



Interference & Noise Robust Image Reconstruction for Magnetic Particle Imaging

Student: Berfin Kavşut

Project Advisor: Emine Ülkü Sarıtaş

Background



- MPI
- X-Space Reconstruction
- Partial field-of-view Center Imaging (PCI)
 - PCI
 - Lumped PCI

X-Space Reconstruction

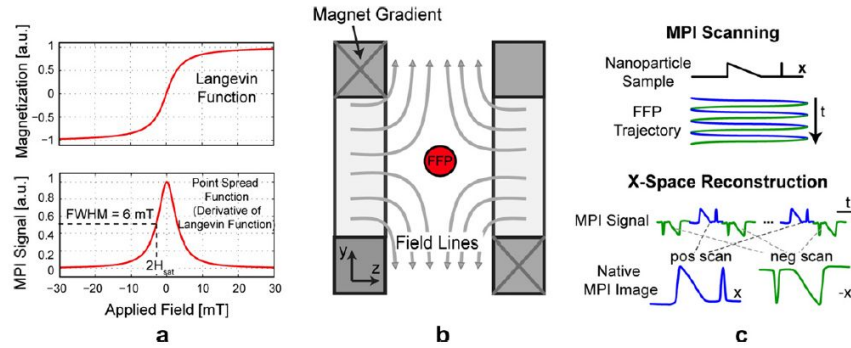


Fig. 1: Principles of x-space MPI imaging. (a) SPIO magnetization characterized by a Langevin function, and Point Spread Function (PSF). (b) Selection fields created by permanent magnets, and Field Free Point (FFP). (c) FFP is moved rapidly with excitation field. In x-space reconstruction, MPI signal is gridded to the instantaneous position of the FFP. [1]

- Speed Compensation

$$\hat{\rho}(x_s(t)) = \rho(x) * h(x)|_{x=x_s(t)}$$

$$s(t) = \gamma \dot{x}_s(t) \hat{\rho}(x_s(t))$$

$$\hat{\rho}(x_s(t)) = \frac{s(t)}{\gamma \dot{x}_s(t)}$$

- Fourier Series of MPI Signal

$$s(t) = \sum_{n=1}^{\infty} S_n \sin(2\pi n f_0 t)$$

- Image Basis of MPI Image

$$\begin{aligned} \hat{\rho}(x) &= \left. \frac{s(t)}{\gamma \dot{x}_s(t)} \right|_{t=\frac{1}{(2\pi f_0)} \arccos\left(\frac{2x}{W}\right)} \\ &= \sum_{n=1}^{\infty} S_n \frac{\sin(2\pi n f_0 t)}{\sin(2\pi f_0 t)} \bigg|_{t=\frac{1}{(2\pi f_0)} \arccos\left(\frac{2x}{W}\right)} \\ &= \sum_{n=1}^{\infty} S_n \frac{\sin\left(\pi \arccos\left(\frac{2x}{W}\right)\right)}{\sin\left(\arccos\left(\frac{2x}{W}\right)\right)} \\ &= \sum_{n=1}^{\infty} S_n U_{n-1}\left(\frac{2x}{W}\right) \end{aligned}$$

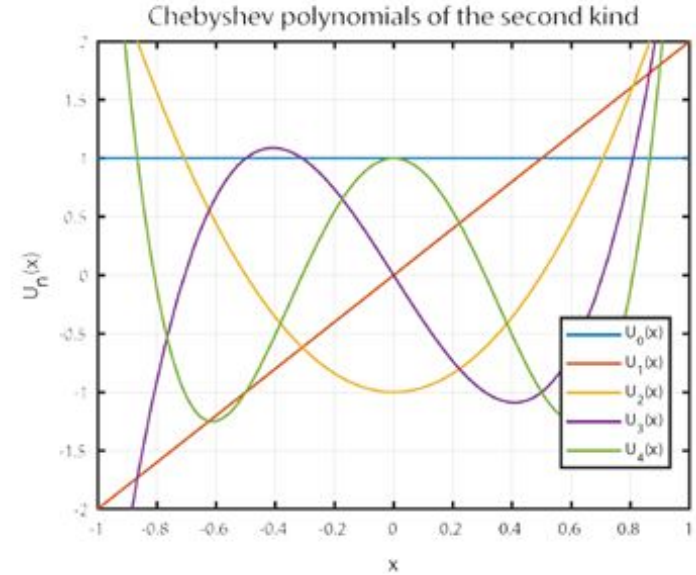


Fig. 2: Chebyshev Polynomials of the Second Kind. [2]

- DC loss in MPI Image

$$\hat{\rho}_{lost}(x) = \alpha S_1 U_0 \left(\frac{2x}{W} \right) = \alpha S_1$$

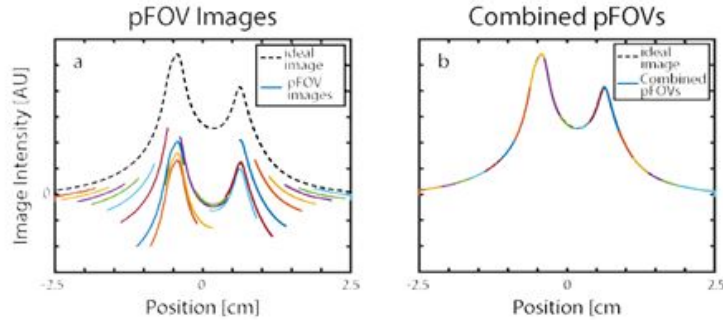


Fig. 3: Standard x-space reconstruction. (a) pFOV images having different DC losses due to high-pass filtering (b) Combining pFOV images by enforcing the smoothness and non-negativity. [2]

- FFP is moved to scan field-of-view (FOV). Due to safety limitations, FOV is divided into partial FOVs (pFOV).
- pFOVs are combined to obtain FOV after speed compensation in x-space reconstruction.
- First harmonic signal is lost due to high-pass filtering MPI signal to get rid of direct feedthrough signal.

Partial FOV Center Imaging (PCI)



- Partial FOV Centering Imaging (PCI) is a novel robust x-space image reconstruction for MPI developed with the motivation of eliminating non-ideal effects on signal. There are two proposed methods: PCI and Lumped PCI.
- PCI and Lumped-PCI methods offer a trade-off between interference and noise robustness. PCI is robust against interference and sensitive to noise, while Lumped PCI is robust against noise and sensitive to interference.

PCI

→ PCI samples MPI signal at the centers of pFOVs and locates them closely by overlapping pFOVs.

$$x_s(t)|_{t=t_{0j}} = x_{0j}, \text{ for } j = 1, \dots, N$$

$$\begin{aligned}\hat{\rho}_0(x_{0j}) &= s(t)|_{t=t_{0j}} \\ &= \alpha \dot{x}_s(t_{0j}) \hat{\rho}(x_{0j}) \\ &= \beta_0 \hat{\rho}(x_{0j}) \text{ for } j = 1, \dots, N\end{aligned}$$

→ Then, applies deconvolution to recover DC loss.

$$\tilde{\rho}_0(x) = \hat{\rho}(x) * h_0(x)$$

$$h_0(x) = \beta_0 \left(\delta(x) - \frac{4}{\pi W} \sqrt{1 - \left(\frac{2x}{W}\right)^2} \right)$$

$$\hat{\rho}(x) = \tilde{\rho}_0(x) *^{-1} h_0(x)$$

Lumped PCI



→ Samples at each position on pFOVs are assigned to their centers.

$$x_s(t)|_{t=t_{kj}} = x_{kj}, \text{ for } j = 1, \dots, N \text{ and } k = -K, \dots, K$$

$$\hat{\rho}_k(x_{oj}) = \beta_k \hat{\rho}(x_{kj})$$

Lumped PCI

- MPI image can be obtained from each raw image.

$$h_k(x) = \beta_k \left(\delta \left(x - (x_{0j} - x_{kj}) \right) - \frac{4}{\pi W} \sqrt{1 - \left(\frac{2x}{W} \right)^2} \right)$$

$$\hat{\rho}(x) = \tilde{\rho}_k(x) *^{-1} h_k(x)$$

- Raw Lumped-PCI image is the summation of all raw images for each position.

$$\tilde{\rho}_{lum}(x) = \sum_{k=-K}^K \tilde{\rho}_k(x)$$

$$h_{lum}(x) = \sum_{k=-K}^K h_k(x)$$

$$\hat{\rho}(x) = \tilde{\rho}_{lum}(x) *^{-1} h_{lum}(x)$$

Deconvolution Kernels

- **Modified Lumped PCI Equations**

$$\tilde{\rho}_{lum}(x) = \sum_{k=-K}^K \alpha_k \tilde{\rho}_k(x)$$

$$\begin{aligned} h_{lum}(x) &= \sum_{k=-K}^K \alpha_k h_k(x) \\ &= \sum_{k=-K}^K \alpha_k \beta_k \delta(x - (x_{0j} - x_{kj})) \end{aligned}$$

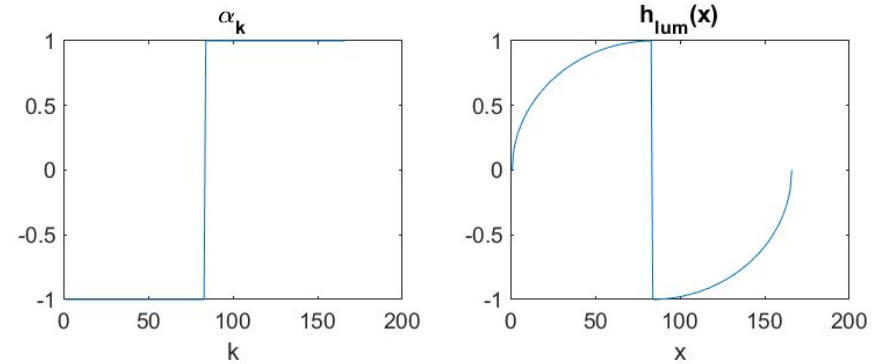


Fig.4 : Deconvolution Kernels for Lumped-PCI

Results

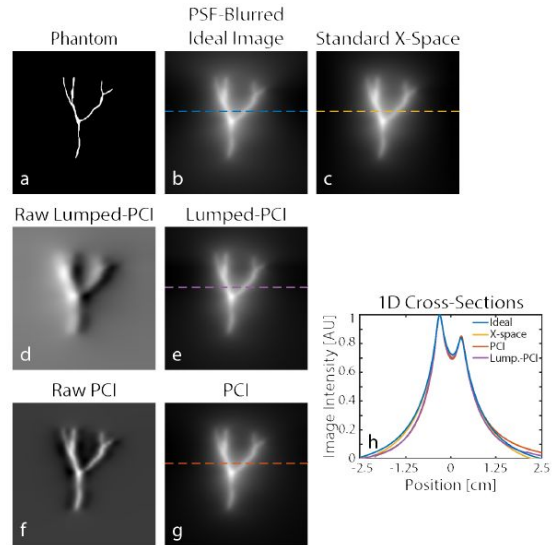


Fig. 5: Image Results for Standard X-Space Reconstruction, Lumped PCI and PCI. [3]

Noise



Noise

- **Signal-to-Noise Ratio (SNR) Definition in Simulations**

$$SNR = 20 \log_{10} \left(\frac{\max_t |s(t)|}{\sigma} \right)$$

- White Gaussian noise
- Noise robustness difference between PCI and Lumped-PCI

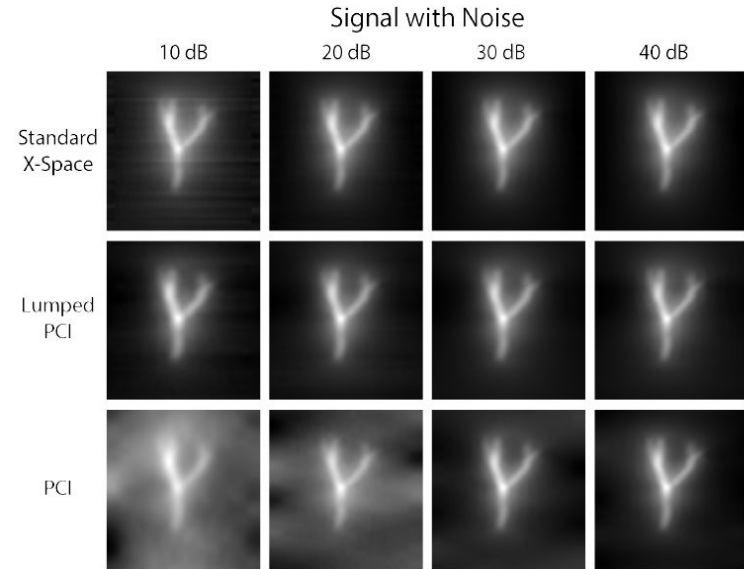


Fig. 6: Comparison of Standard X-Space Reconstruction, Lumped PCI and PCI Results at Different SNR Levels. [3]

Noise Analysis

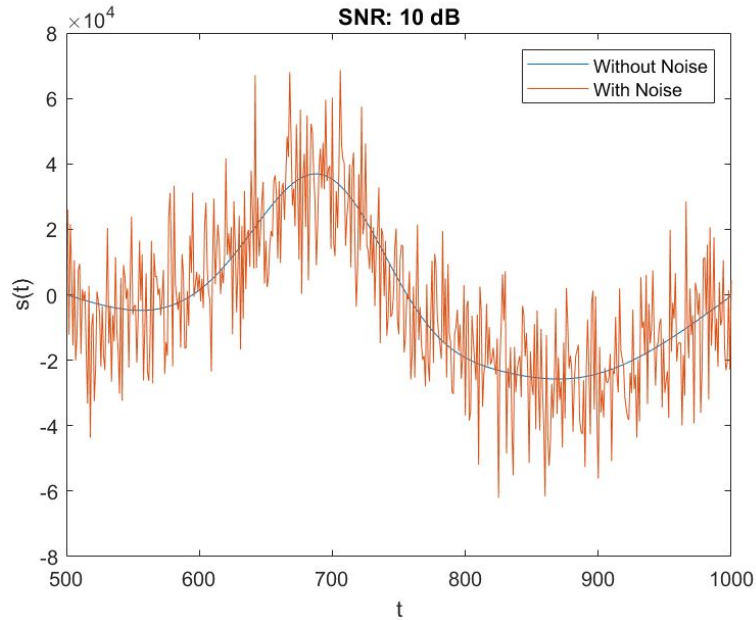


Fig. 7: Noise Profile in Middle Line (#26) and Middle pFOV (#291)

- Noise

$$s(t) = s_{ideal}(t) + n(t)$$

SNR Optimization for Lumped PCI

- Weighting each sample before raw Lumped-PCI creation such that SNR of raw Lumped-PCI image will be optimized.

$$\begin{aligned}\tilde{\rho}_{lum,ideal} &= \sum_{k=-K}^K \beta_k \tilde{\rho}_k, \text{ where } k = -K, \dots, K \\ \tilde{\rho}_{lum} &= \sum_{k=-K}^K (\beta_k \tilde{\rho}_k + n_k) \\ &= \sum_{k=-K}^K c_k (\beta_k \tilde{\rho}_k + n_k) \\ &= \sum_{k=-K}^K c_k \beta_k \tilde{\rho}_k + \sum_{k=-K}^K c_k n_k\end{aligned}$$

$$SNR = 20 \log_{10} \left(\frac{|\sum_{k=-K}^K c_k \beta_k \tilde{\rho}_k|}{\sigma'} \right)$$

$$\sigma' = \sqrt{\sum_{k=-K}^K c_k^2 \sigma^2} = \sigma \sqrt{\sum_{k=-K}^K c_k^2}$$

SNR Optimization for Lumped PCI

$$\begin{aligned}
 \frac{\partial}{\partial c_j} \left(\frac{\sum_{k=-K}^K c_k \beta_k \tilde{\rho}_k}{\sigma \sqrt{\sum_{k=-K}^K c_k^2}} \right) &= \frac{\sigma \beta_j \tilde{\rho}_j \sqrt{\sum_{k=-K}^K c_k^2} - \sum_{k=-K}^K c_k \beta_k \tilde{\rho}_k \cdot \left(\frac{1}{2 \sqrt{\sum_{k=-K}^K c_k^2}} 2 c_j \sigma \right)}{\sigma^2 \sum_{k=-K}^K c_k^2} \\
 &= \beta_j \tilde{\rho}_j \sqrt{\sum_{k=-K}^K c_k^2} - \sum_{k=-K}^K c_k \beta_k \tilde{\rho}_k \cdot \frac{c_j}{\sqrt{\sum_{k=-K}^K c_k^2}} \\
 &= \beta_j \tilde{\rho}_j \sum_{k=-K}^K c_k^2 - \left(\sum_{k=-K}^K c_k \beta_k \tilde{\rho}_k \right) \cdot c_j = 0
 \end{aligned}$$

SNR Optimization for Lumped PCI

- Signal samples at each gridding position should be weighted by itself.
- For conserving Linear Shift-Invariance (LSI) property of Lumped-PCI, PSF blurred image is assumed to be uniform.

$$c_j = \beta_j \tilde{\rho}_j \frac{\sum_{k=-K}^K c_k^2}{\sum_{k=-K}^K c_k \beta_k \tilde{\rho}_k}$$

$$\mu = \frac{\sum_{k=-K}^K c_k^2}{\sum_{k=-K}^K c_k \beta_k \tilde{\rho}_k}$$

$$c_j = \mu \beta_j \tilde{\rho}_j$$

Results

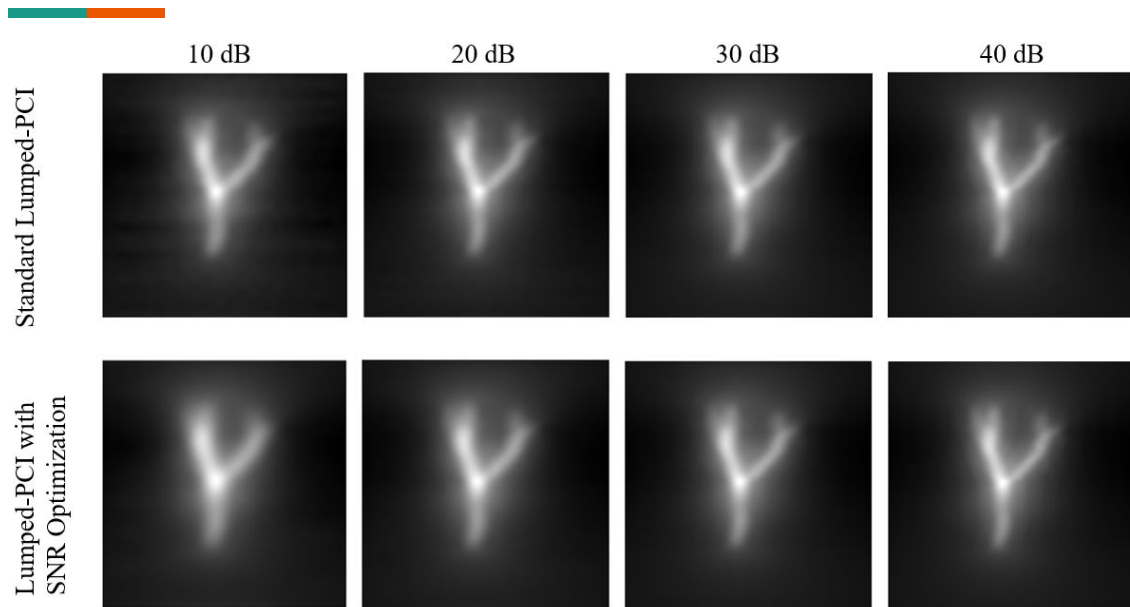


Fig. 8: Results of Lumped-PCI with SNR Optimization

Interference



Interference

- **Signal-to-Interference Ratio (SIR) Definition in Simulations**

$$SIR = 20 \log_{10} \left(\frac{\max_f |S_n(f)|}{\gamma_n} \right)$$
$$S_n(f) = \begin{cases} S(f), & \left(n - \frac{1}{2}\right) f_0 < f < \left(n + \frac{1}{2}\right) f_0 \\ 0, & \text{otherwise} \end{cases}$$
$$f = n f_0$$

→ Interference robustness difference between PCI and Lumped-PCI

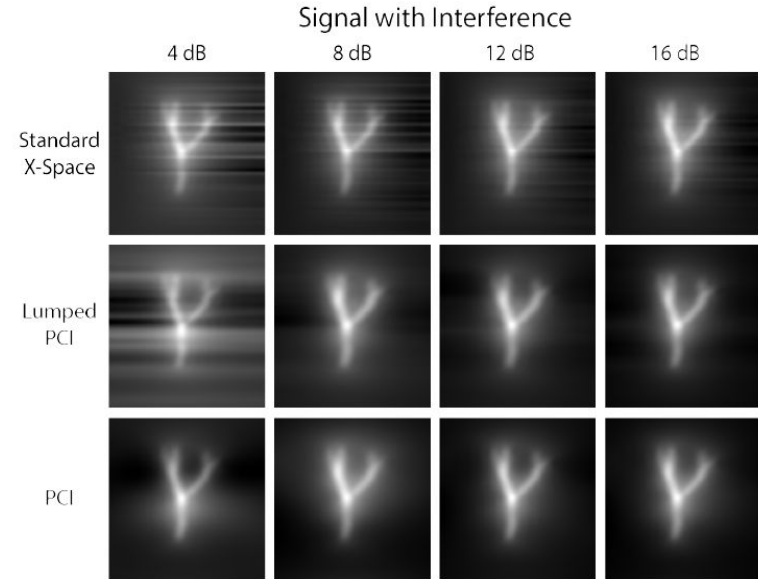


Fig. 9: Comparison of Standard X-Space Reconstruction, Lumped PCI and PCI Results at Different SIR Levels. [3]

Interference Analysis

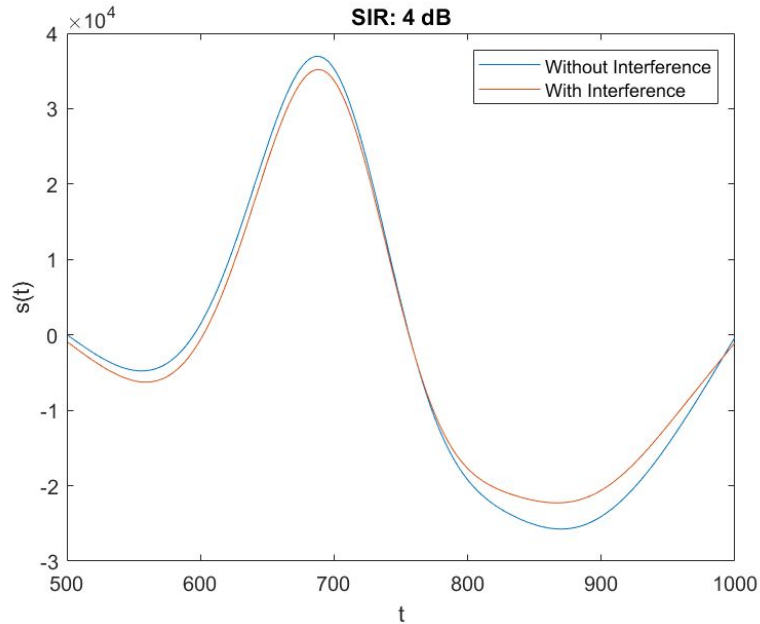


Fig. 10: Interference Profile in Middle Line (#26) and Middle pFOV (#291)

- Harmonic Interference

$$s(t) = s_{ideal}(t) + i(t)$$

$$s(t) = \sum_n S_n \sin(2\pi f_n t) + \sum_n I_n \sin(2\pi f_n t + \phi_n)$$

$$\text{where } I_n \sim U(0, \gamma_n) \text{ and } \phi_n \sim N\left(0, \left(\frac{\pi}{10}\right)^2\right)$$

Interferencing Signal Decomposition

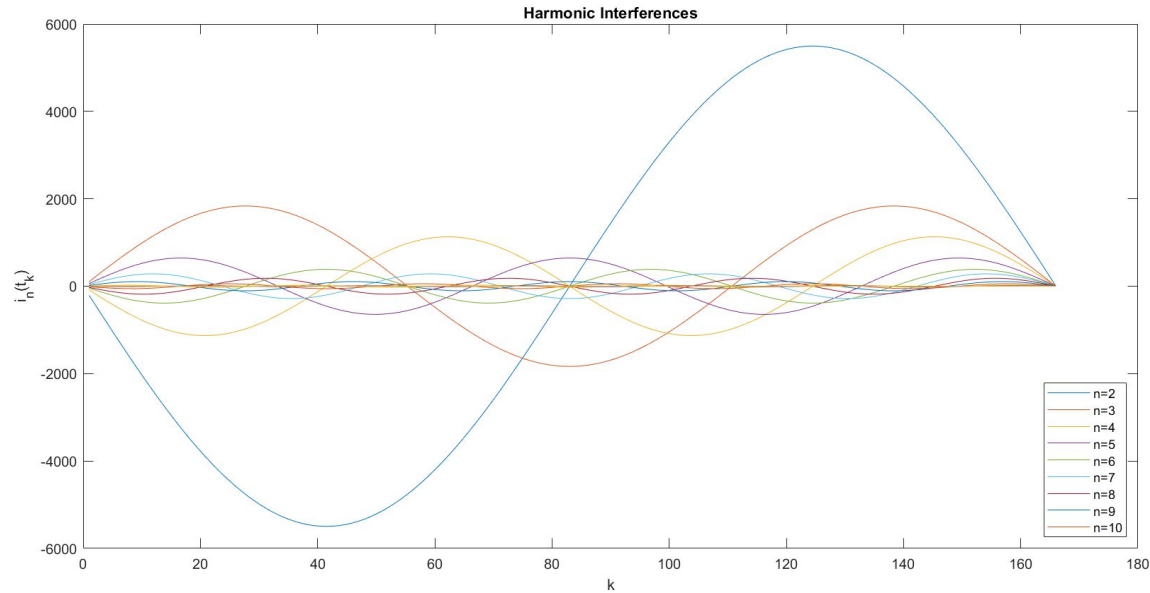


Fig. 11: Expectances of Harmonic Interferences for Middle Line.

Interferencing Signal

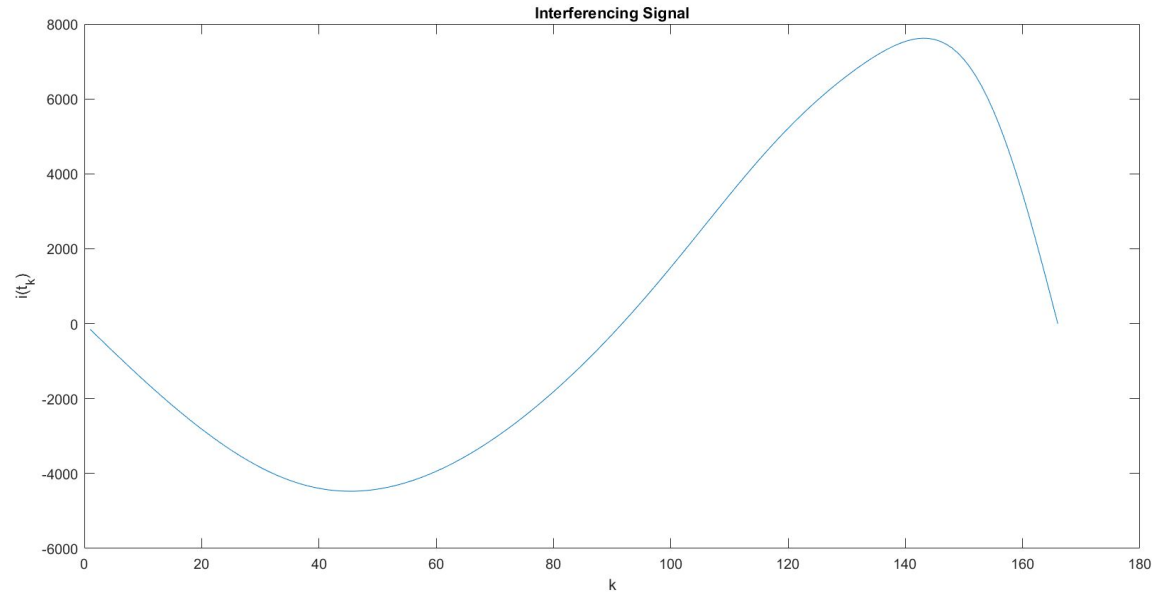


Fig. 12: Total Expectance of Harmonic Interferences for Middle Line

Interferencing Signal Decomposition

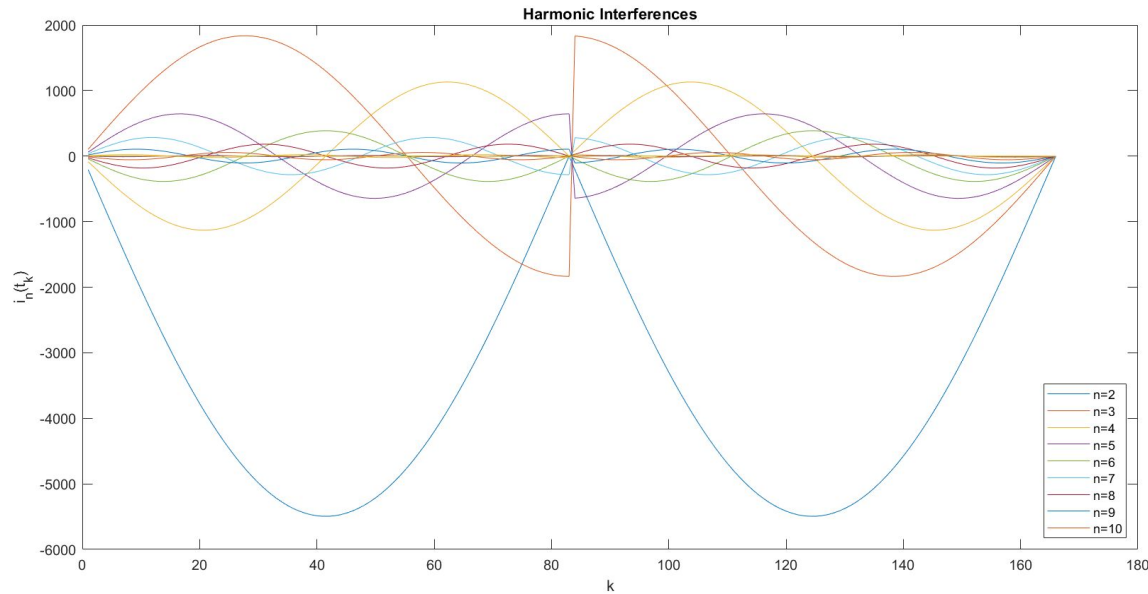


Fig. 13: Expectances of Harmonic Interferences for Middle line
(Kernel signs are included.)

Interferencing Signal

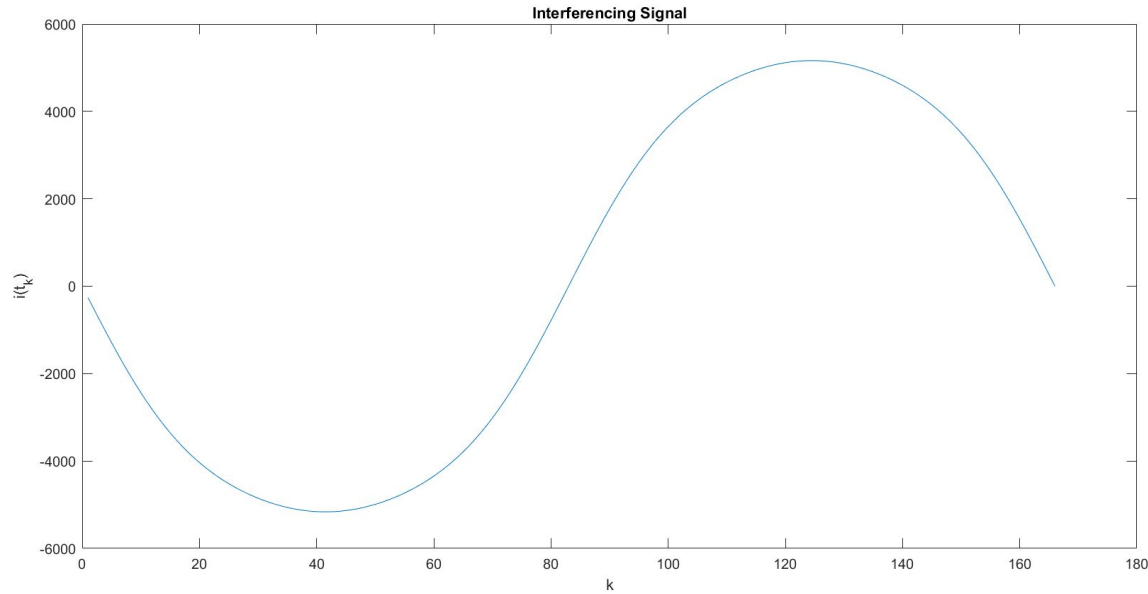


Fig. 14: Total Expectance of Harmonic Interferences for Middle Line.
(Kernel signs are included.)

Results

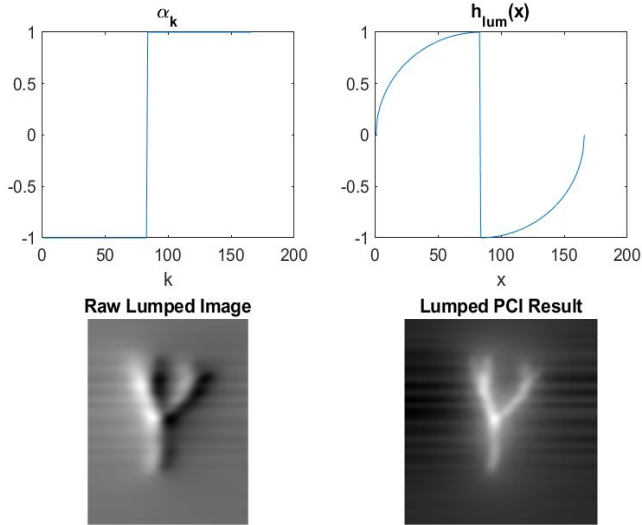


Fig. 15: Lumped PCI result at 4 dB SIR level.
Kernel is the standard kernel.

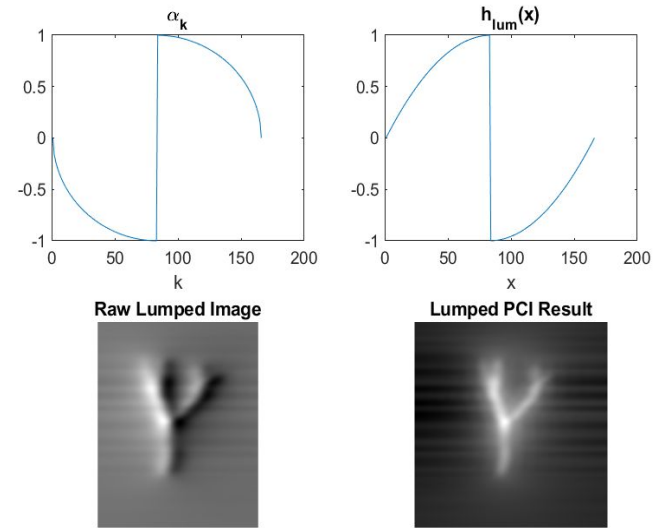


Fig. 16: Lumped PCI result at 4 dB SIR level.
Kernel is the kernel used in SNR optimization
for Lumped PCI.

Results

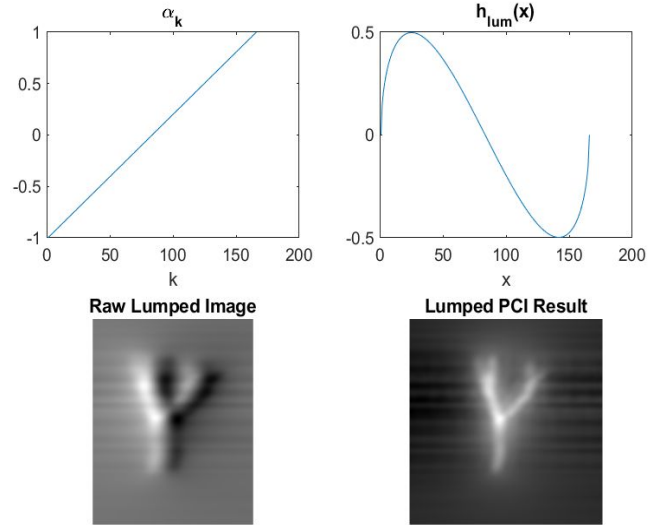


Fig. 17: Lumped PCI result at 4 dB SIR level.
Kernel is used for emphasizing edges.

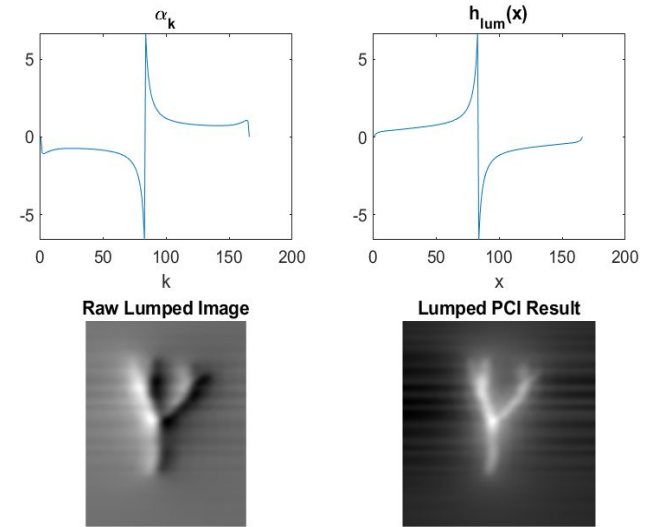


Fig. 18: Lumped PCI result at 4 dB SIR level.
Kernel is used for suppressing samples with maximum interference.

Discussion



Discussion



- Noise robustness and interference robustness were studied separately.
- Noise robustness of Lumped PCI was increased by SNR optimization technique.
- Interference profile for Lumped PCI was analysed.
- Interference robustness of Lumped PCI could not be increased yet.

Future Work



Future Work



New Approaches for Interference Robustness of Lumped-PCI

- Weighting strategy for summing harmonic interference expectations to be zero. Kernel has the property of odd function. Interference at certain harmonics are already summed up to be zero.
- Another approach may be using PCI image for Lumped-PCI to boost interference robustness.

References



- [1] P.W. Goodwill, E.U. Saritas, L.R. Croft, T.N. Kim, K.M. Krishnan, D.V. Schaffer, S.M. Conolly, X-Space MPI: magnetic nanoparticles for safe medical imaging, *Adv. Mater.* 24 (2012) 3870–3877.
- [2] S. Kurt, “Rapid and Robust Image Reconstruction for magnetic Particle imaging,” M.S. Thesis, Grad. Sch. of Engineering and Science, Bilkent University, 2020.
- [3] S. Kurt, Y. Muslu and E. U. Saritas, "Partial FOV Center Imaging (PCI): A Robust X- Space Image Reconstruction for Magnetic Particle Imaging," in *IEEE Transactions on Medical Imaging*, vol. 39, no. 11, pp. 3441-3450, Nov. 2020. doi: 10.1109/TMI.2020.2995410
- [4] P. W. Goodwill and S. M. Conolly, “The x-space formulation of the magnetic particle imaging process: 1-D signal, resolution, bandwidth, SNR, SAR, and magnetostimulation,” *IEEE Transactions on Medical Imaging*, vol. 29, no. 11, pp. 1851– 1859, 2010.
- [5] K. Lu, P. W. Goodwill, E. U. Saritas, B. Zheng, and S. M. Conolly, “Linearity and shift invariance for quantitative magnetic particle imaging,” *IEEE Transactions on Medical Imaging*, vol. 32, no. 9, pp. 1565–1575, 2013.