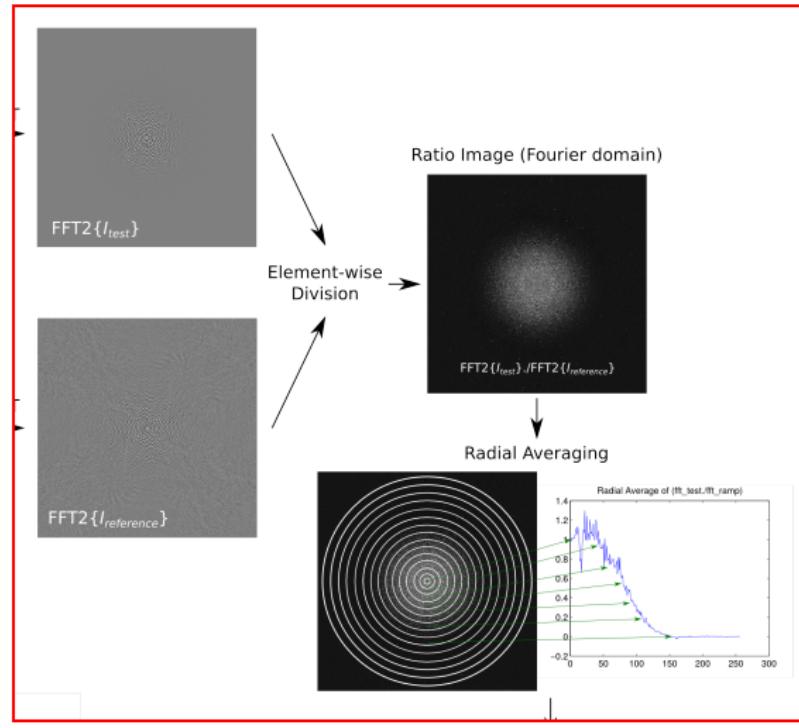


Figure : Flowchart of kernel extraction via proposed method



Ratio image and radial distribution



Step 3: Multiply by reference kernel (if known)

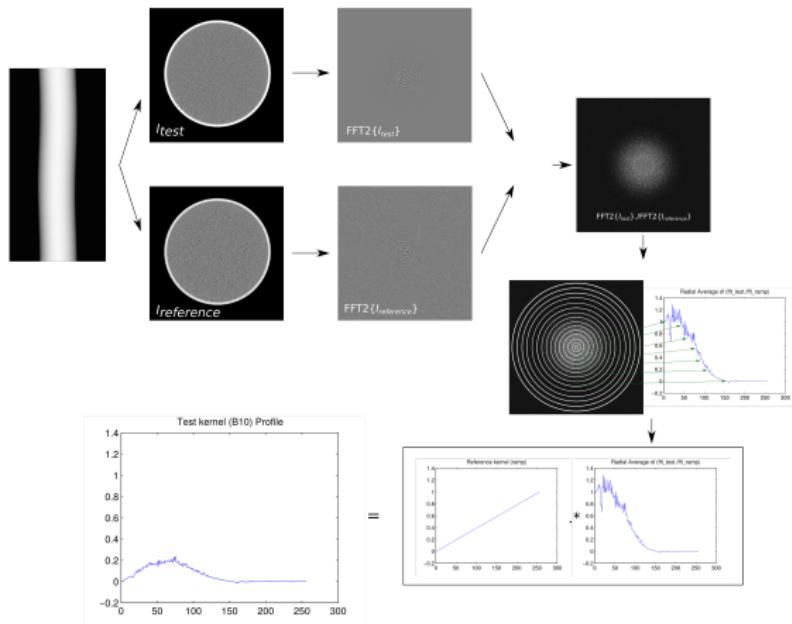


Figure : Flowchart of kernel extraction via proposed method



Step 3: Multiply by reference kernel (if known)

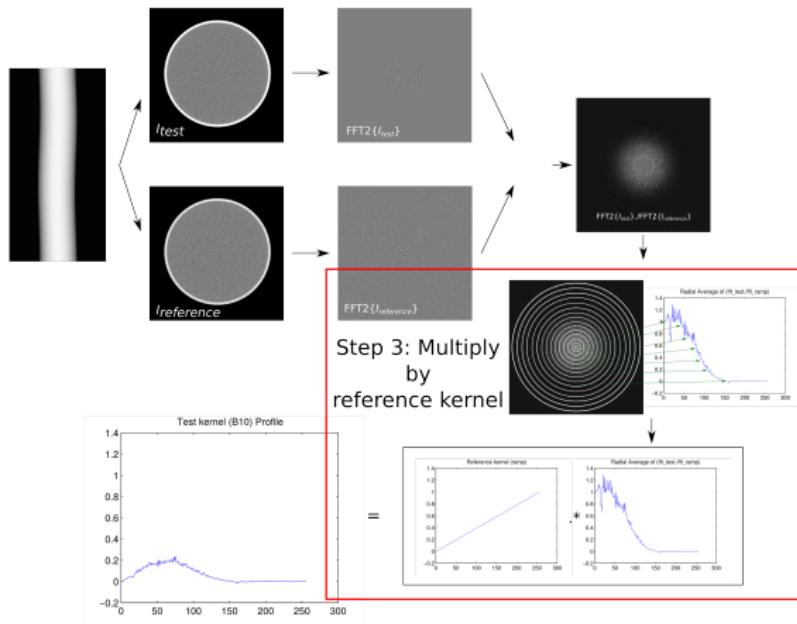
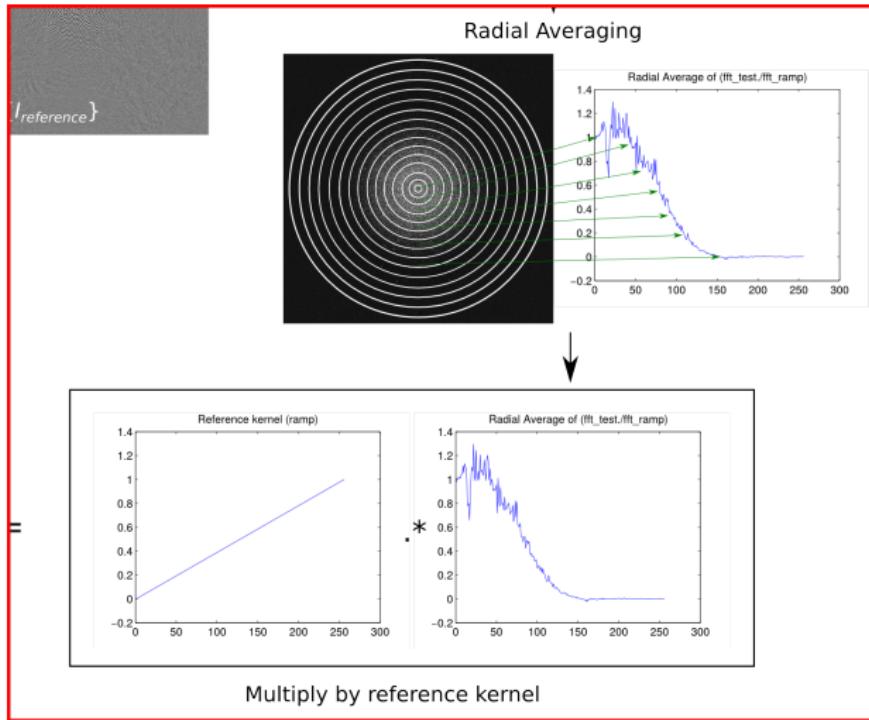


Figure : Flowchart of kernel extraction via proposed method



Multiply by reference kernel



And finally....

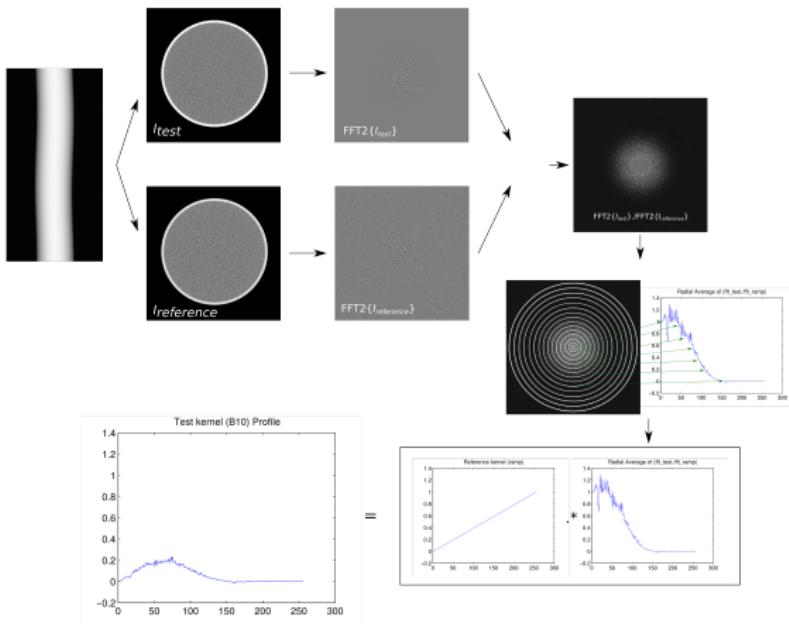


Figure : Flowchart of kernel extraction via proposed method



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

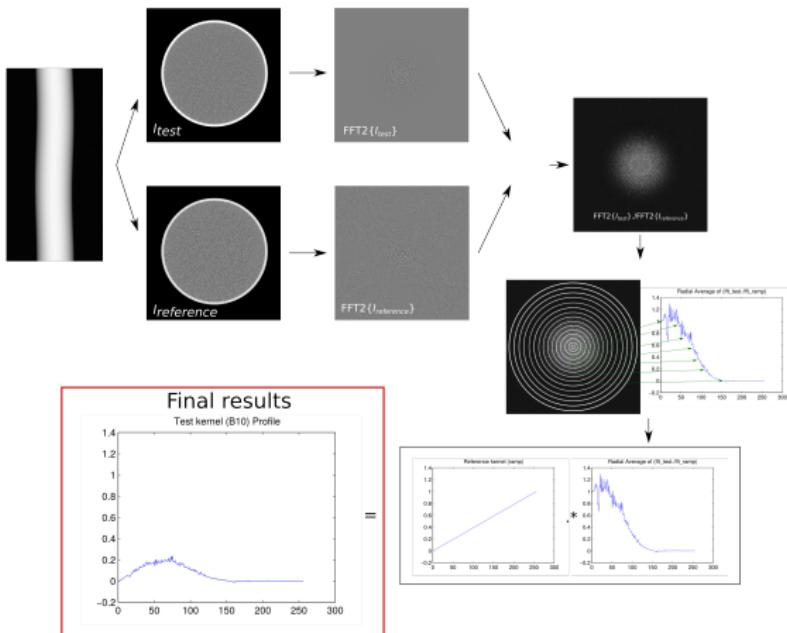
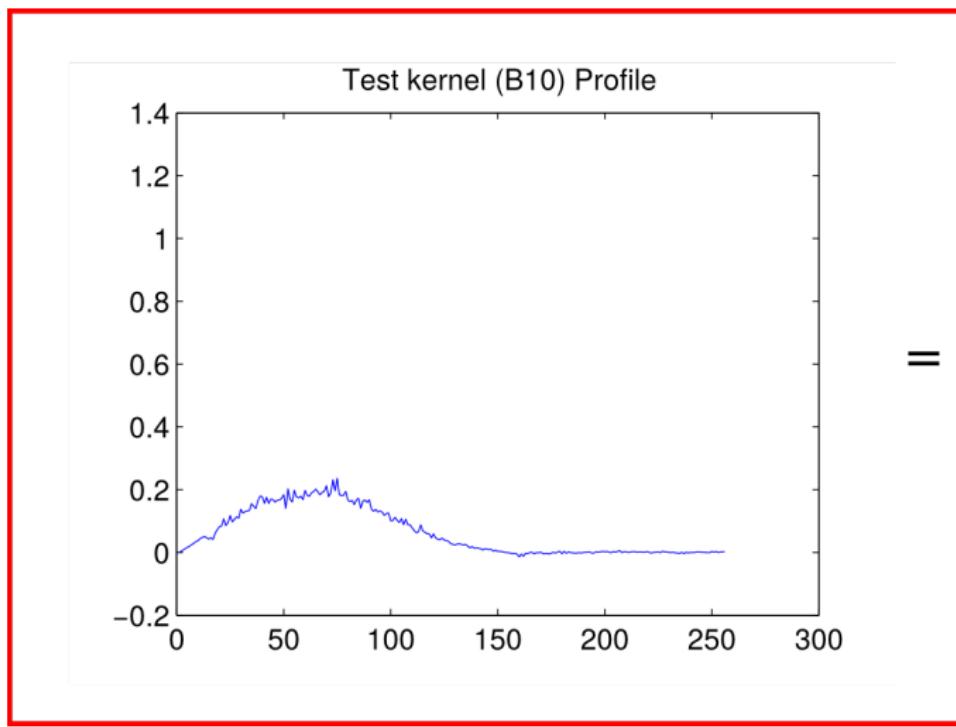


Figure : Flowchart of kernel extraction via proposed method



Final kernel profile



Outline

1 Introduction

- Motivation
- Aims

2 Kernel Extraction Approach

- Assumptions
- Overview

3 Evaluation

- Methods
- Results
- Conclusions



- 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).

¹<http://github.com/FreeCT/FreeCT> wFBP, submitted to *MedPhys*

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).

¹<http://github.com/FreeCT/FreeCT> wFBP, submitted to MedPhys



Analysis

- 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)

Analysis

- 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)

Analysis

- 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)

Hypothesis: If the method works...

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



Outline

1 Introduction

- Motivation
- Aims

2 Kernel Extraction Approach

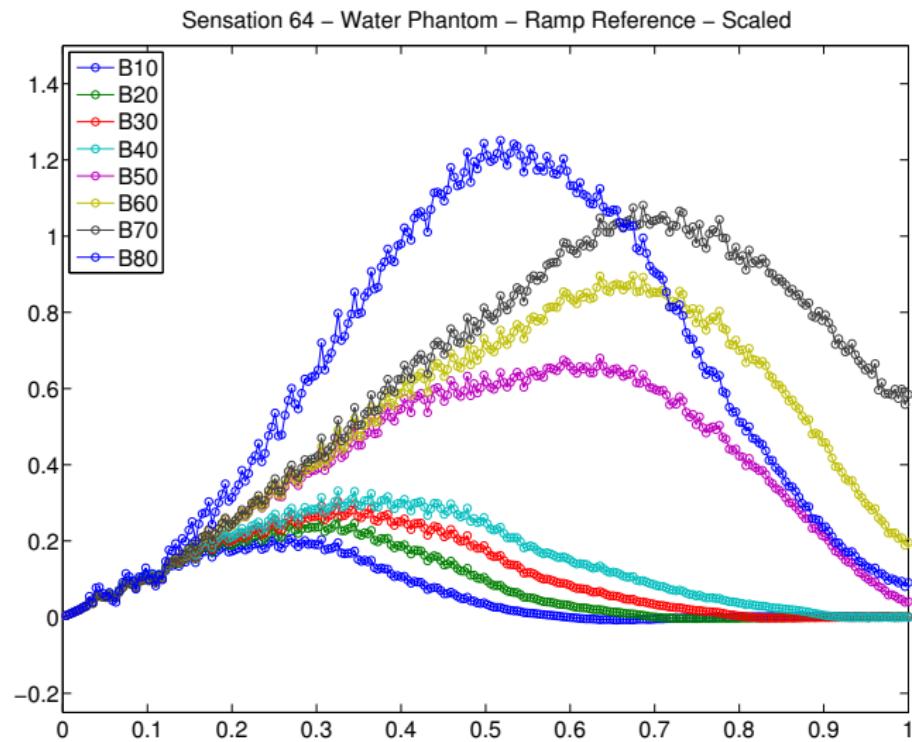
- Assumptions
- Overview

3 Evaluation

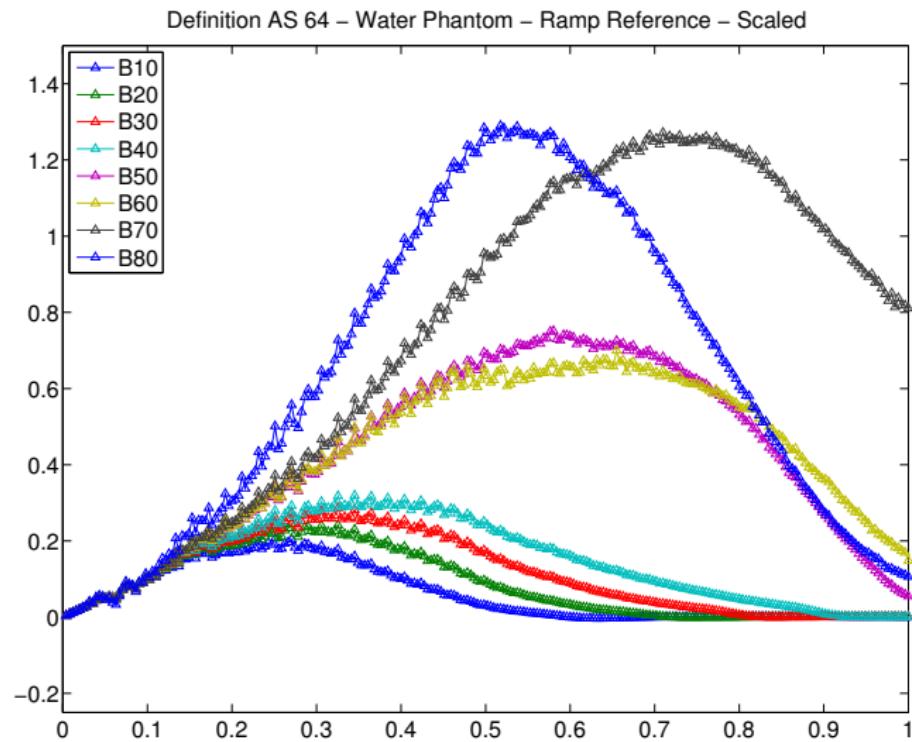
- Methods
- Results
- Conclusions



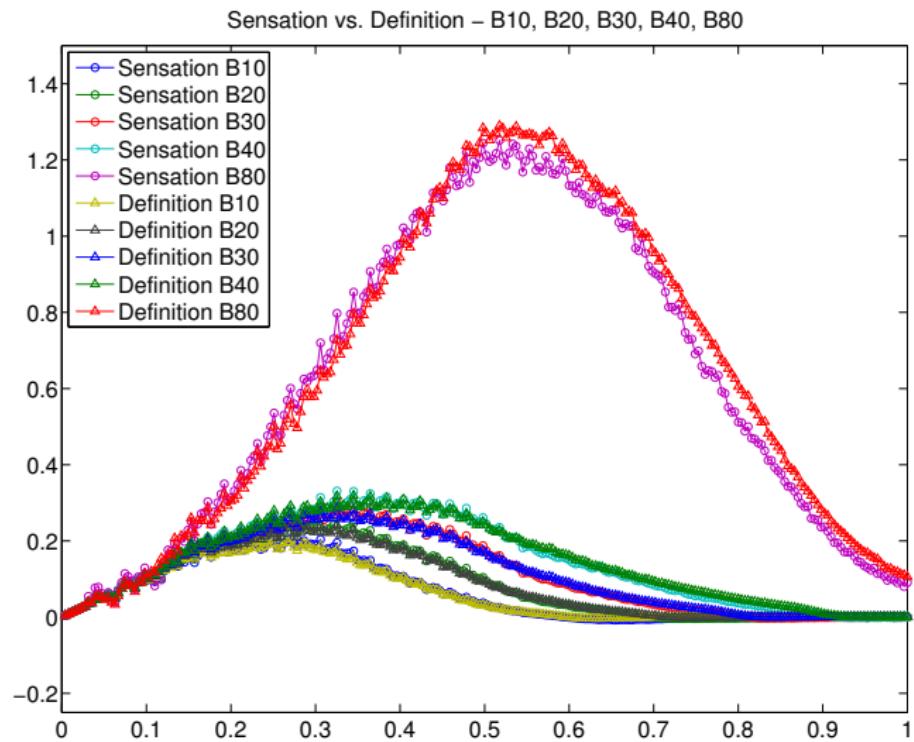
Results: Sensation 64



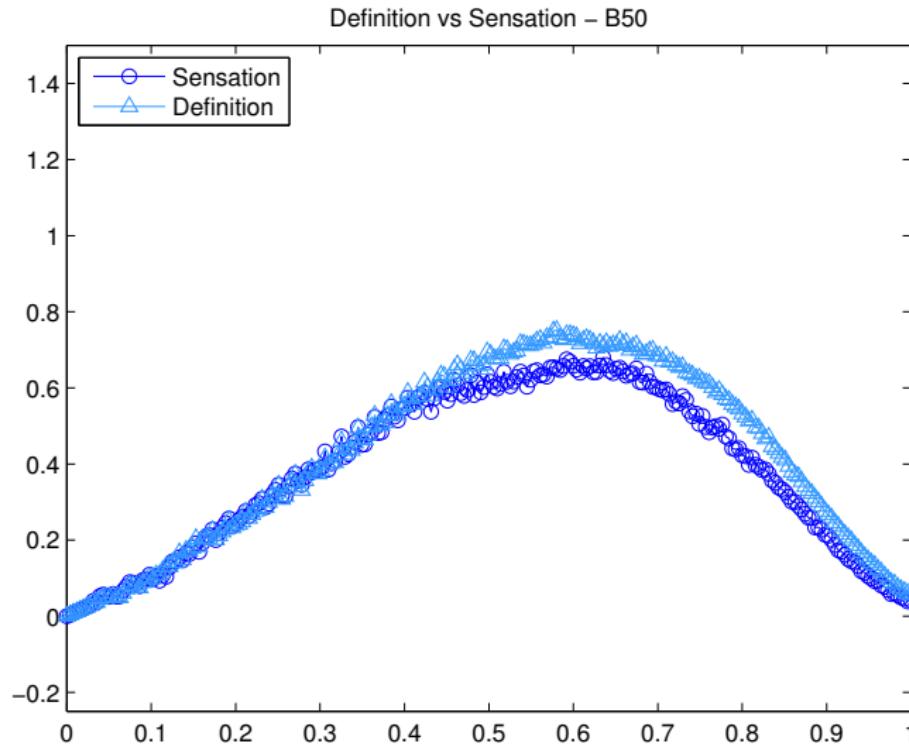
Results: Definition AS 64



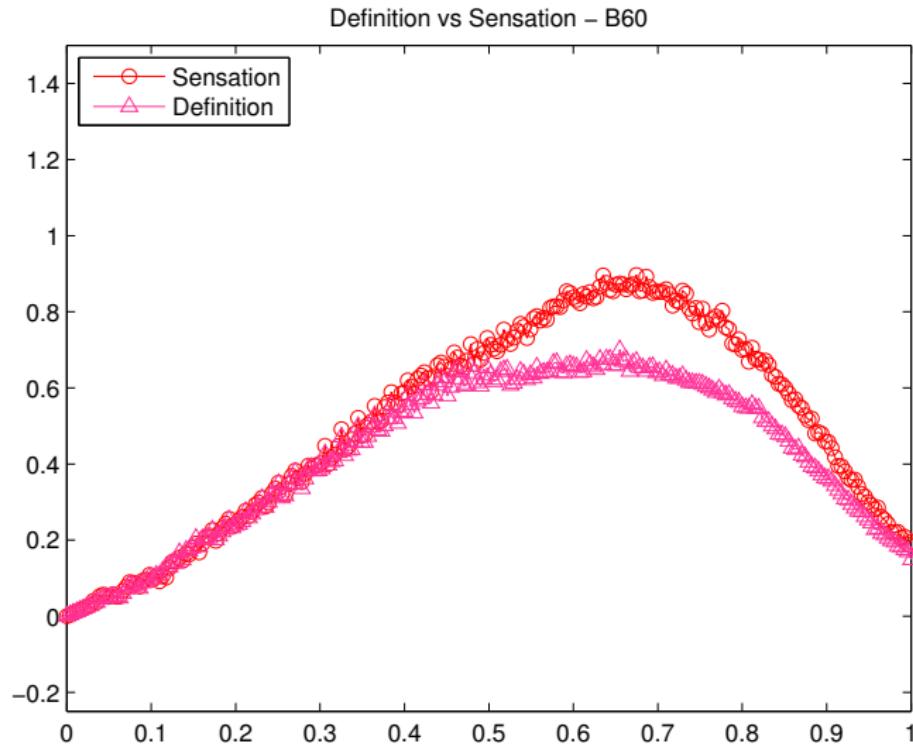
The Good



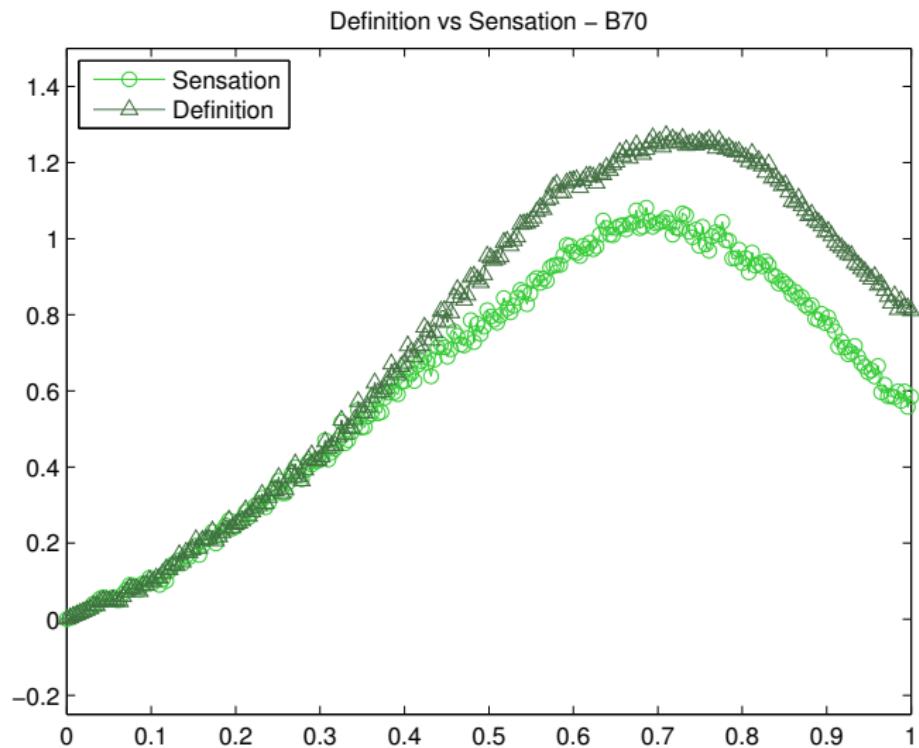
The Not-So-Good - B50



The Not-So-Good - B60



The Not-So-Good - B70

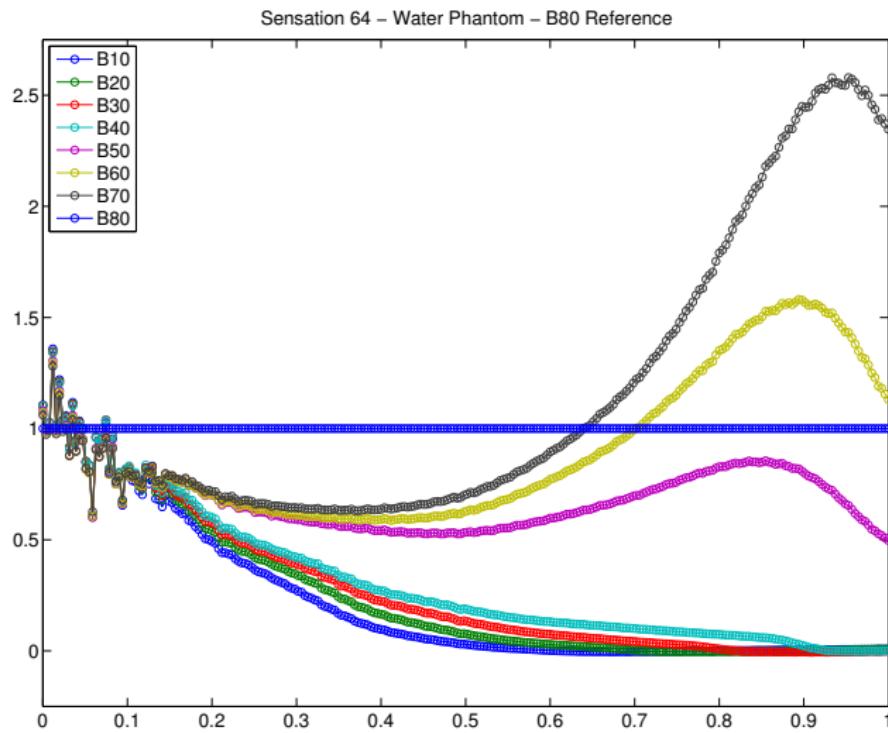


Possible causes

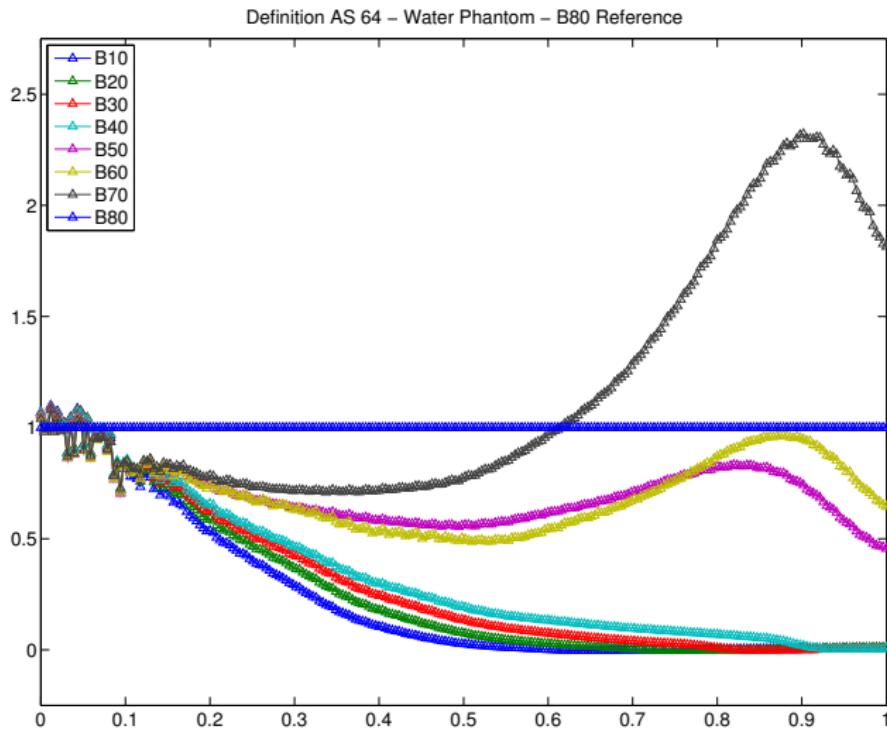
- Ramp kernel reconstruction
- Reconstruction kernels are different between scanners



Results: Sensation 64, B80 Reference



Results: Definition AS 64, B80 Reference



Possible causes

- Ramp kernel reconstruction
- *Reconstruction kernels are different between scanners*

Outline

1 Introduction

- Motivation
- Aims

2 Kernel Extraction Approach

- Assumptions
- Overview

3 Evaluation

- Methods
- Results
- Conclusions



Conclusions

- 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



Conclusions

- 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



Further Work

- 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!

