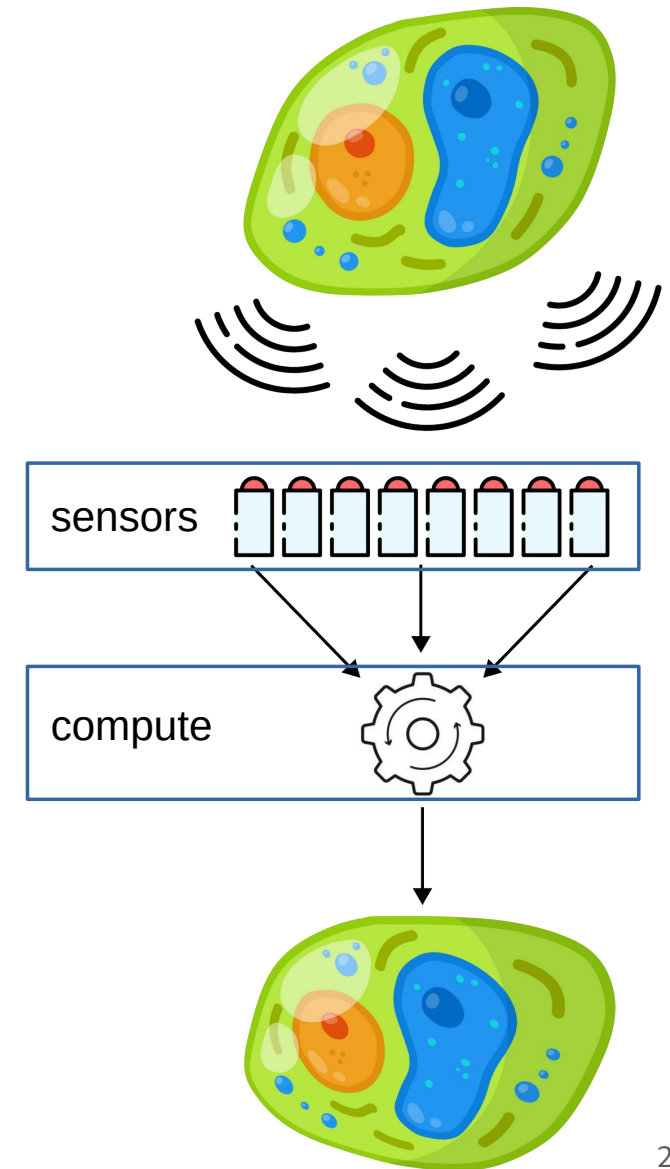
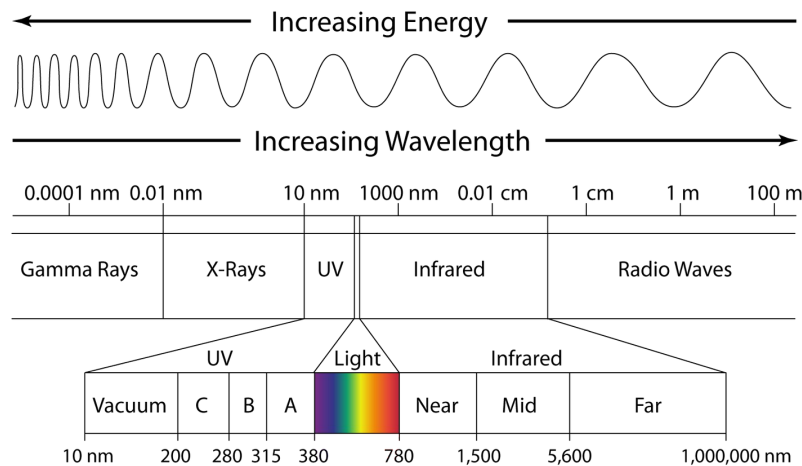


Pyxu: a Modular Approach to Imaging across Domains and Scales

Sepand Kashani

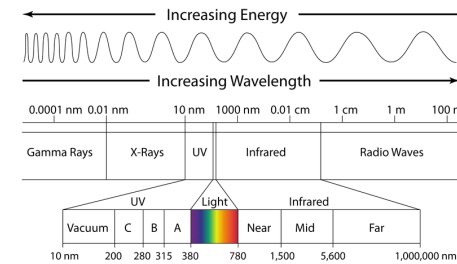
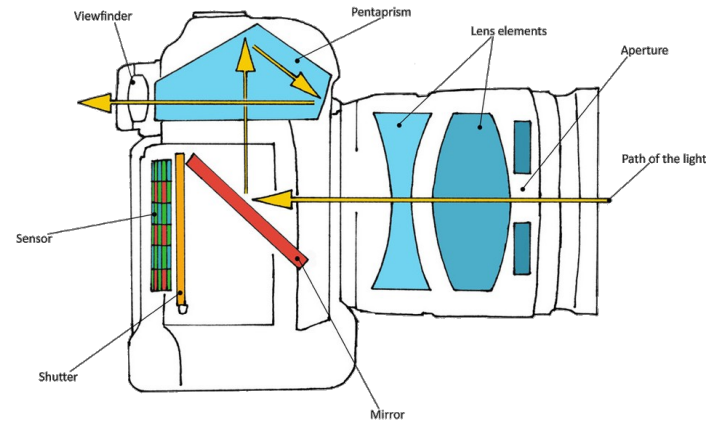
Computational Imaging (CI)

- Inverse problem involving **image**-like quantities
 - Capture signals from real world with sensors
 - Infer something about quantity of interest via **computation**



CI: Optical Imaging

- Capture scene radiance
 - Visible light enters camera
 - Recorded on pixel detector



scene



sensor recording

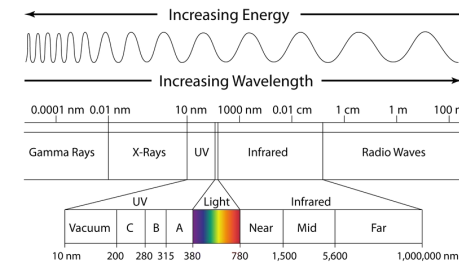
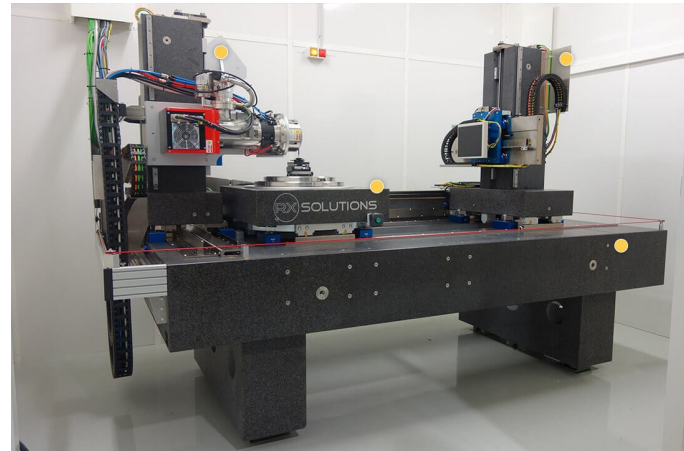


measurement

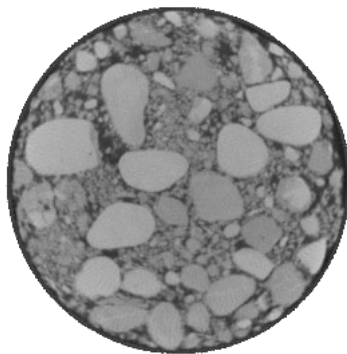
computation

CI: Tomography

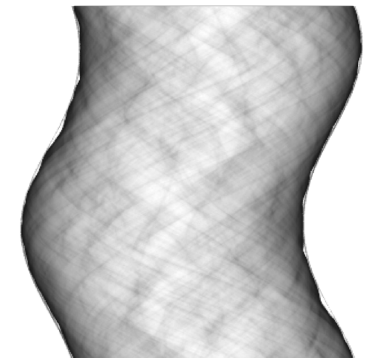
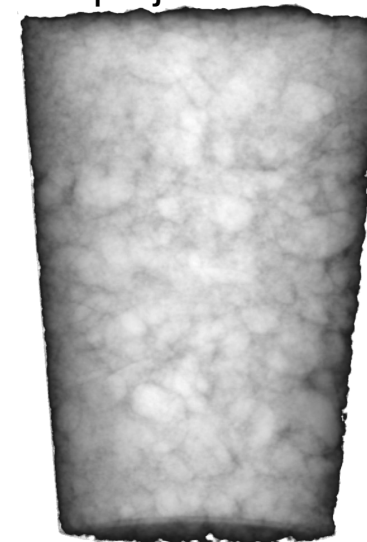
- Determine volume absorption profile
 - Project X-rays through object
 - Record shadows from different directions



3D volume



2D projections

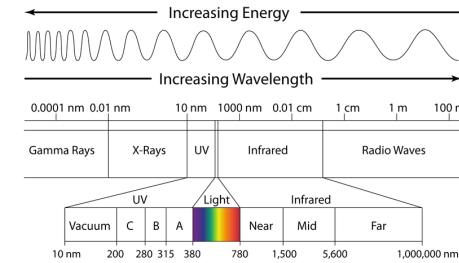
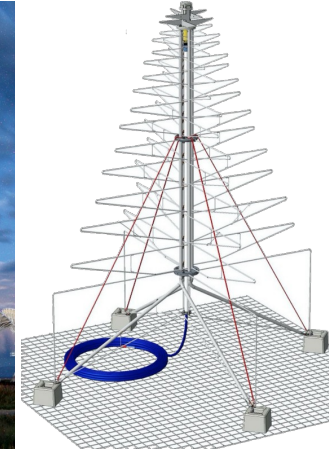


measurement

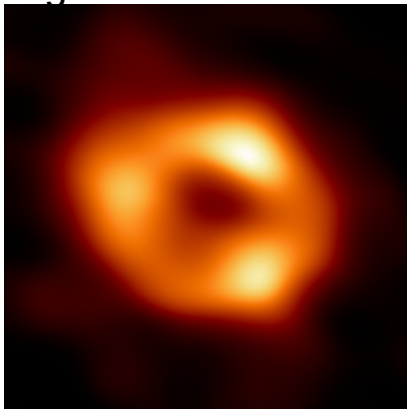
computation

CI: Radio-Interferometry

- Determine sky brightness distribution
 - Stars emit radio emissions (among other things)
 - Recorded on Earth with antennas



Sgr A*



measurement

computation

antenna time series

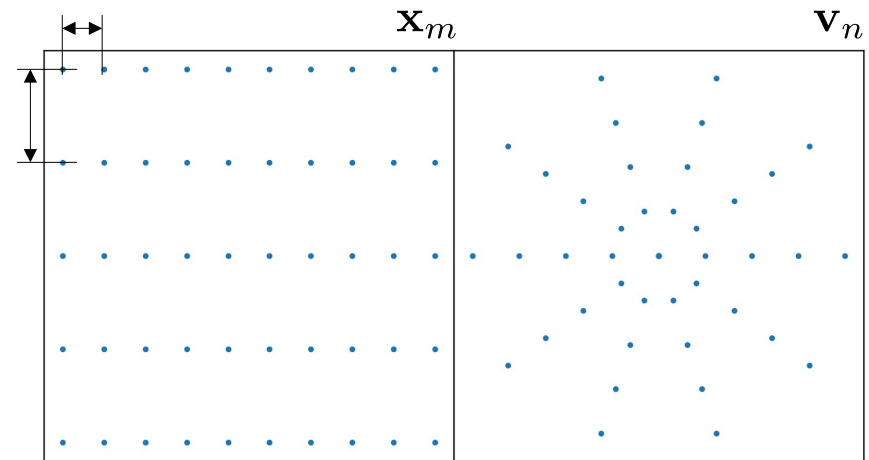


Solving CI Inverse Problems

- Relate quantity of interest with measurements

$$\mathbf{y} = [\mathcal{A}_Q \circ \cdots \circ \mathcal{A}_1] \mathbf{f} + \mathbf{n}$$

- Typical forward models A encountered in imaging (after discretization):



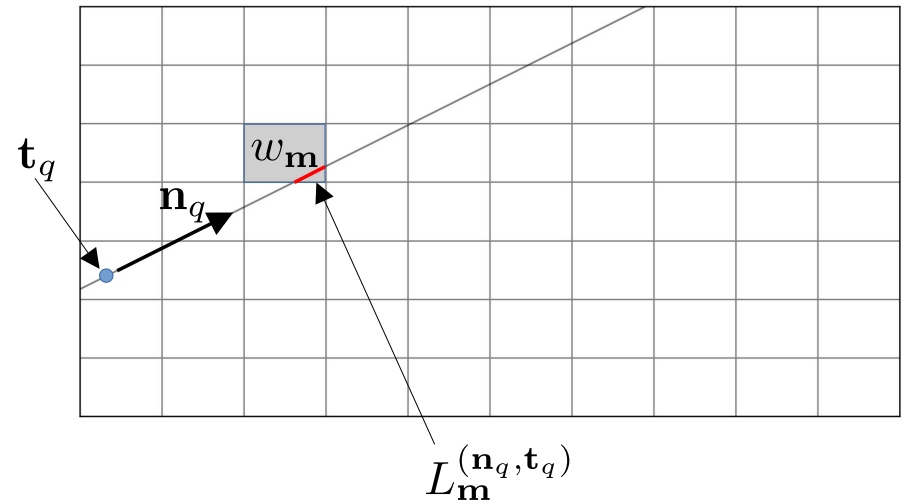
Fourier Sums

$$[\mathbf{z}_n]_{n=1}^N = \sum_{m=1}^M w_m e^{-j2\pi \langle \mathbf{x}_m, \mathbf{v}_n \rangle}$$

\mathbb{R}^D knots

Line Integrals

$$[\mathbf{z}_{\mathbf{n}_q, \mathbf{t}_q}]_{q=1}^Q = \sum_{\mathbf{m} \in \mathbb{Z}^D} w_{\mathbf{m}} L_{\mathbf{m}}^{(\mathbf{n}_q, \mathbf{t}_q)}$$



The CI Software Landscape

- Rich software tools for all types of imaging
 - Silo software: re-inventing the wheel
 - Slow dissemination across domains

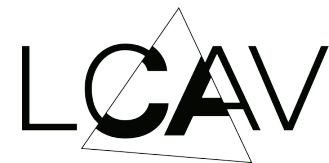




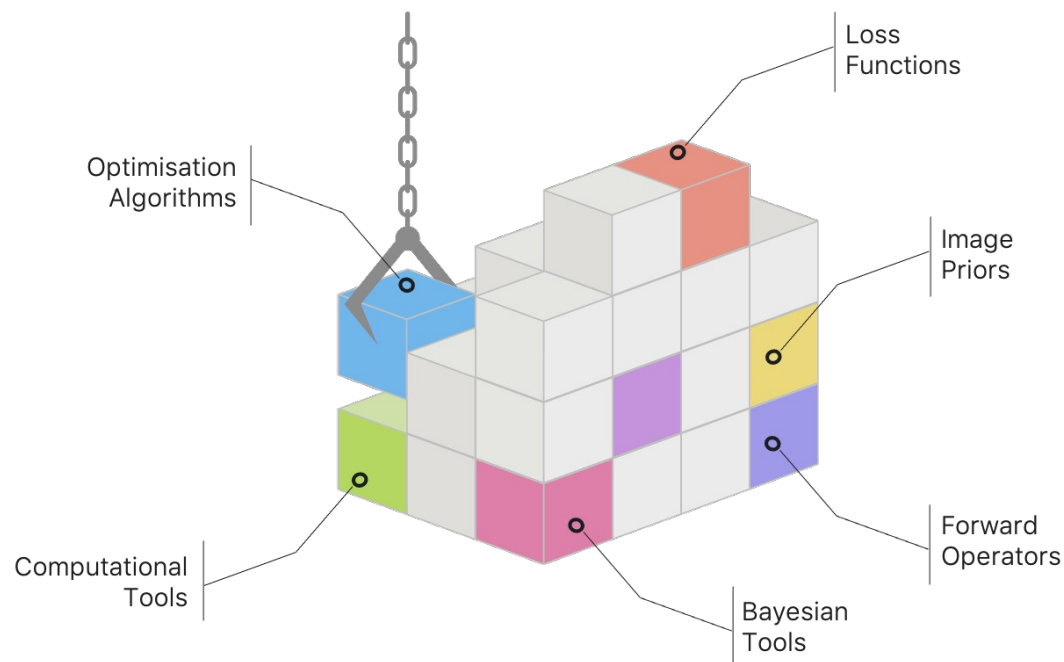
<https://pyxu-org.github.io/>

EPFL

Center
for Imaging



- Python library to design/deploy CI pipelines
 - CPU/GPU imaging operators (CT, Fourier, ...)
 - Reconstruction algorithms
 - Strong interoperability with ML ecosystem
 - Speed up R&D loop
- Share compute-part between domains and applications.



LenslessPiCam [Bezzam et al.]

Divergent X-ray tomography reconstruction and optimisation [Haouchat¹

An Angular Framework for Ultrasound Imaging [Hériard-Dubreuil]

Neural Manifolds Through the Lens of Connectome Spectral Analysis [Rue Queralt]

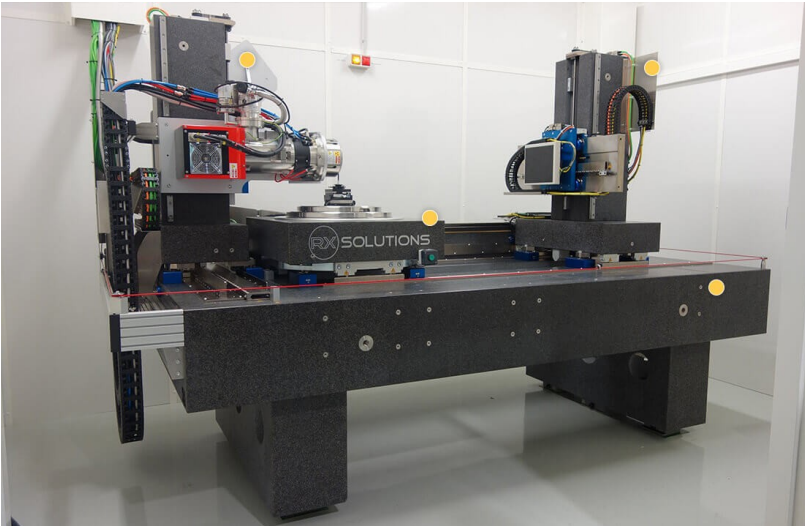
HVOX: Scalable Interferometric Synthesis and Analysis of Spherical Sky Maps [Kashani et al.]

PolyCLEAN: Atomic Optimization for Super-Resolution Imaging and Uncertainty Estimation in Radio Interferometry [Jarret et al.]

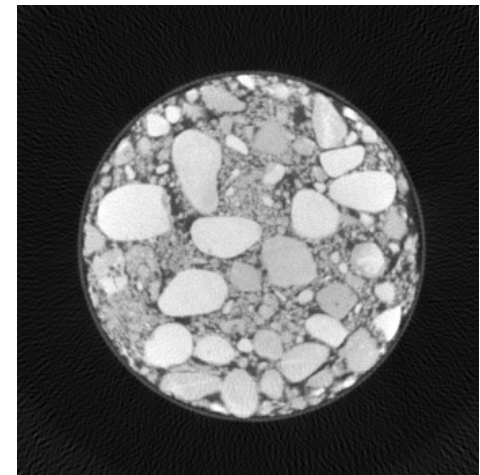


Example: CT Reconstruction

- High-resolution Cone-beam CT scanner.
- Goal: Speed up acquisition at equivalent reconstruction quality.



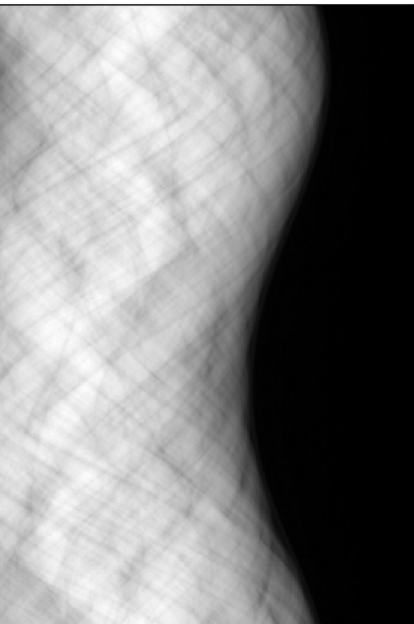
Device software



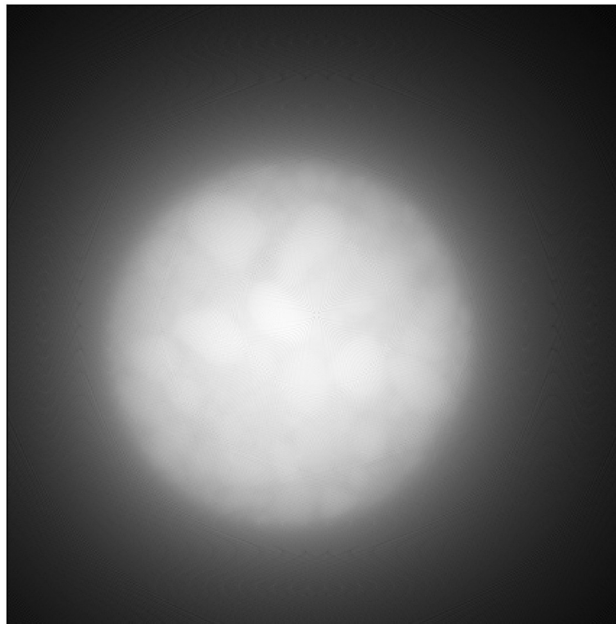
Match the device output 1

- Parse XML file → extract scan geometry.
- Build CT projector digital twin via Pyxu.

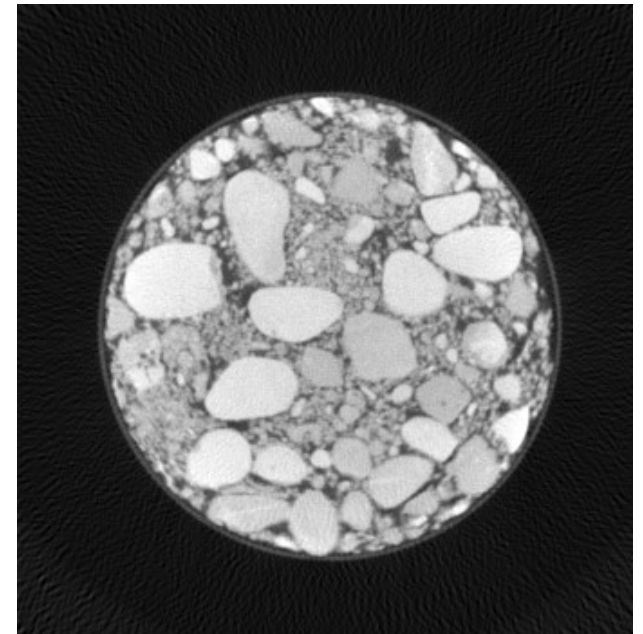
Raw data



$$\mathbf{x} = \mathbf{A}^* \mathbf{y}$$



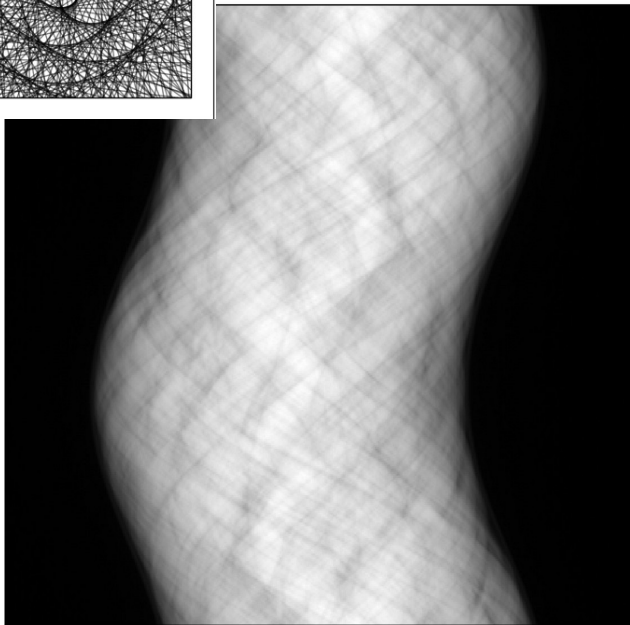
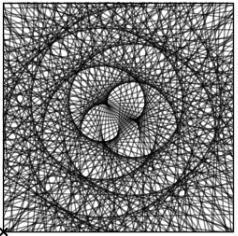
Device software



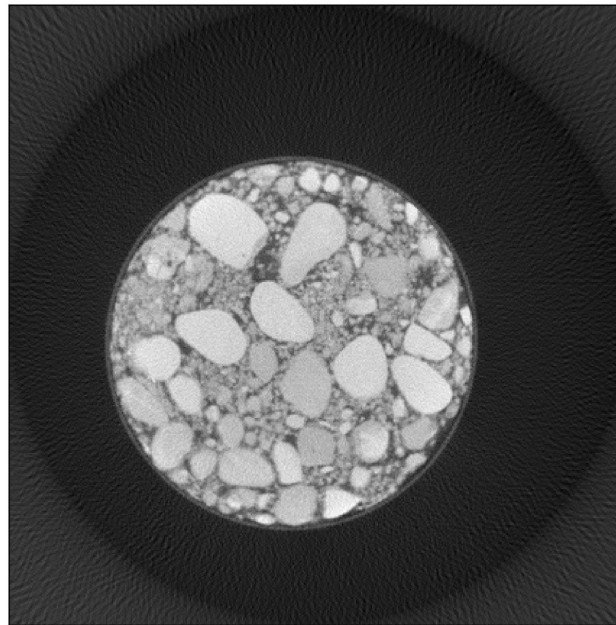
Match the device output 2

- Direct inversion: Filtered Back-Projection

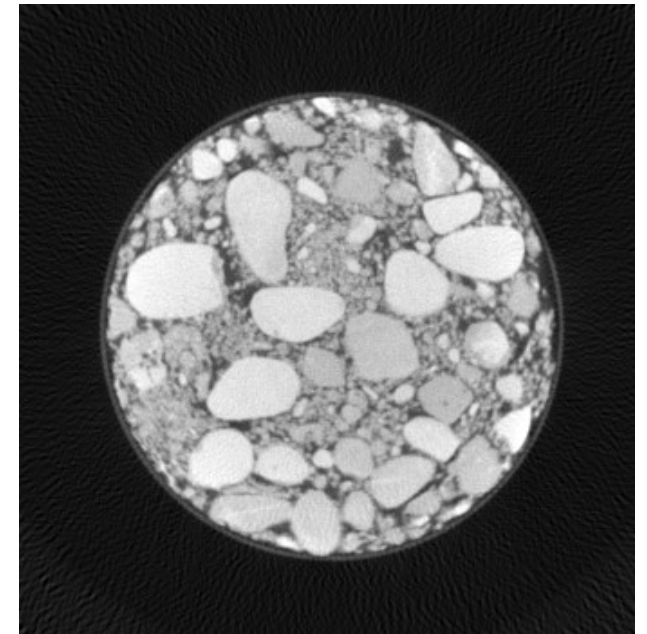
Raw data



$$\mathbf{x} = \mathbf{A}^{-1}\mathbf{y}$$



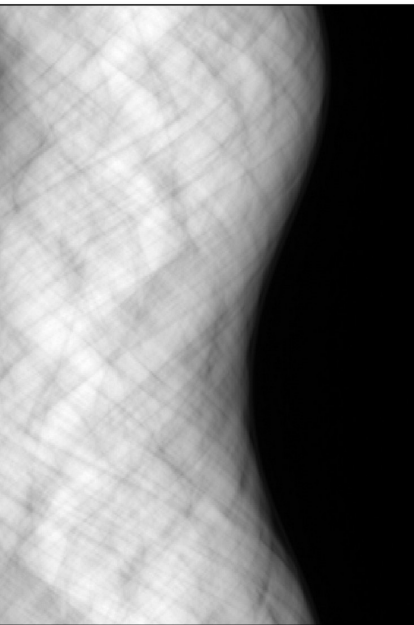
Device software



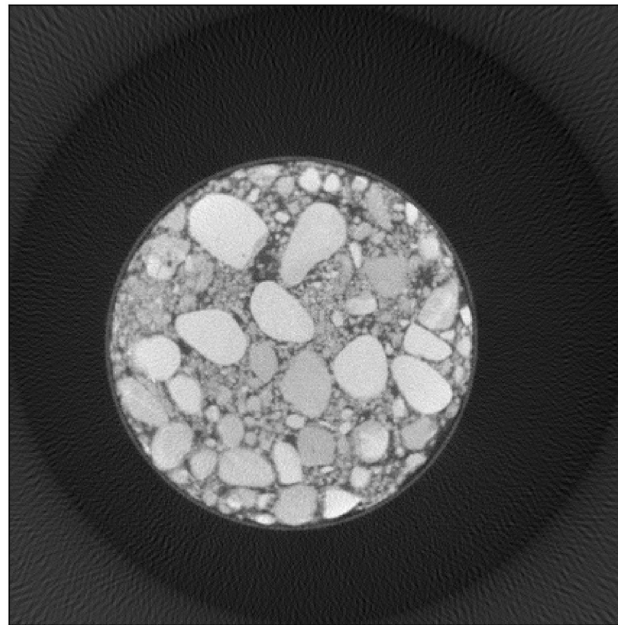
Direct Inversion with 25% data

- Speeds up acquisition
- Direct inversion breaks down

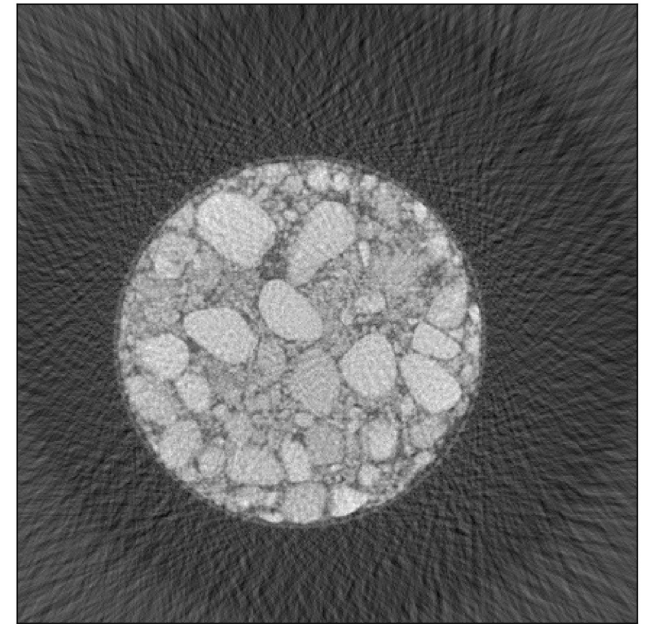
Raw data



$$\mathbf{x} = \mathbf{A}^{-1}\mathbf{y} \quad (\text{full data})$$

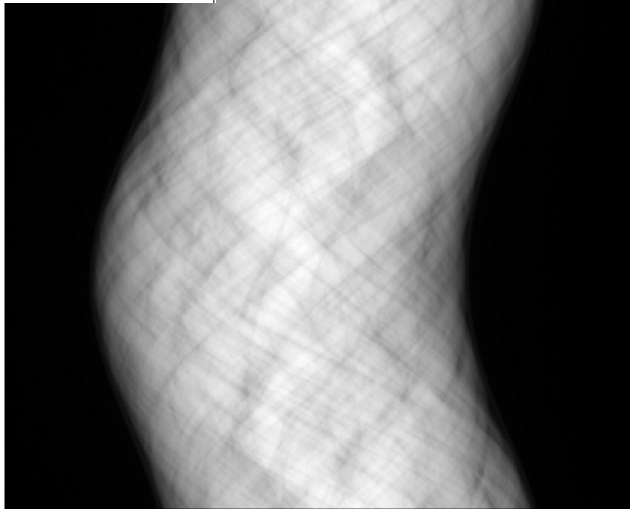
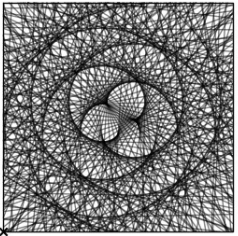


$$\mathbf{x} = \mathbf{A}^{-1}\mathbf{y} \quad (25\% \text{ data})$$

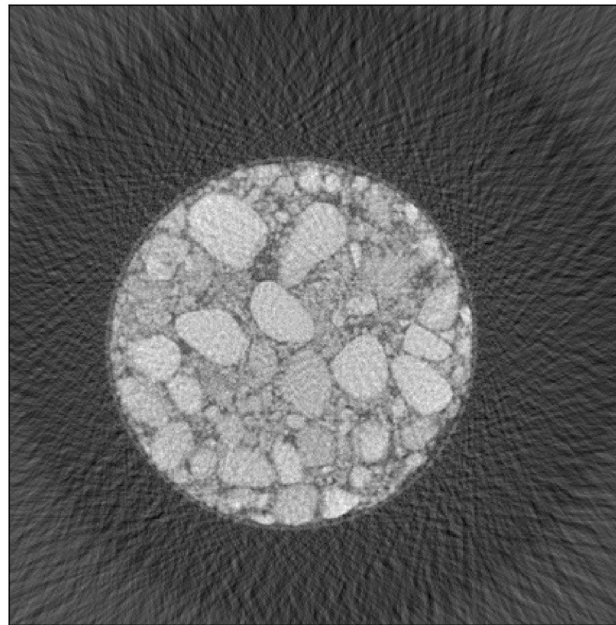


Model-Based Reconstruction 1

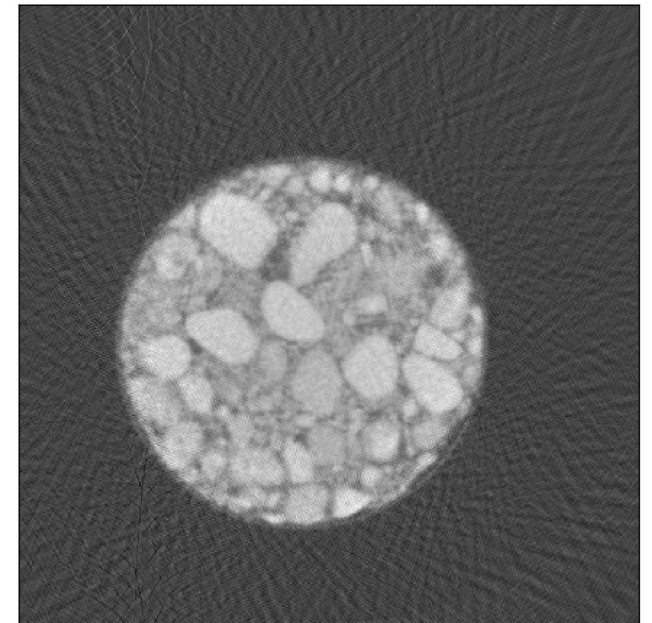
Raw data



$$\mathbf{x} = \mathbf{A}^{-1}\mathbf{y} \quad (25\% \text{ data})$$

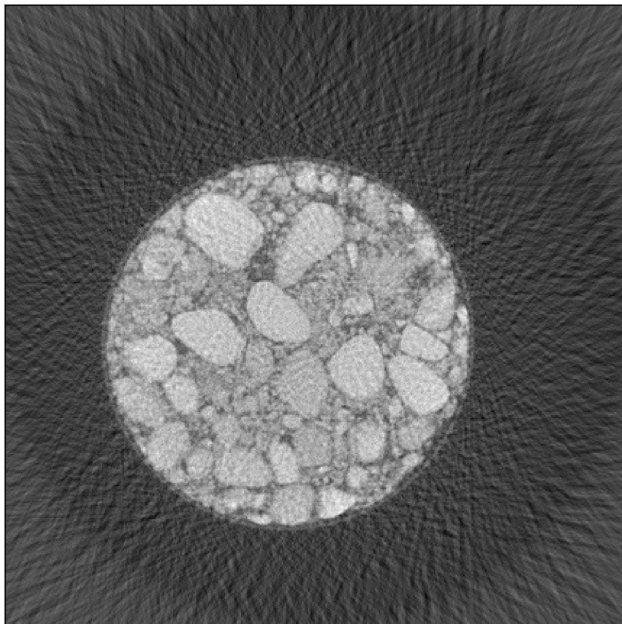


$$\arg \min_{\mathbf{z}} \|\mathbf{y} - \mathbf{A}\mathbf{z}\|_2^2 + \lambda \|\mathbf{z}\|_2^2$$

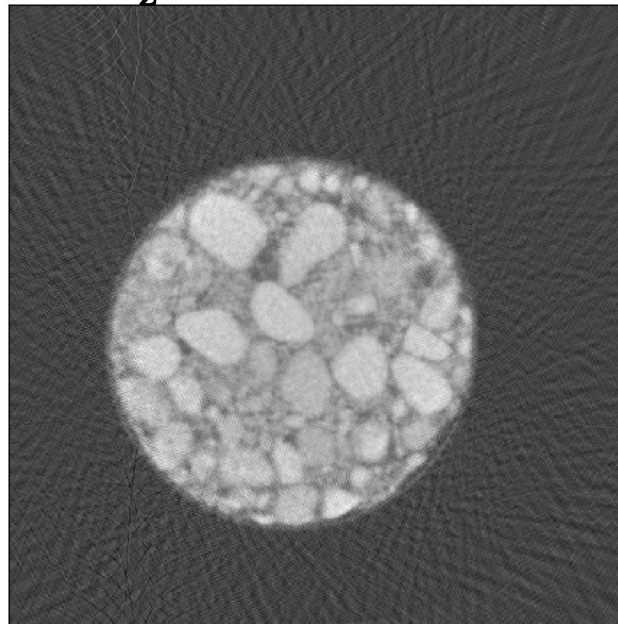


Model-Based Reconstruction 2

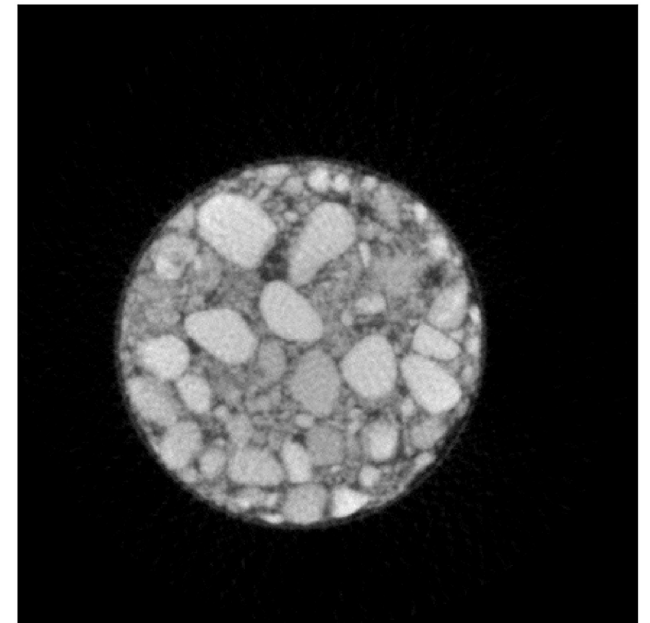
$$\mathbf{x} = \mathbf{A}^{-1}\mathbf{y} \quad (25\% \text{ data})$$



$$\arg \min_{\mathbf{z}} \|\mathbf{y} - \mathbf{A}\mathbf{z}\|_2^2 + \lambda \|\mathbf{z}\|_2^2$$

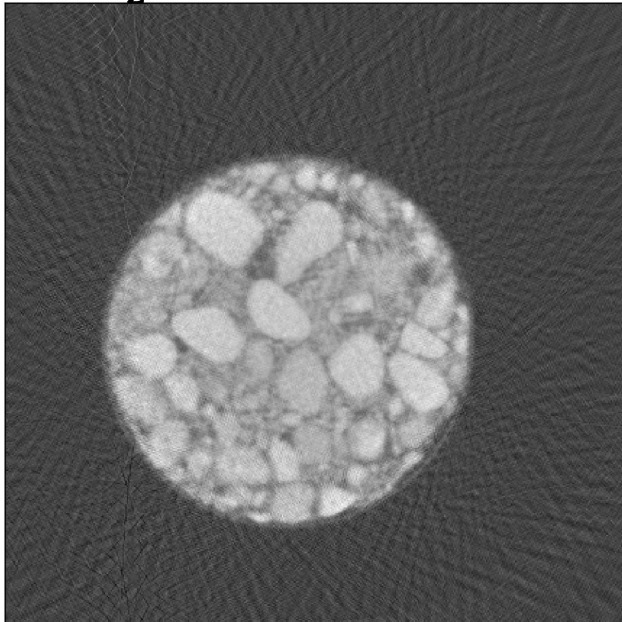


support constraint + positivity

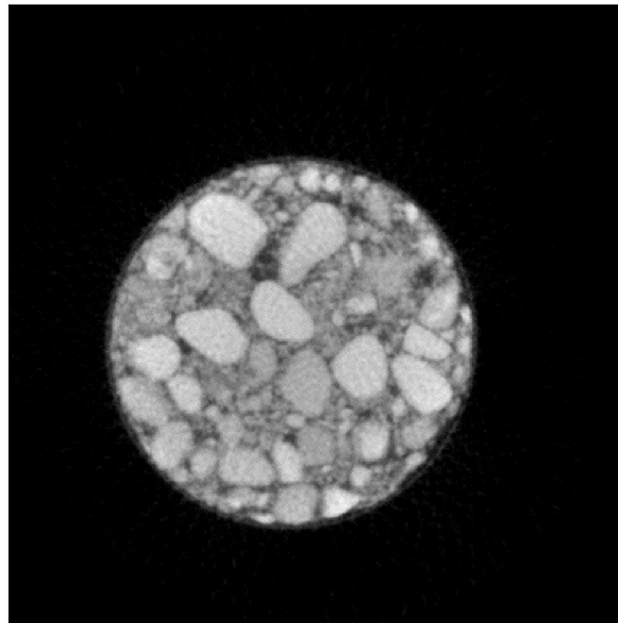


Model-Based Reconstruction 3

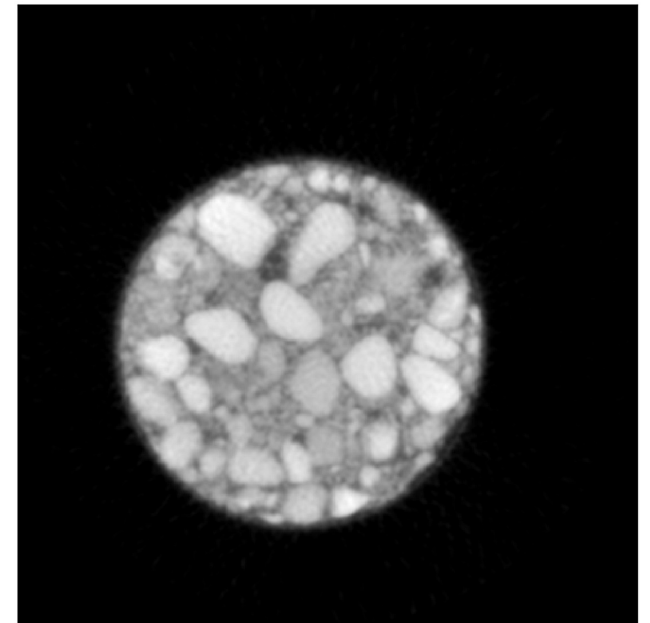
$$\arg \min_{\mathbf{z}} \|\mathbf{y} - \mathbf{A}\mathbf{z}\|_2^2 + \lambda \|\mathbf{z}\|_2^2$$



support constraint + positivity

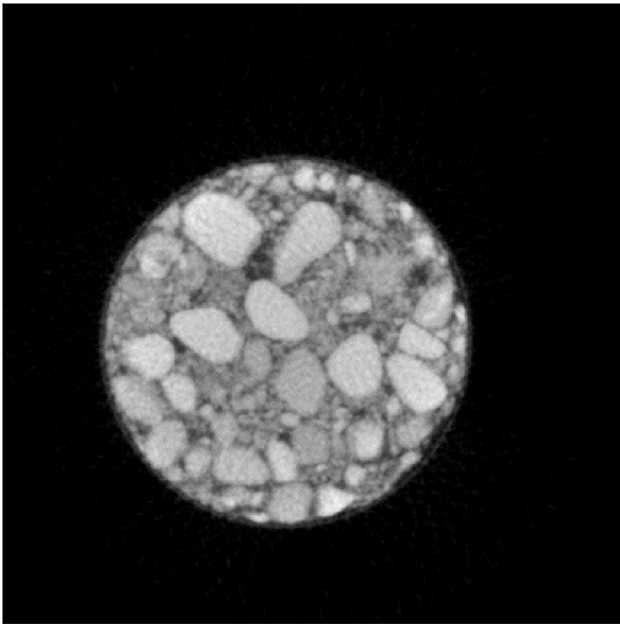


+ TV regularization

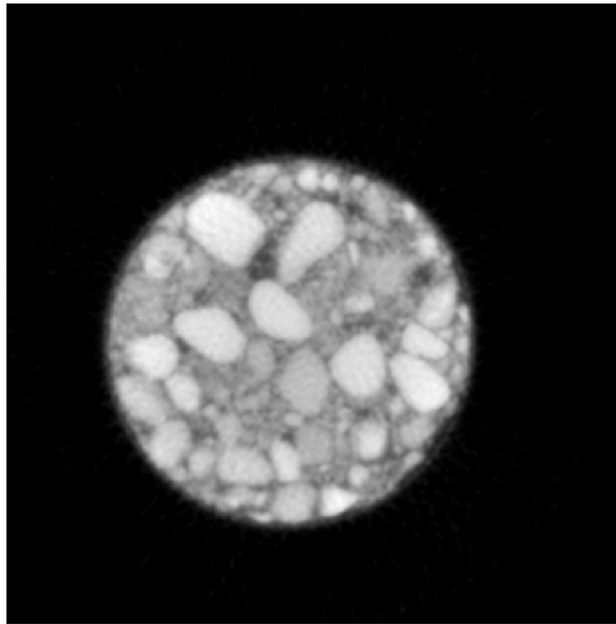


Model-Based Reconstruction 4

support constraint + positivity



+ TV regularization



+ Plug-and-play denoiser

