

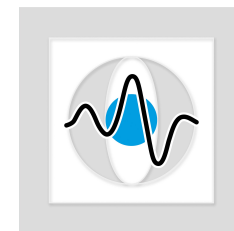
# Medical Image Processing for Diagnostic Applications

## Parallel Beam – Reconstruction Steps

Online Course – Unit 34

Andreas Maier, Joachim Hornegger, Markus Kowarschik, Frank Schebesch

Pattern Recognition Lab (CS 5)



# Topics

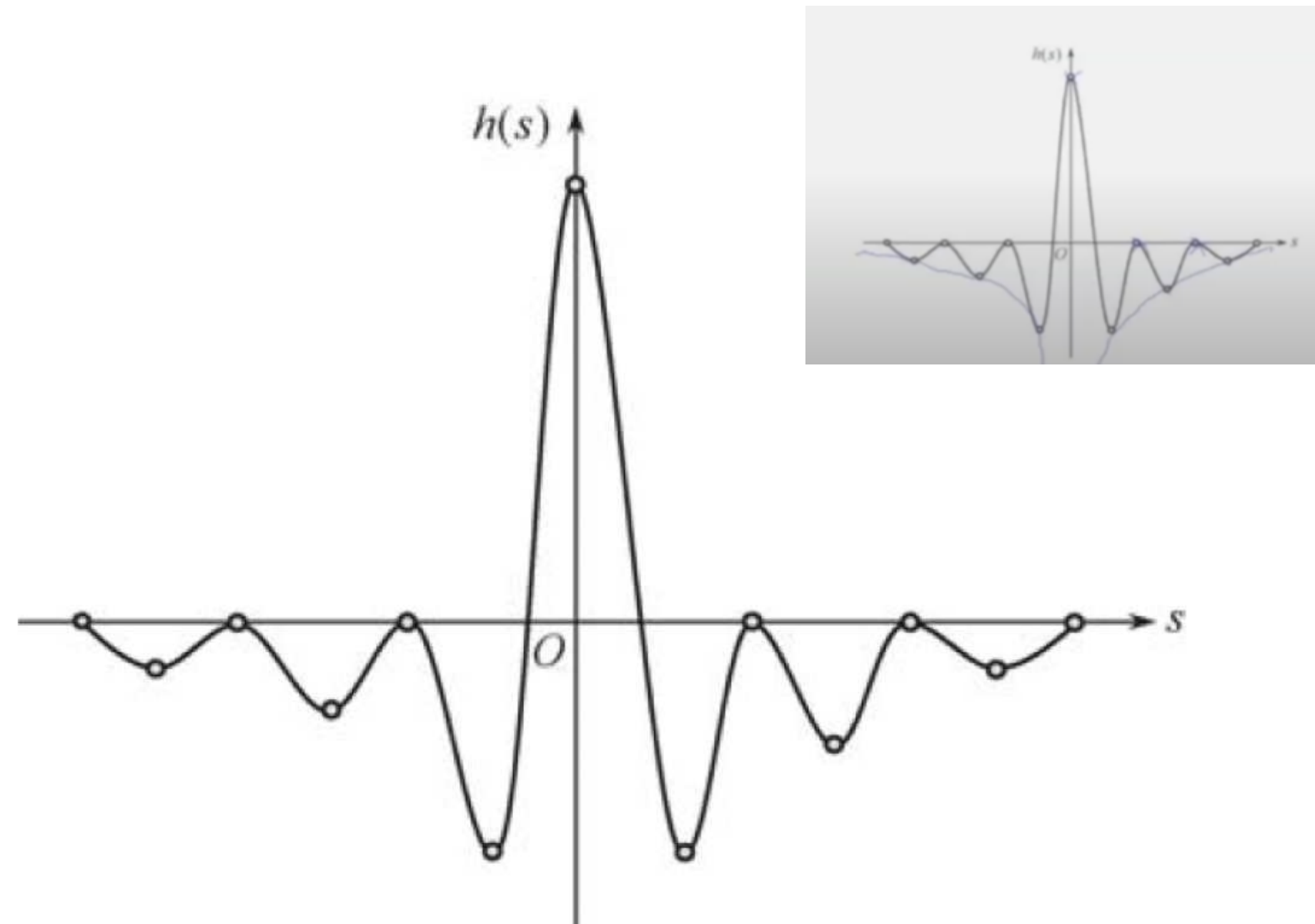
How to Implement a Parallel Beam Algorithm - Part 2

- Discrete Spatial vs. Continuous Frequency Version
- Practical Algorithm
- Backprojection Example

## Summary

- Take Home Messages
- Further Readings

## Ram-Lak: Discrete Spatial Form of the Ramp Filter



we only sample at integer position  
the real signal oscillates ??

Figure 1: Continuous and discrete graph of the Ram-Lak filter (Zeng, 2009)

# Discrete Spatial vs. Continuous Frequency Version

- Continuous frequency representation of the ramp filter:

$$H(\omega) = |\omega|$$

- Discrete spatial form:

$$h(n\tau) = \begin{cases} \frac{1}{4\tau^2} & n = 0, \\ 0 & n \text{ even}, \\ -\frac{1}{\pi^2(n\tau)^2} & n \text{ odd} \end{cases}$$

# Discrete Spatial vs. Continuous Frequency Version

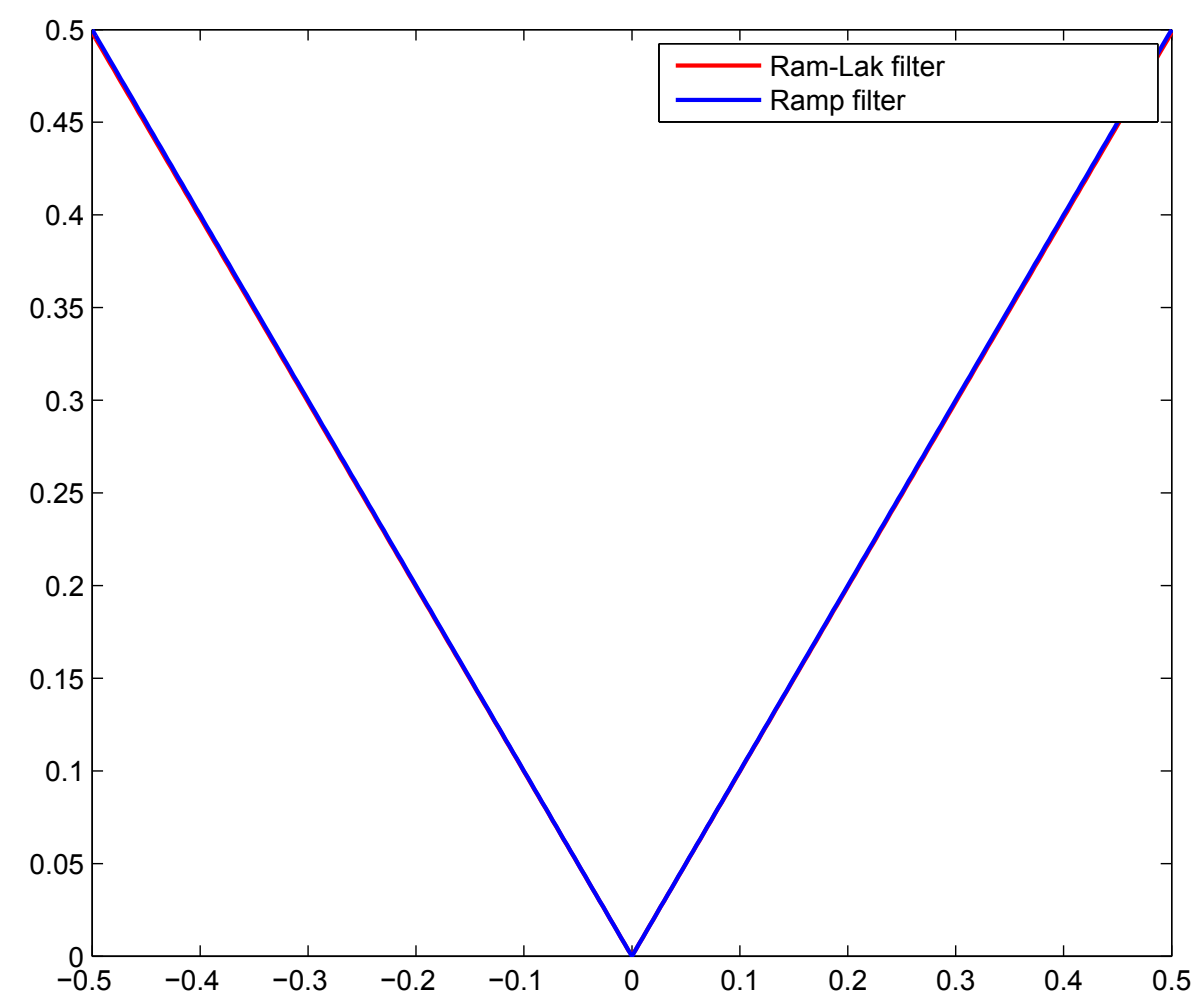
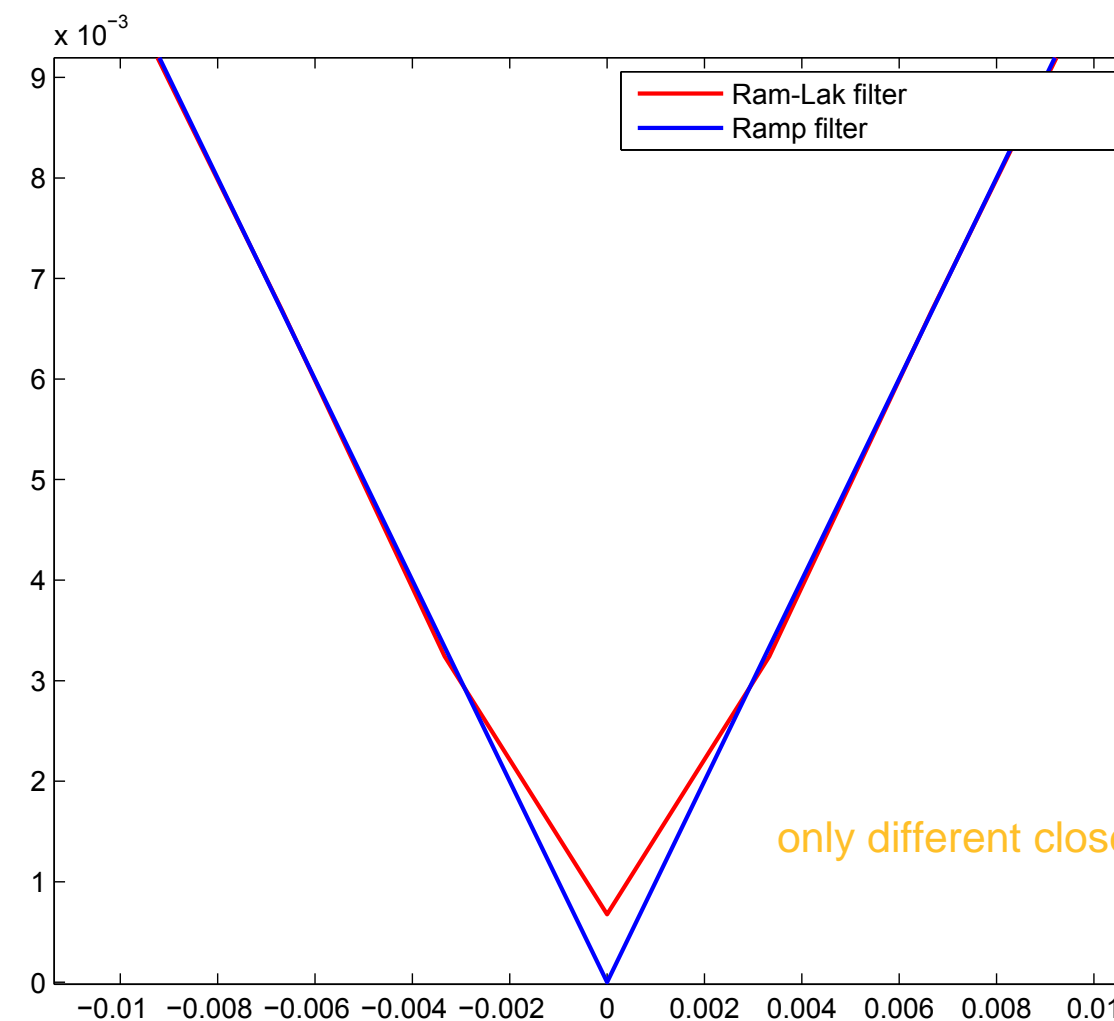


Figure 2: Plot of the ramp filter, whole scale

# Discrete Spatial vs. Continuous Frequency Version



only different close to zero -> with RAM the mean is not zero :)

Figure 3: Comparison of the ramp and the Ram-Lak filter, zoomed in at zero

## Example: Homogeneous Cylinder after Filter

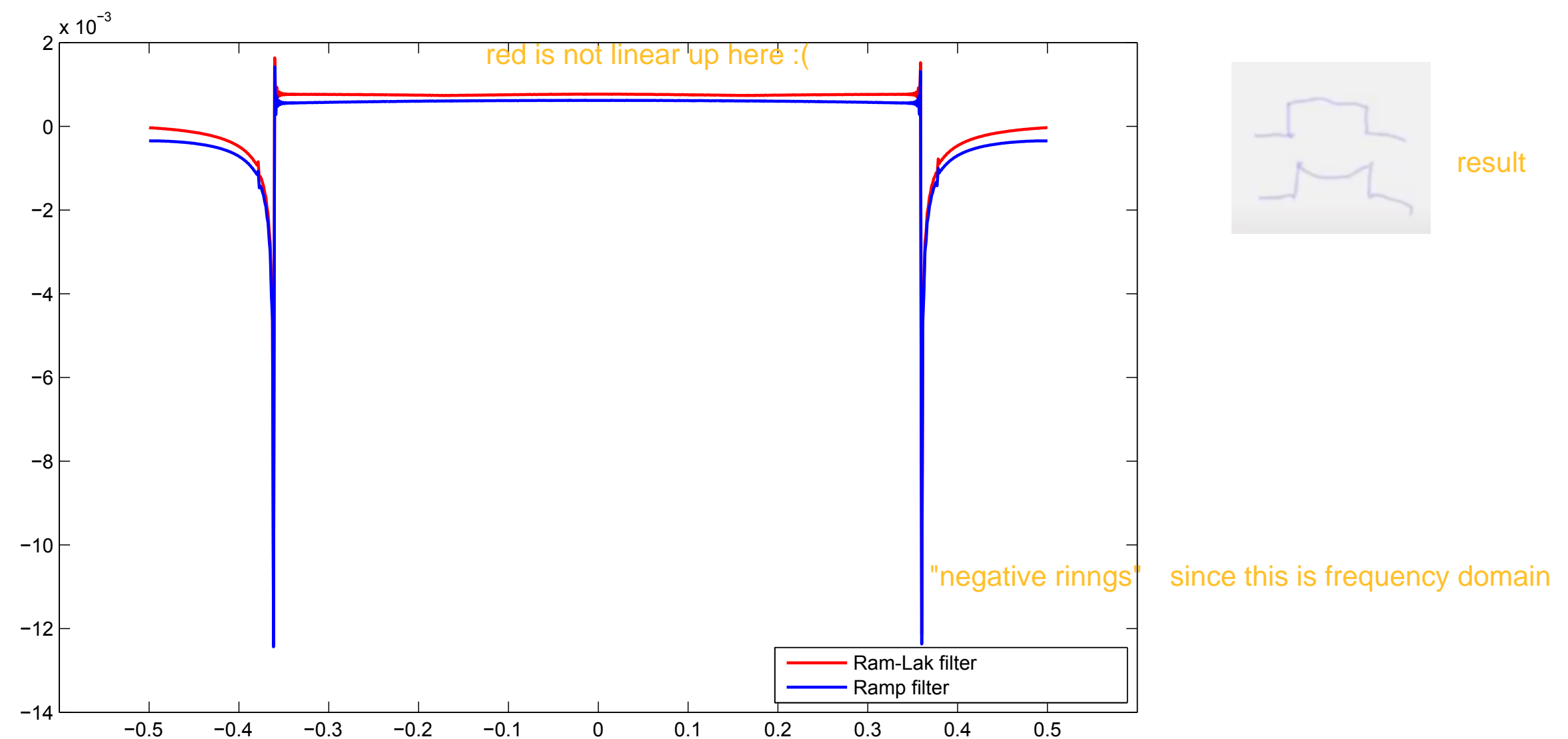


Figure 4: Filtered projection profile of the cylinder phantom



# Practical Algorithm - Filtering

1. Precompute filter  $h(s)$  in spatial domain  $O(N)$
2. Transform filter to frequency domain  $H(\omega)$  via FFT  $O(N \log N)$
3. For each of  $\#P$  projections:
  - Compute FFT of  $p(s, \theta)$   $O(N \log N)$
  - Apply filter  $P(\omega, \theta) \cdot H(\omega)$   $O(N)$
  - Compute filtered projection  $q(s)$  via iFFT  $O(N \log N)$

Total complexity:  $O(N + N \log N + \#P(N + 2N \log N)) = O(\#P N \log N)$



# Practical Algorithm - Backprojection

1. Initialize  $f(x, y) = 0$   $O(N^2)$
2. For each of  $N \times N$  pixels:
  - For each of  $\#P$  projections:
    - Compute  $s = x \cos \theta + y \sin \theta$   $O(1)$
    - Update  $f(x, y) += q(s, \theta)$   $O(1)$

Total complexity:  $O(N^2 + N^2 \#P(1 + 1)) = O(N^2 \#P)$

# Practical Algorithm: Overall Complexity

- Apply filter on the detector row:

$$O(\#P \, N \log N)$$

- Backproject:

$$O(\#P \, N^2)$$

# Backprojection Example

backprojection      just smear the intensity over the image

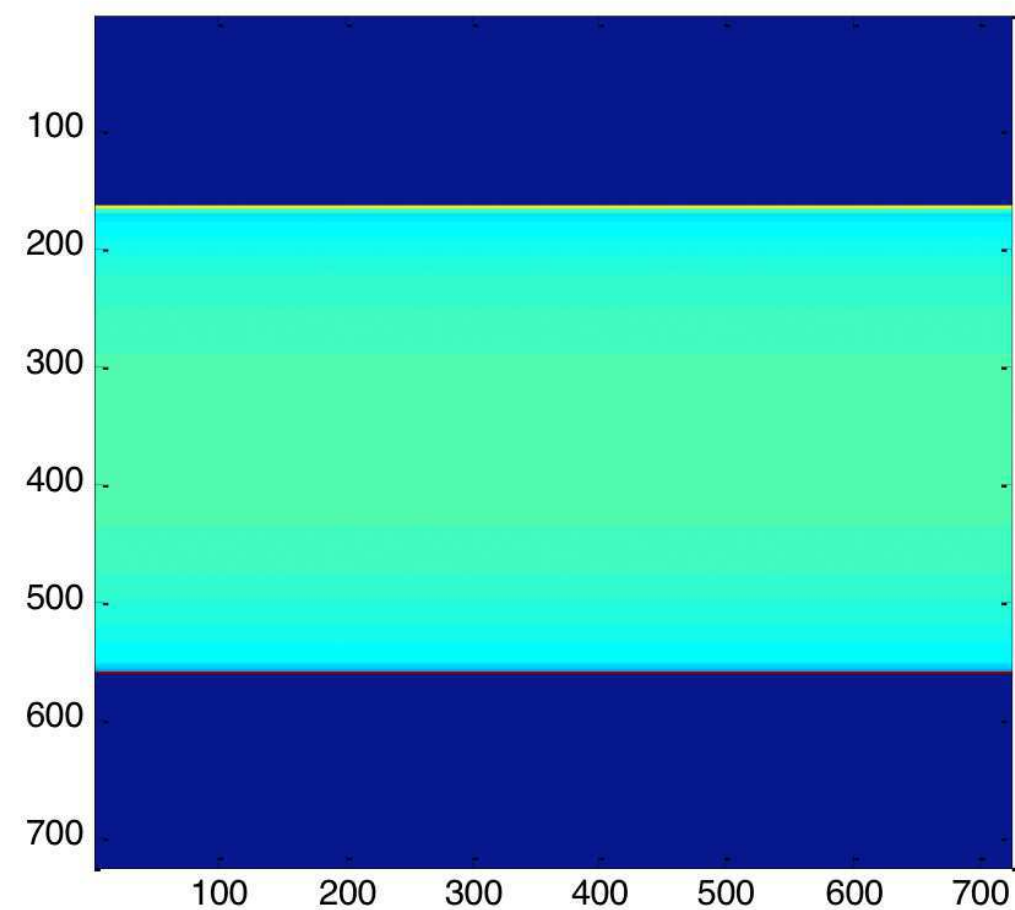
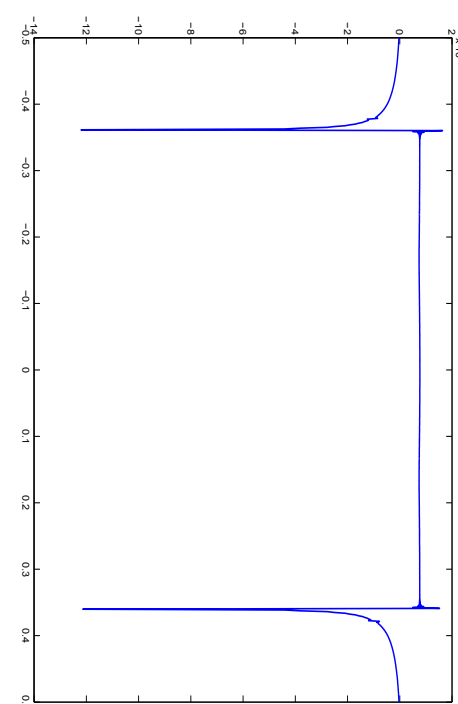


Figure 5: Backprojection of a single projection

# Backprojection and Fourier Slice Theorem

one backprojection is one line in k space

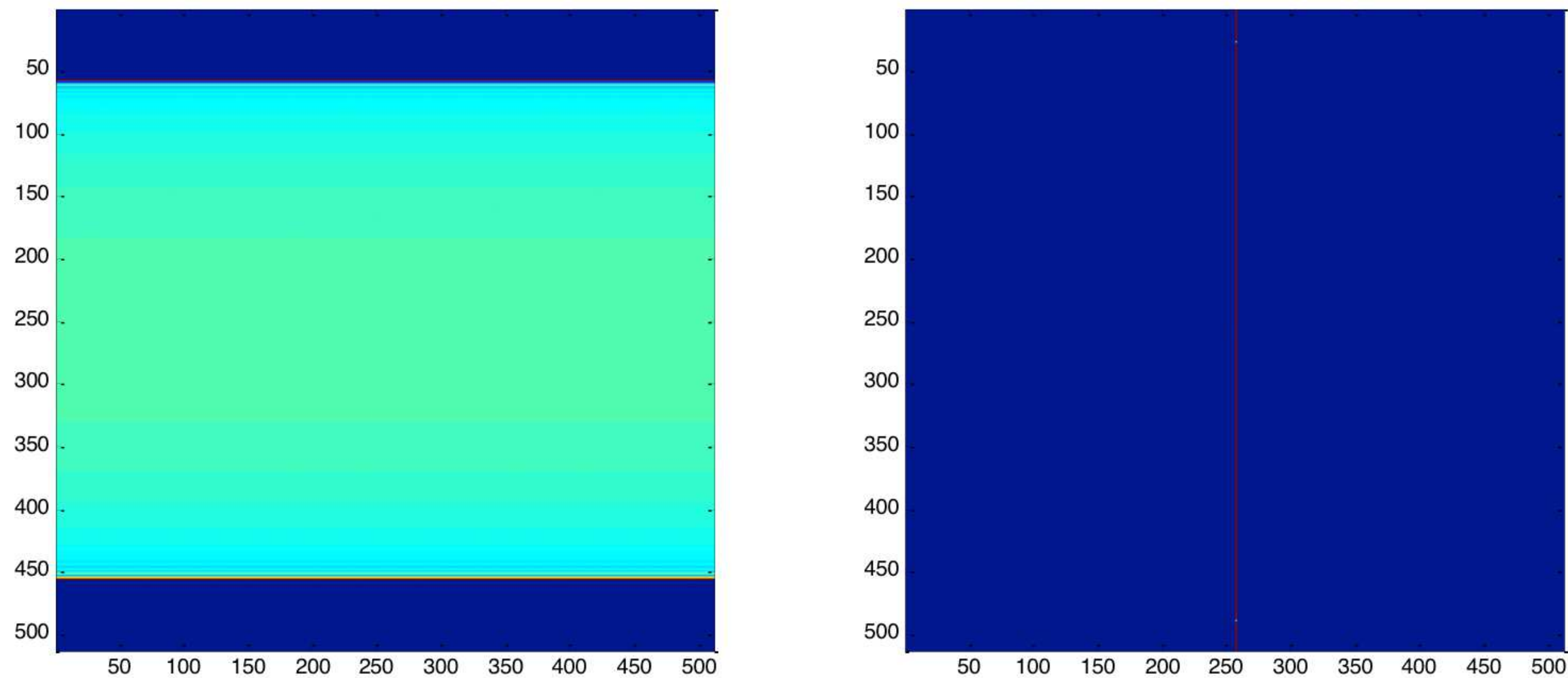


Figure 6: Backprojection (left) of a single line in the Fourier space (right)

# Backprojection Example

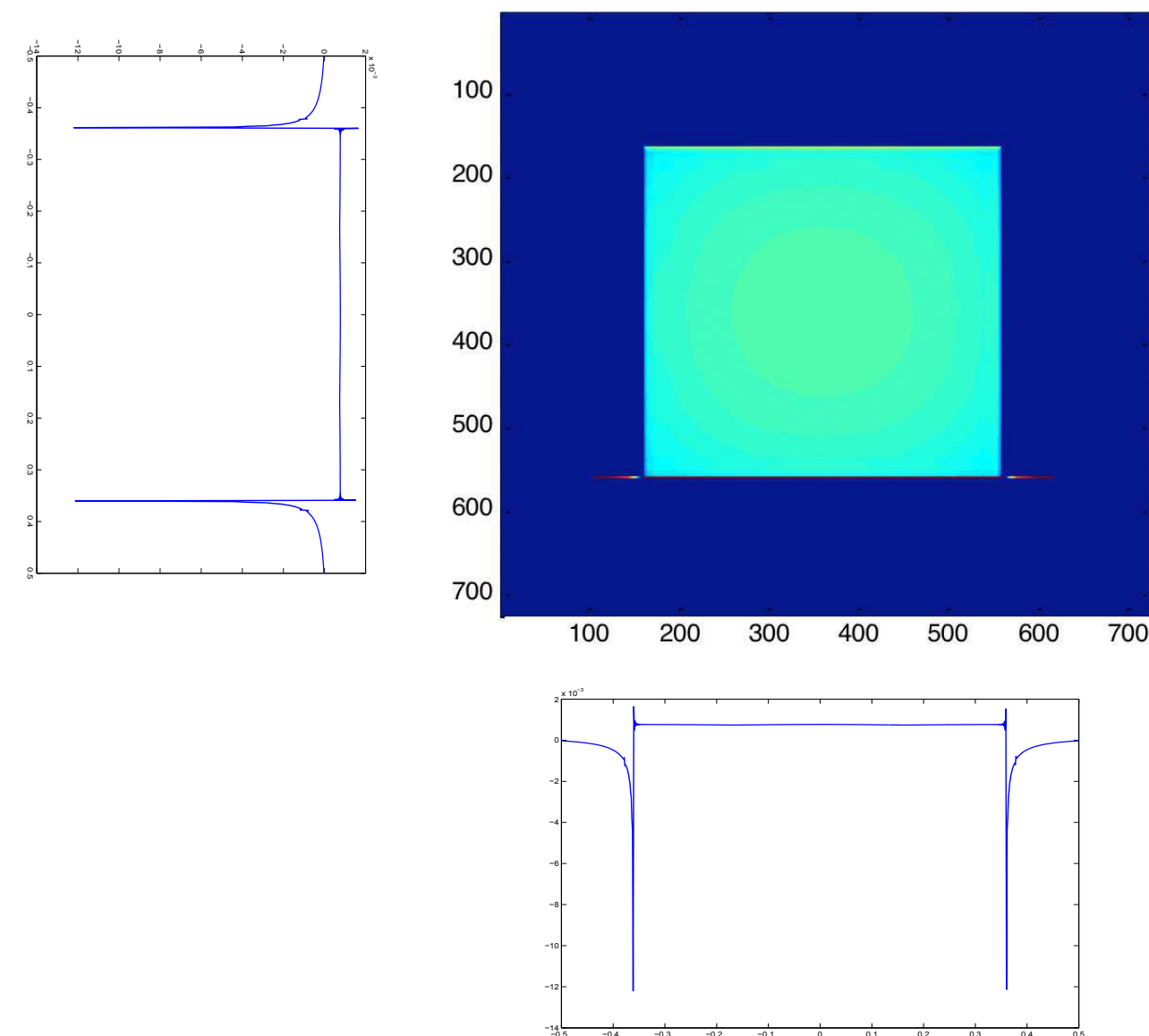


Figure 7: Backprojection of two projections ( $0^\circ$ ,  $90^\circ$ )

# Backprojection Example

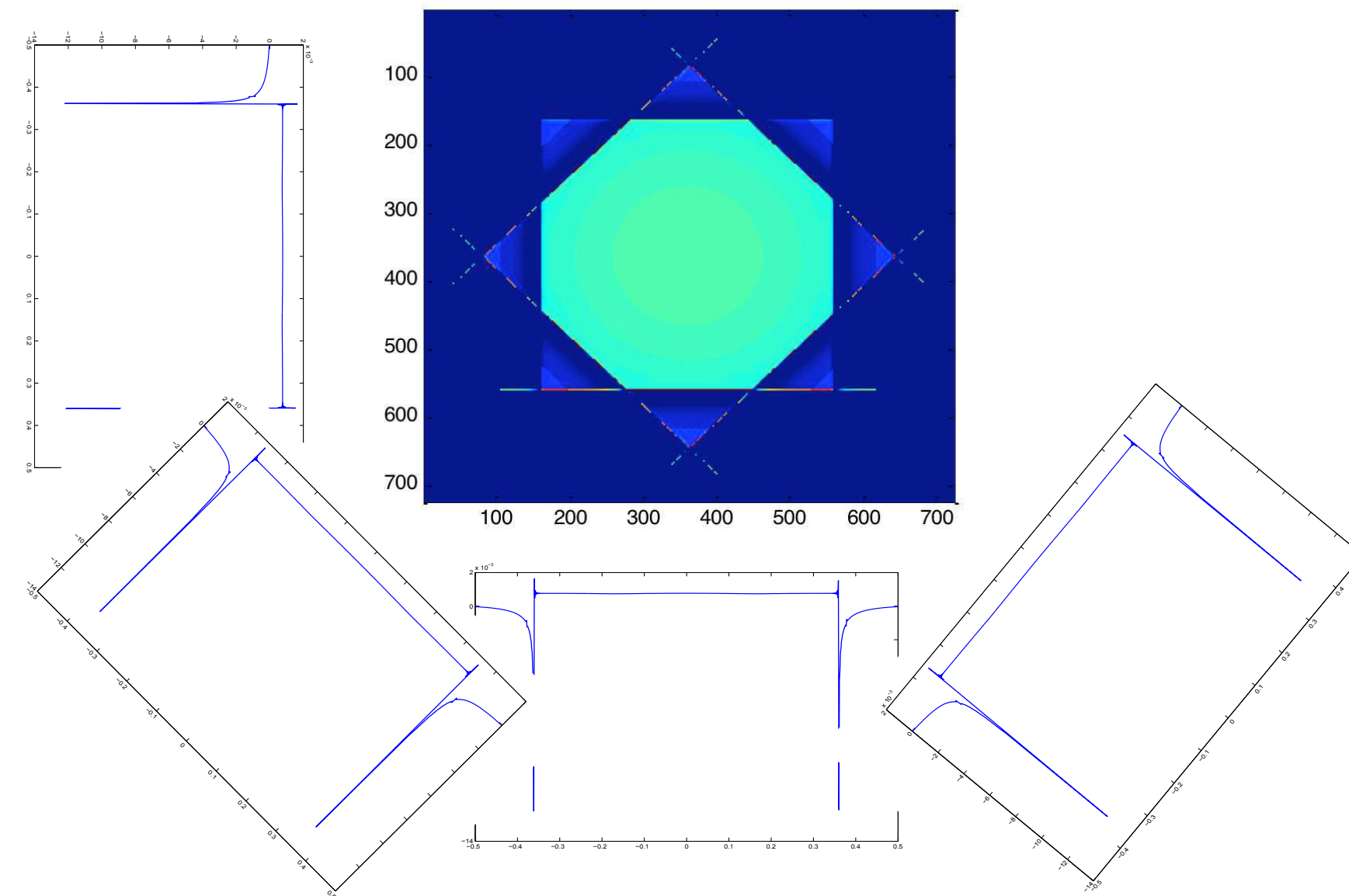


Figure 8: Backprojection of four projections ( $0^\circ$ ,  $45^\circ$ ,  $90^\circ$ ,  $135^\circ$ )

# Backprojection Example

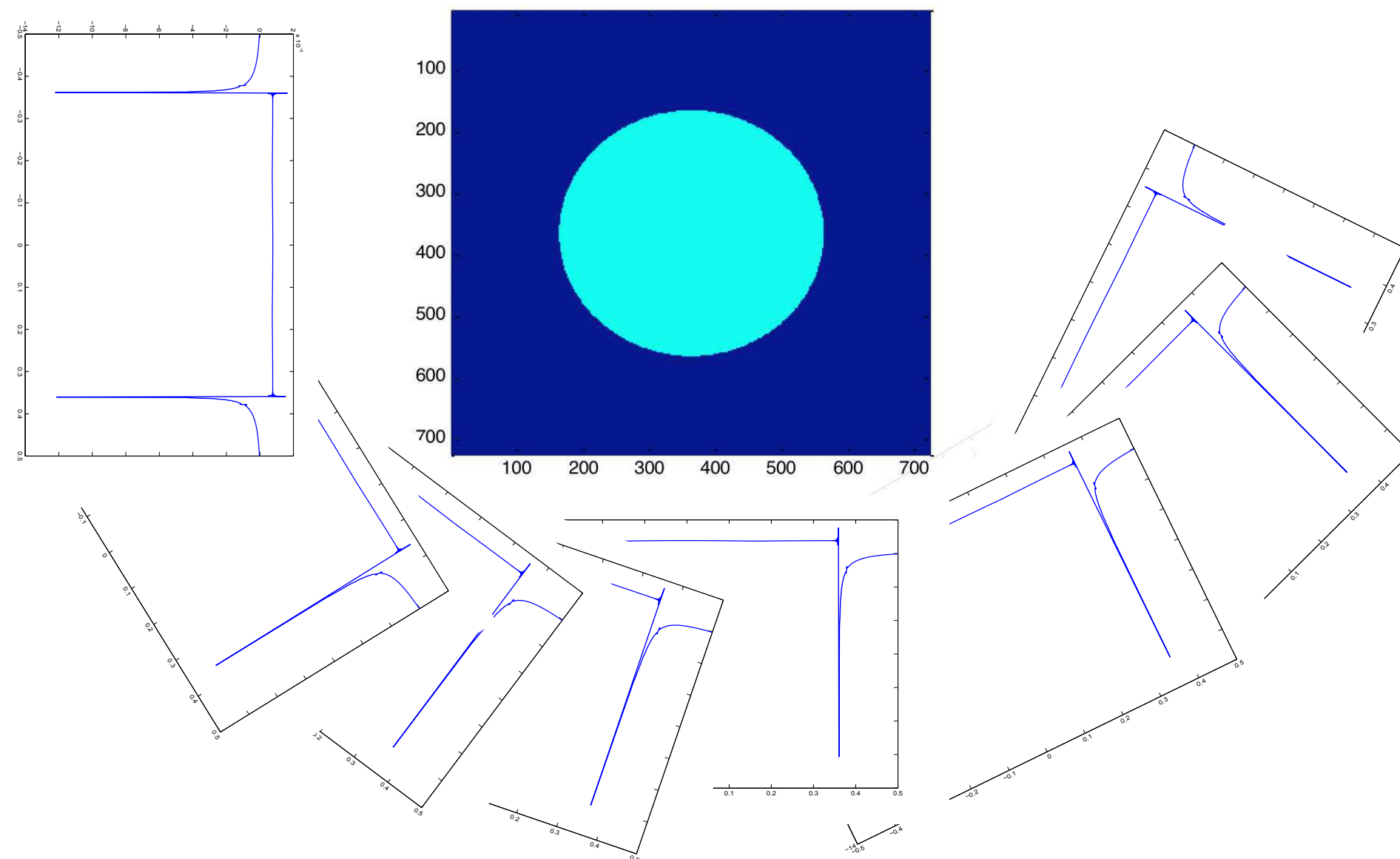


Figure 9: Backprojection of multiple projections ( $0^\circ$ - $180^\circ$ )



# Topics

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## Summary

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## Take Home Messages

- Although the original ramp filter converges to zero at zero frequency, the **Ram-Lak** filter takes **low frequencies** into account. This enables discrete computations to be more accurate.
- The **filtering** has a complexity of  $O(N \log N)$ , and the **backprojection** a complexity of  $O(N^2)$  per projection.
- Increasing the number of projections improves the reconstruction result.

## Further Readings

The original Ram-Lak article is:

G. N. Ramachandran and A. V. Lakshminarayanan. “Three-dimensional Reconstruction from Radiographs and Electron Micrographs: Application of Convolutions instead of Fourier Transforms”. In: *Proceedings of the National Academy of Sciences of the United States of America* 68.9 (Sept. 1971), pp. 2236–2240

The derivation shown in this unit is based on a document by [Martin Berger](#).

The concise reconstruction book from ‘Larry’ Zeng:

Gengsheng Lawrence Zeng. *Medical Image Reconstruction – A Conceptual Tutorial*. Springer-Verlag Berlin Heidelberg, 2010. DOI: [10.1007/978-3-642-05368-9](https://doi.org/10.1007/978-3-642-05368-9)

Another mathematical examination of filtered backprojection can be found in

Thorsten Buzug. *Computed Tomography: From Photon Statistics to Modern Cone-Beam CT*. Springer Berlin Heidelberg, 2008. DOI: [10.1007/978-3-540-39408-2](https://doi.org/10.1007/978-3-540-39408-2)