

Mathematics of Computational Imaging Systems

Forward model

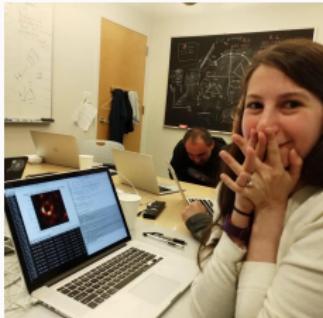
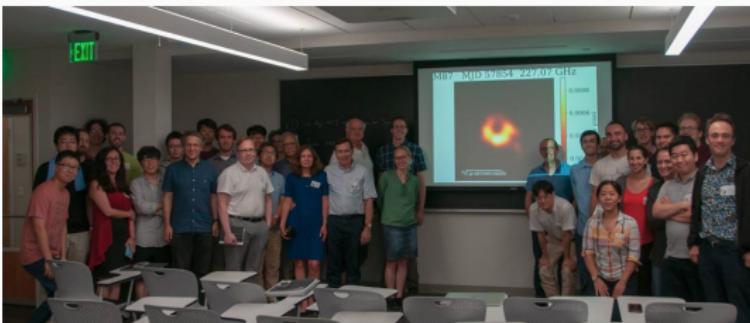
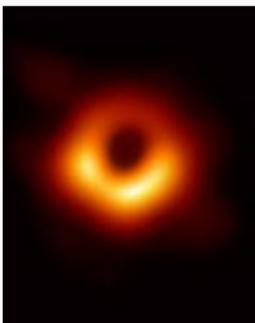
Nelly Pustelnik

CNRS, ENSL, Laboratoire de physique, F-69342 Lyon, France

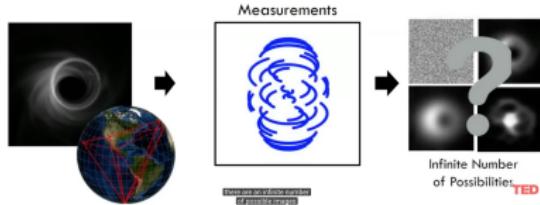


FONDATION
SIMONE ET CINO
DEL DUCA
INSTITUT DE FRANCE

Computational Imaging Systems: serving other sciences

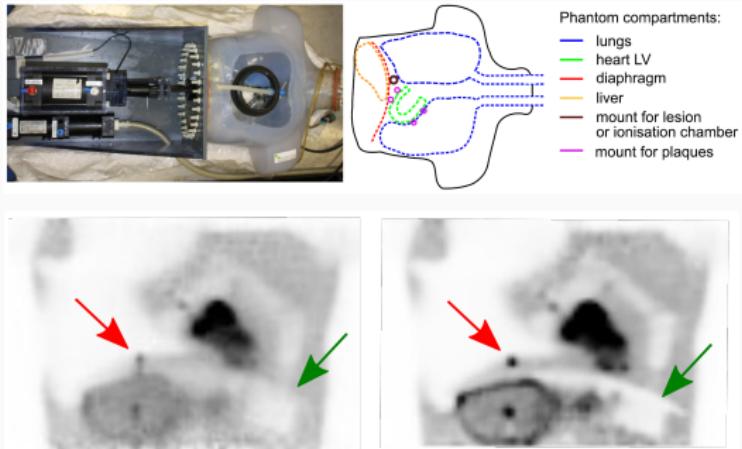


Reconstructing an Image



Black hole, Galaxy M87, Event Horizon Telescope (EHT)

Image analysis: serving other sciences



(source : F. Jolivet)

H2020 Nexis Project

Direct model

→ Variables of interest

- $z \in \mathbb{R}^M$: data/measurements.
- $\bar{x} \in \mathbb{R}^N$: unknown image.
- $\hat{x} \in \mathbb{R}^N$: estimated image.

→ Forward model:

$$z = \mathcal{D}(A\bar{x})$$

Stochastic degradation Linear operator

Direct model

→ Variables of interest

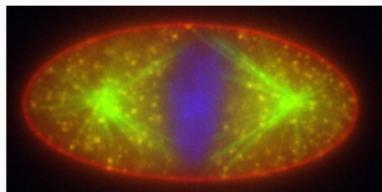
- $z \in \mathbb{R}^M$: data/measurements.
- $\bar{x} \in \mathbb{R}^N$: unknown image.
- $\hat{x} \in \mathbb{R}^N$: estimated image.

→ Forward model:

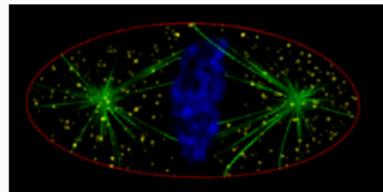
$$z = \mathcal{D}(A\bar{x})$$

Stochastic degradation **Linear** operator

© ISBI 2013 challenge



Degraded image z



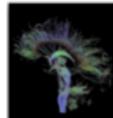
Original image \bar{x}

Direct model in medecine: MRI

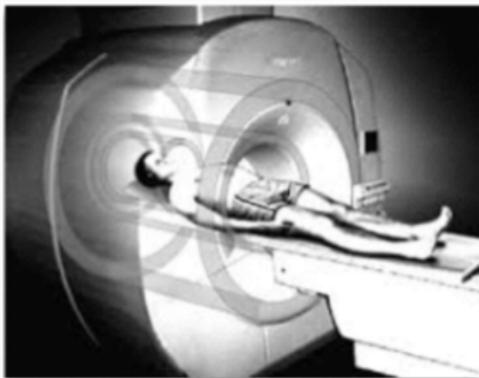
$$\mathbf{z} \simeq \mathbf{A}\bar{\mathbf{x}}$$



2D Structural MRI



6D Diffusion MRI

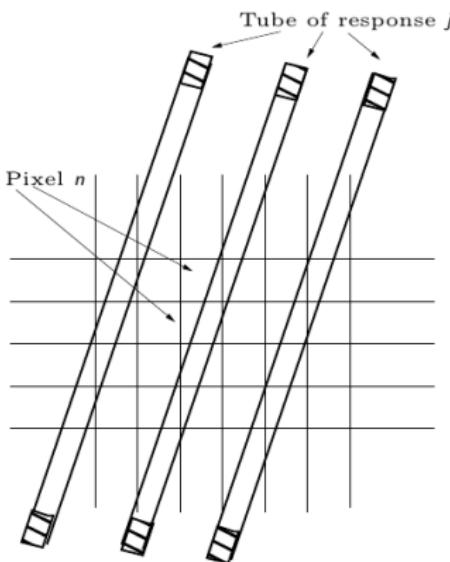


- $\bar{\mathbf{x}} \in \mathbb{R}^N$: vectorized original (unknown) image.
- $\mathbf{A} = \mathbf{MF}: \mathbb{R}^N \rightarrow \mathbb{C}^M$: measurement operator selecting (mask $\mathbf{M}: \mathbb{C}^N \rightarrow \mathbb{C}^M$) Fourier coefficients (2D Fourier transform $\mathbf{F}: \mathbb{R}^N \rightarrow \mathbb{C}^N$).
- \mathbf{z} : vector containing the observed values (undersampled Fourier coefficients).

Direct model in medicine: Tomography

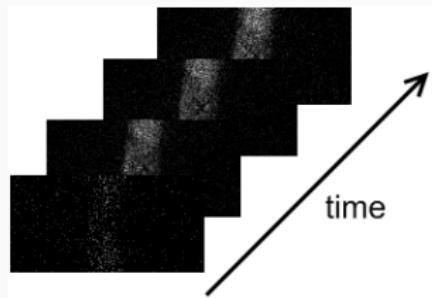


$$\mathbf{z} \simeq \mathbf{A}\bar{\mathbf{x}}$$

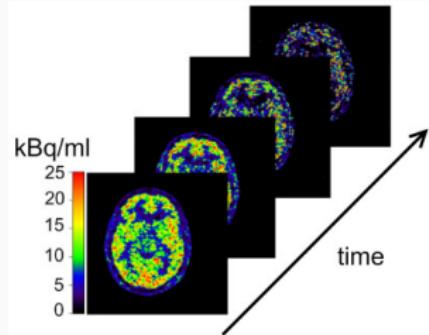
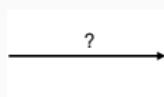


- $\bar{\mathbf{x}} = (\bar{x}_n)_{1 \leq n \leq N} \in \mathbb{R}^N$: vector consisting of the (unknown) values of the original image of size $N = N_1 \times N_2$.
- $\mathbf{A} = (A_{m,n})_{1 \leq m \leq M, 1 \leq n \leq N}$: probability to detect an event in the tube/line of response.
- $\mathbf{z} = (z_m)_{1 \leq m \leq M} \in \mathbb{R}^M$: vector containing the observed values (sinogram).

Application examples in medicine: Tomography

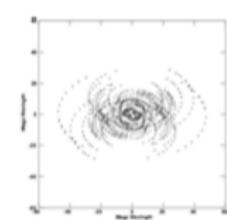


Degraded images



Reconstructed images

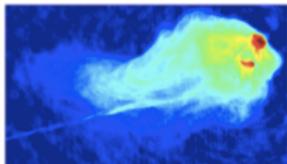
Direct model in Astronomy: Radio-interferometry



Fourier sampling



Very Large Array, New Mexico

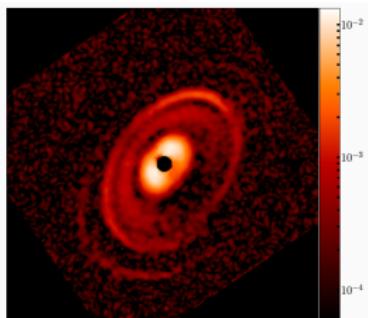
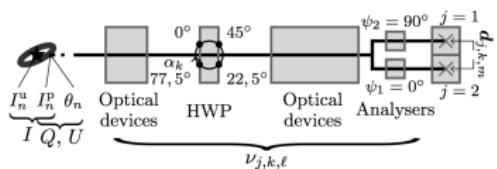


Cygnus A

$$\mathbf{z} \simeq \mathbf{A}\bar{\mathbf{x}}$$

- $\bar{\mathbf{x}} \in \mathbb{R}^N$: vectorized original (unknown) 2D image.
- $\mathbf{A} = \mathbf{GF}$: $\mathbb{R}^N \rightarrow \mathbb{C}^M$: measurement operator selecting Fourier coefficients.
 - $\mathbf{F}: \mathbb{R}^N \rightarrow \mathbb{C}^{\bar{N}}$: 2D Fourier transform (with zero-padding),
 - $\mathbf{G} \in \mathbb{C}^{M \times \bar{N}}$: (de)-gridding matrix modelling non-uniform (undersampled) Fourier transform, and direction (in)dependent effect (calibration artefacts).
- \mathbf{z} : vector containing the observed values (undersampled Fourier coefficients)

Application examples in astronomy: High-contrast imagery



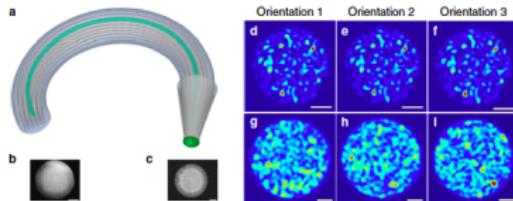
RXJ 1615 (Avenhaus et al. 2018)

$$z_{j,k} \simeq \underbrace{\sum_{m=1}^3 \nu_{j,k,m} T_{j,k} H S_m}_{A_{j,k}}$$

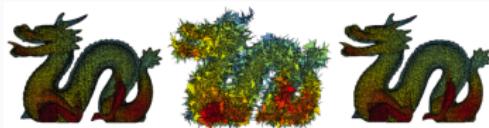
- $H: \mathbb{R}^N \rightarrow \mathbb{R}^N$: invariant blur.
- $T_{j,k}: \mathbb{R}^N \rightarrow \mathbb{R}^{M_j}$: geometric transform of the j -th polariser during the k -th acquisition.
- Stokes versus Jones formalisms:

$$\begin{aligned} I_{j,k}^{\text{det}} &= \frac{1}{2} I_u + I_p \cos^2(\theta - 2\alpha_k - \psi_j) \\ \Leftrightarrow I_{j,k}^{\text{det}} &= \nu_{j,k,1} I + \nu_{j,k,2} Q + \nu_{j,k,3} U \end{aligned}$$

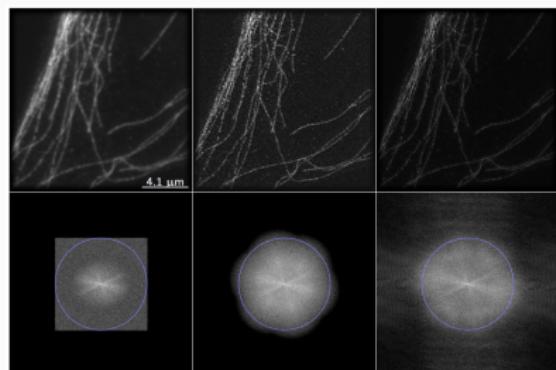
Many others!



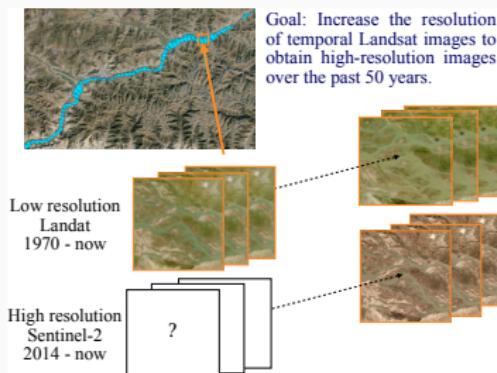
Photon imaging



3D Mesh denoising



Structured Illumination Microscopy



Remote sensing