



Jean-Luc Starck

<http://jstarck.cosmostat.org>





Radio-Interferometry & CosmoStat



• Radio Image Reconstruction

- H. Garsden et al, A&A, 575, A90, 2015.
- S. Farrens et al, Astronomy and Computing, 32, 2020.

UnivEarthS 2014-2017 (P.I. S. Corbel)
COSMIC 2016-2022 (CO-I J.L. Starck)

Julien Girard



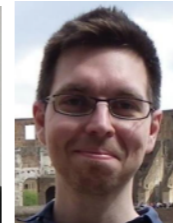
S. Corbel



H. Garsden



S. Farrens

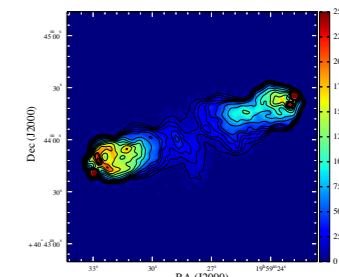
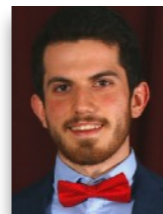


• Radio Weak Lensing

- F. Nammour, et al, "Galaxy Image Restoration with Shape Constraint", Journal of Fourier Analysis and Applications, 27, 88, 2021.
- F. Nammour et al, "ShapeNet: Shape Constraint for Galaxy Image Deconvolution", Astronomy and Astrophysics, 663, id.A692022, 2022.

ANR TOSCA 2023-2027(542k€, P.I. V. Pettorino)
ERA CHAIR TITAN 2023-2027 (2.5M€, P.I. J.-L. Starck)

Fadi Namour François Lanusse Julien Girard



• Transient sky

- B. Chiche et al, "Deep Learning-Based Deconvolution for Interferometric Radio Transient Reconstruction", **Astronomy and Astrophysics**, in press, 2023.

CIFRE 2019-2022
ARGOS 2023-2025 (2.8M€, co-I J.L. Starck)

Julien Girard Benjamin Chiche



• Epoch of Reionization (EoR)

- E. Chapman et al , "The Scale of the Problem : Recovering Images of Reionization with GMCA", MNRAS, 429, 2013.
- E. Chapman, et al., "Cosmic dawn and Epoch of Reionization foreground removal with the SKA", SKA science book, 2015.
- I. Carruci et al, "Recovery of 21-cm intensity maps with sparse component separation", 499, 2020.
- Y. Guimard et al , "Statistical and morphological component separation of foregrounds in convolved HI skymaps.", IEEE Workshop on Signal Processing, 2022

ERC Lena 2016-2021 (1.5M€, P.I. J. Bobin)
ERA CHAIR TITAN 2023-2027

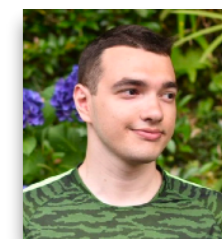
J. Bobin



Isabella Carucci



Y. Guimard



• Theory

- S. Casas et al, Linear and non-linear Modified Gravity forecasts with future surveys, Physics of the Dark Universe, Volume 18, 2023.
- S. Casas et al, Constraining gravity with synergies between radio and optical cosmological surveys, Physics of the Dark Universe, Volume 39, 2023

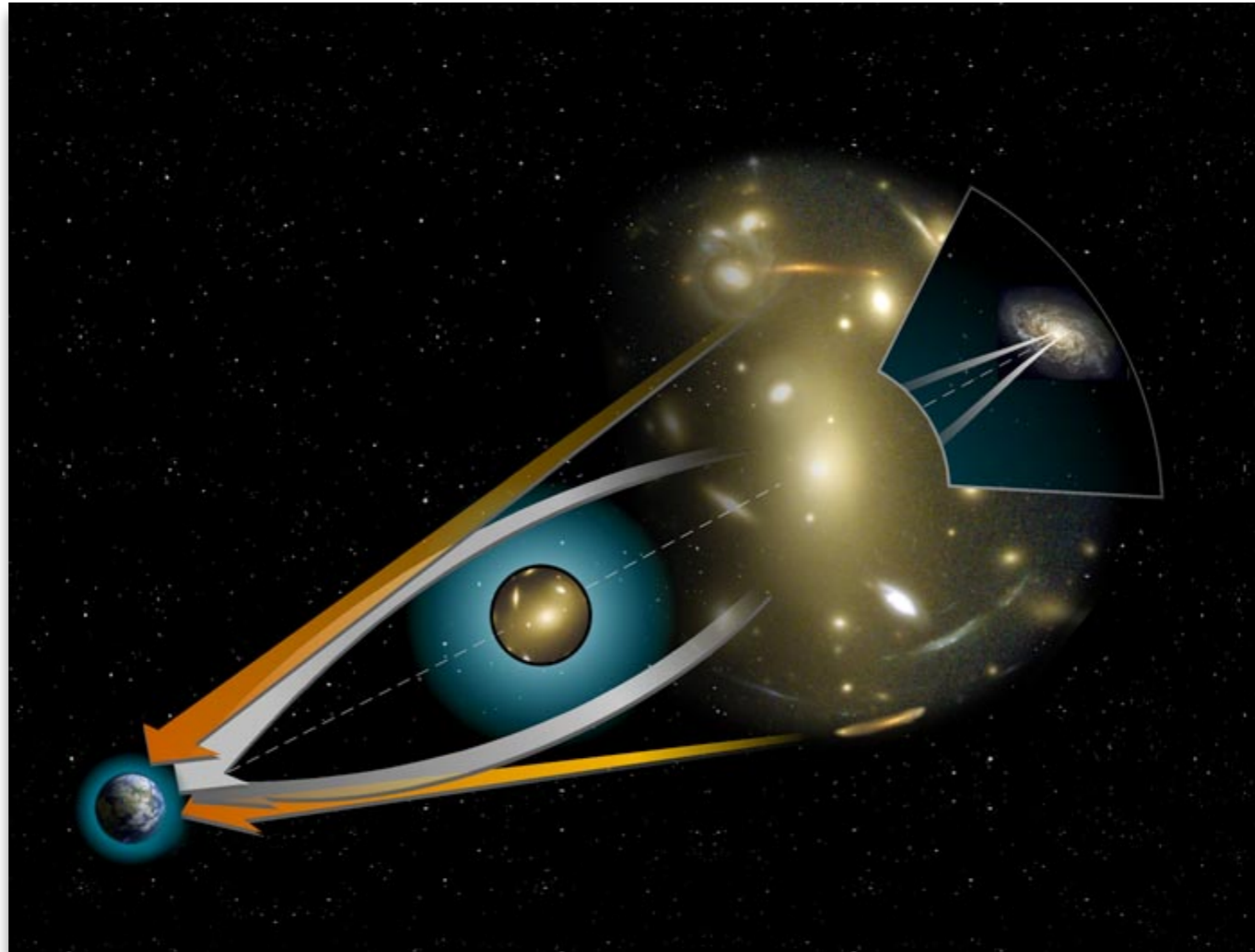
ANR TOSCA 2023-2027

V. Pettorino Santiago Casas



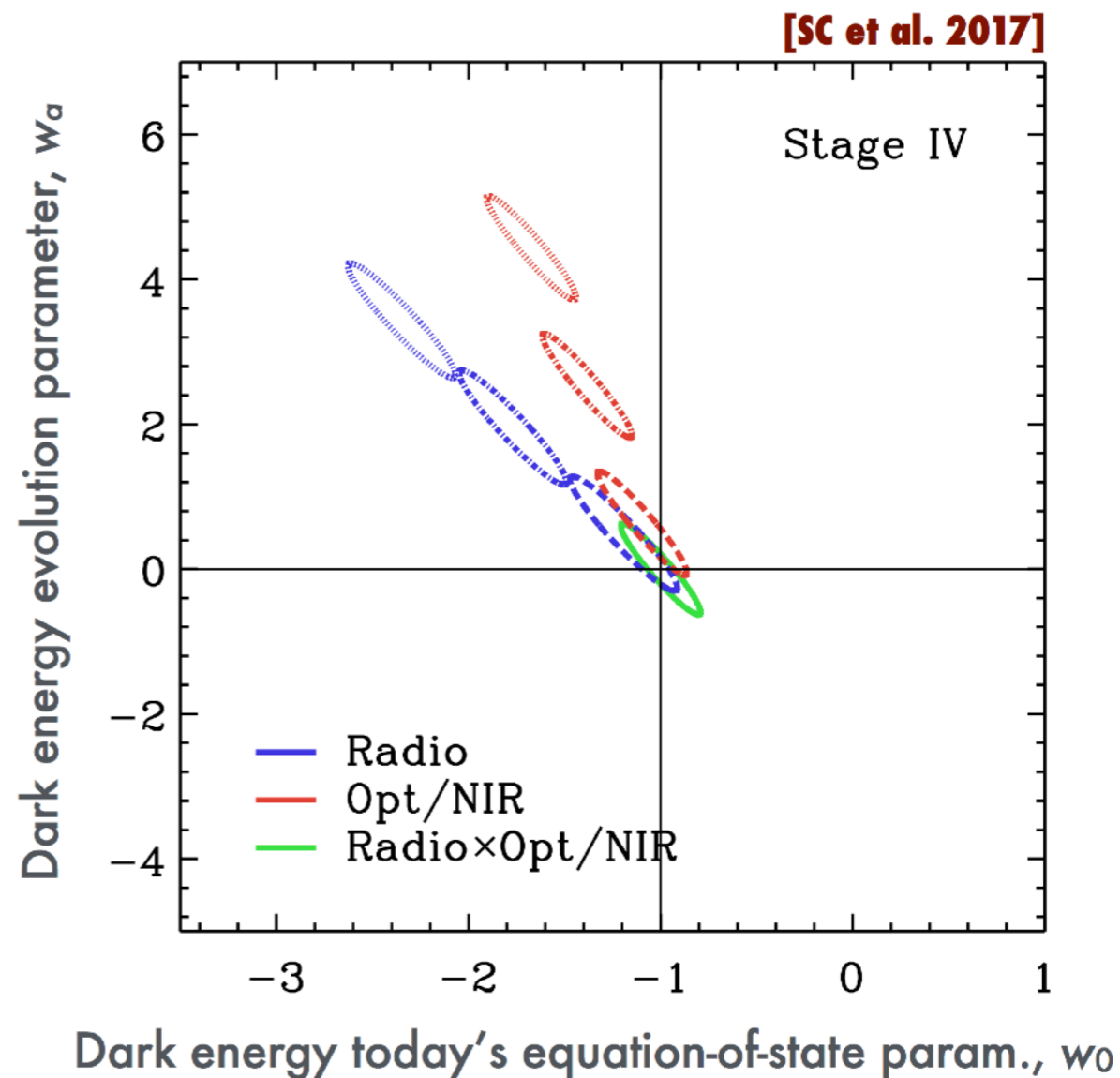
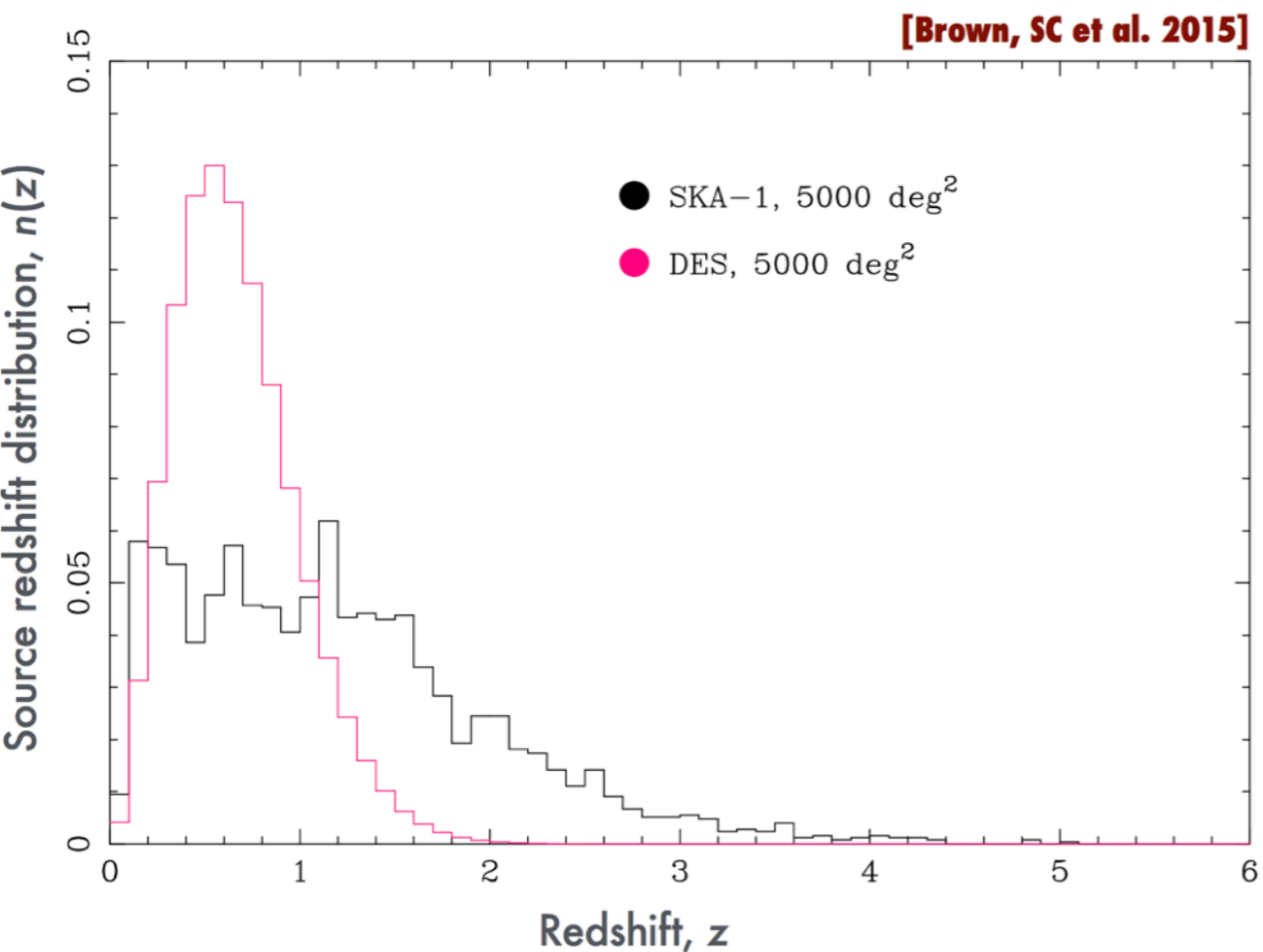


Dark Matter & Weak Lensing





Weak Lensing





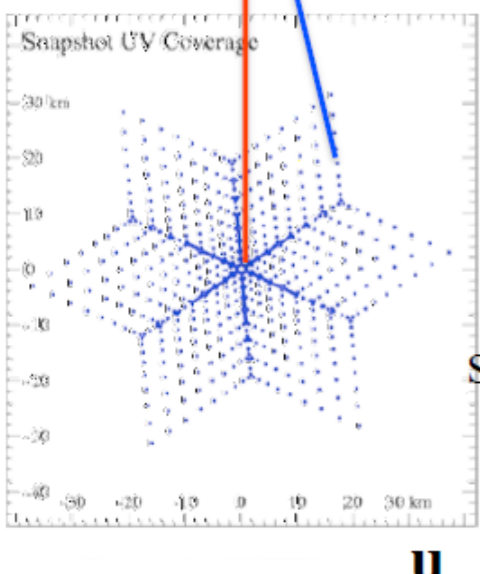
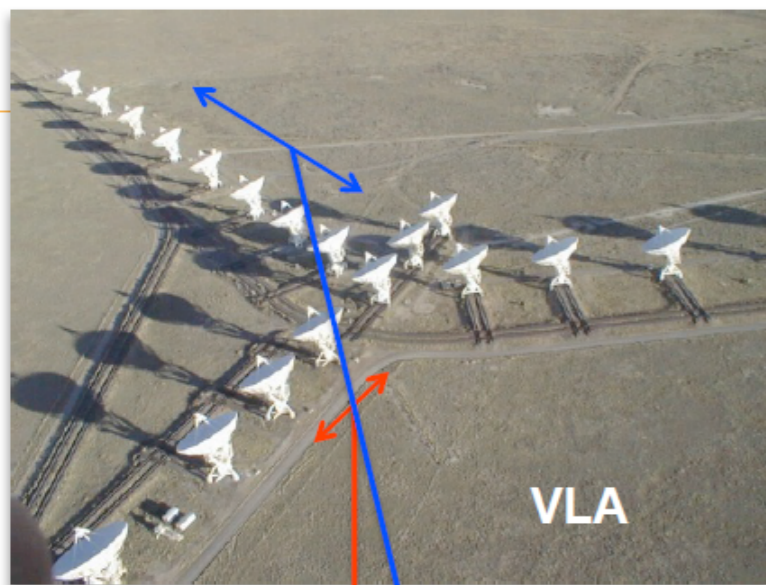
Detection + Classification stars/galaxies



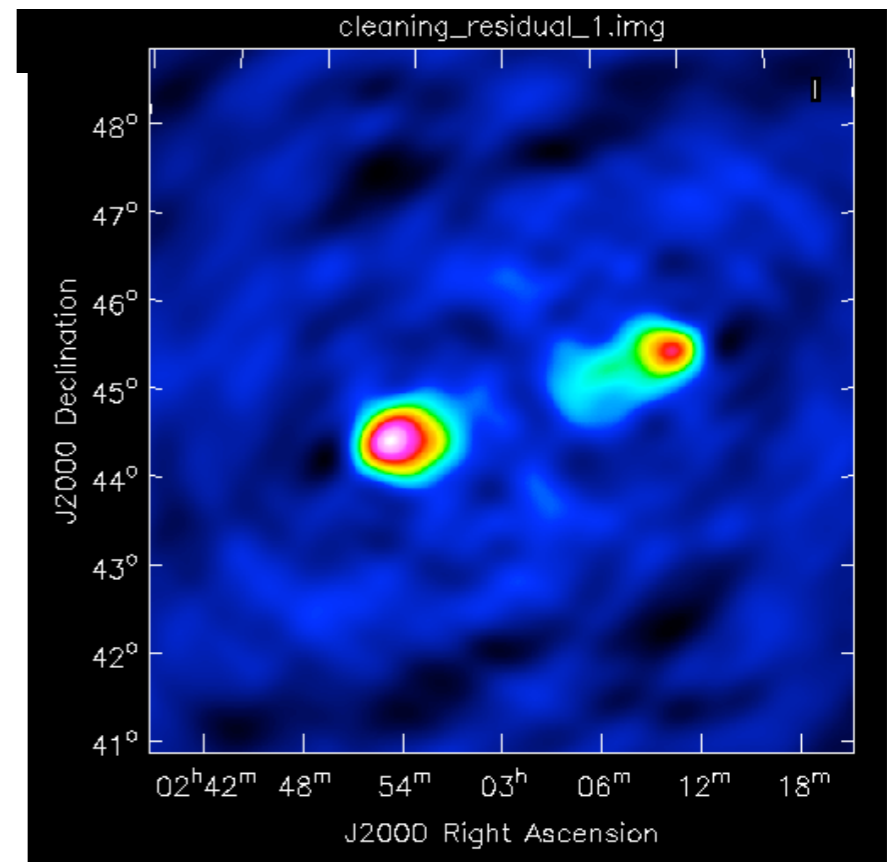
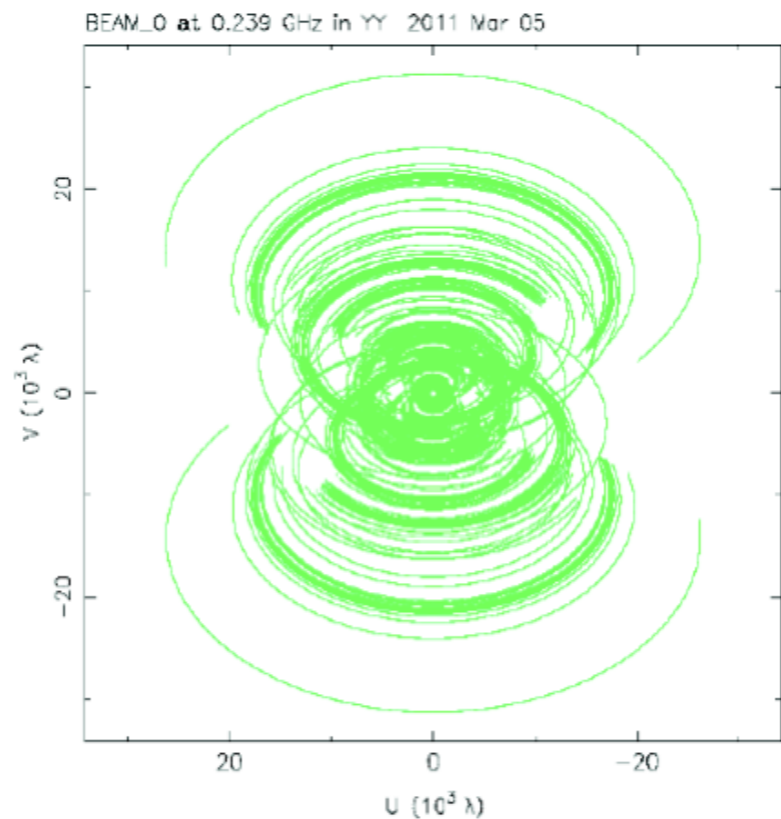
Galaxies

Stars





(u,v)
plane
sampling

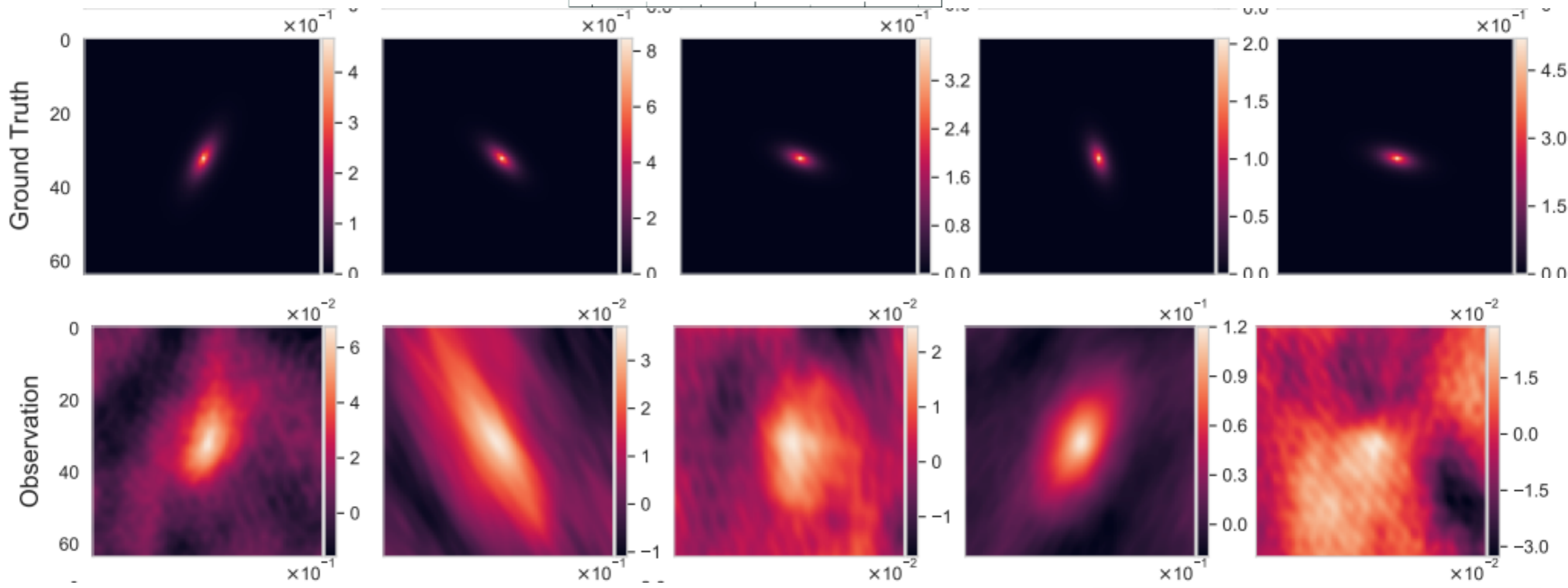
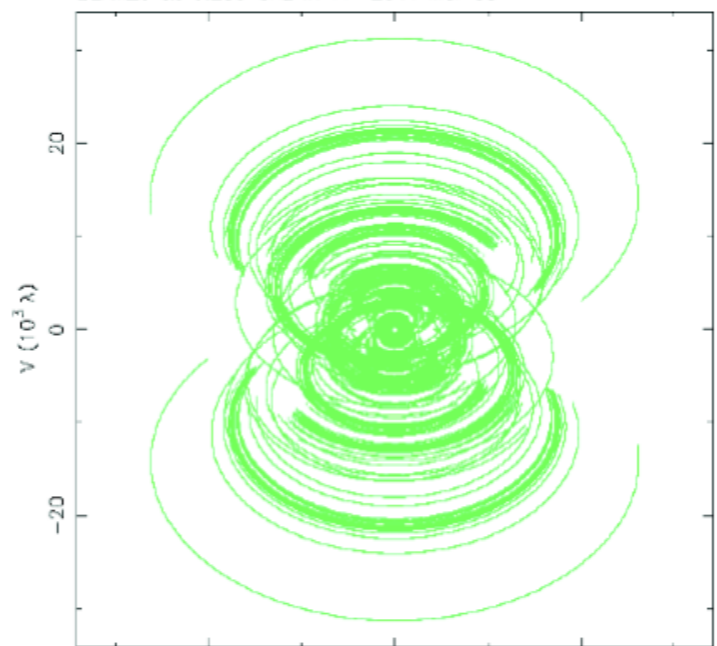




Radio Galaxies



BEAM_0 at 0.239 GHz in YY 2011 Mar 05





- **Reconstruct images from visibilities and applied standard methods used in optics**

=> iterative deconvolution methods produce images with structures in the residuals that dominate the cosmological signal (Patel et al, 2014,2015).

- **Fit a galaxy model directly to the visibilities (Rivi et al, 2016; Rivi et al, 2018)**

=> Limited by the source number density in the field of view, because of nearby galaxies residuals in the extraction (Rivi et al 2019).

- **Joint fitting of all galaxies directly to the visibilities using an Hamiltonian Monte Carlo (HMC) sampler (Rivi et al, 2019).**

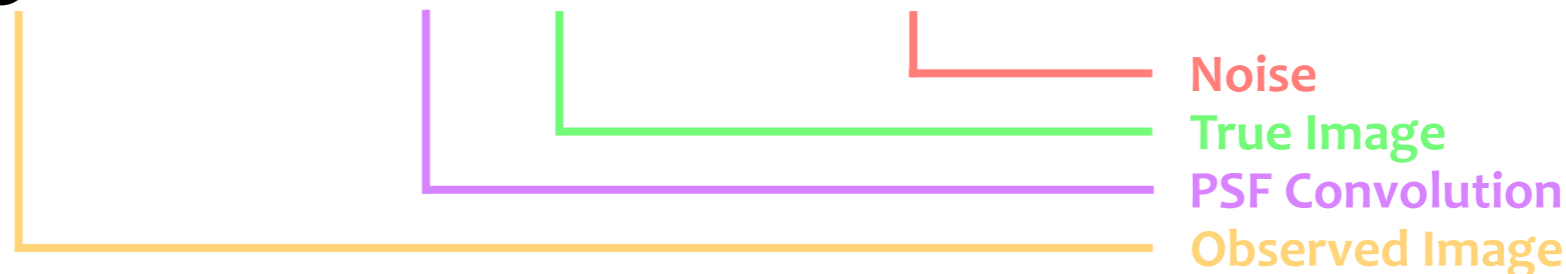
=> Seems promising, however much higher computational cost.



Deconvolution & Shape Constraint



$$\mathbf{y} = \mathbf{H}\mathbf{x} + \mathbf{n}$$



$$U_1 = \begin{pmatrix} 1 & 1 & \dots & 1 \\ 2 & 2 & \dots & 2 \\ \vdots & \vdots & \ddots & \vdots \\ n & n & \dots & n \end{pmatrix} \quad U_2 = \begin{pmatrix} 1 & 2 & \dots & n \\ 1 & 2 & \dots & n \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 2 & \dots & n \end{pmatrix} \quad U_3 = \begin{pmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{pmatrix}$$

$$U_4 = \begin{pmatrix} 1^2 + 1^2 & 1^2 + 2^2 & \dots & 1^2 + n^2 \\ 2^2 + 1^2 & 2^2 + 2^2 & \dots & 2^2 + n^2 \\ \vdots & \vdots & \ddots & \vdots \\ n^2 + 1^2 & n^2 + 2^2 & \dots & n^2 + n^2 \end{pmatrix} \quad U_5 = \begin{pmatrix} 1^2 - 1^2 & 1^2 - 2^2 & \dots & 1^2 - n^2 \\ 2^2 - 1^2 & 2^2 - 2^2 & \dots & 2^2 - n^2 \\ \vdots & \vdots & \ddots & \vdots \\ n^2 - 1^2 & n^2 - 2^2 & \dots & n^2 - n^2 \end{pmatrix} \quad U_6 = \begin{pmatrix} 1 & 2 & \dots & n \\ 2 & 4 & \dots & 2n \\ \vdots & \vdots & \ddots & \vdots \\ n & 2n & \dots & n^2 \end{pmatrix}$$

$$e_1(X) = \frac{\langle X, U_5 \rangle \langle X, U_3 \rangle - \langle X, U_1 \rangle^2 + \langle X, U_2 \rangle^2}{\langle X, U_5 \rangle \langle X, U_3 \rangle - \langle X, U_1 \rangle^2 - \langle X, U_2 \rangle^2}, \quad e_2(X) = \frac{2(\langle X, U_6 \rangle \langle X, U_3 \rangle - \langle X, U_1 \rangle \langle X, U_2 \rangle)}{\langle X, U_5 \rangle \langle X, U_3 \rangle - \langle X, U_1 \rangle^2 - \langle X, U_2 \rangle^2}$$

Statistical properties related to ellipticities are preserved through the blurring operation

$$\begin{bmatrix} \langle Y, U_1 \rangle \\ \vdots \\ \langle Y, U_6 \rangle \end{bmatrix} = \begin{bmatrix} \langle HX, U_1 \rangle \\ \vdots \\ \langle HX, U_6 \rangle \end{bmatrix}$$

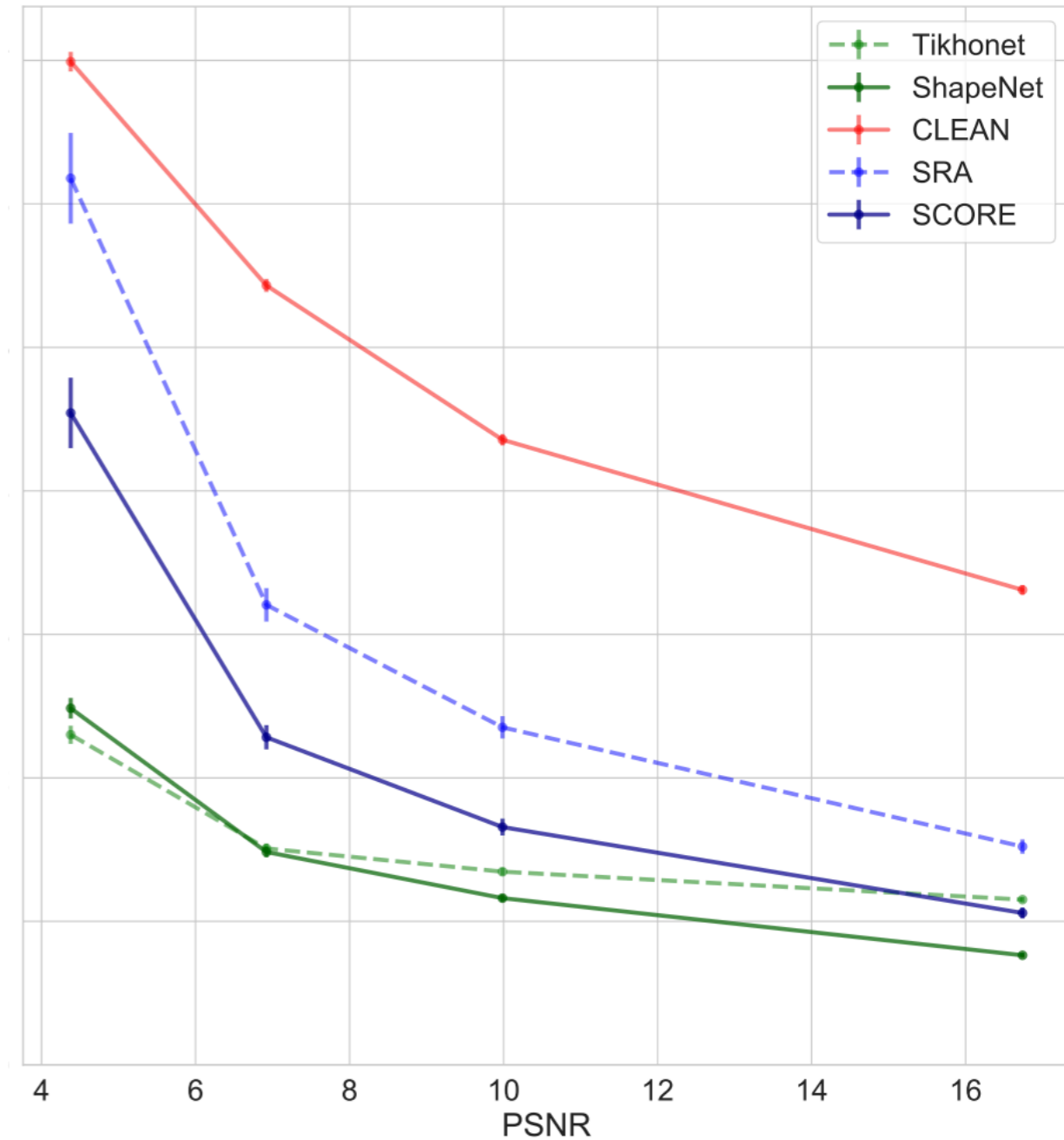
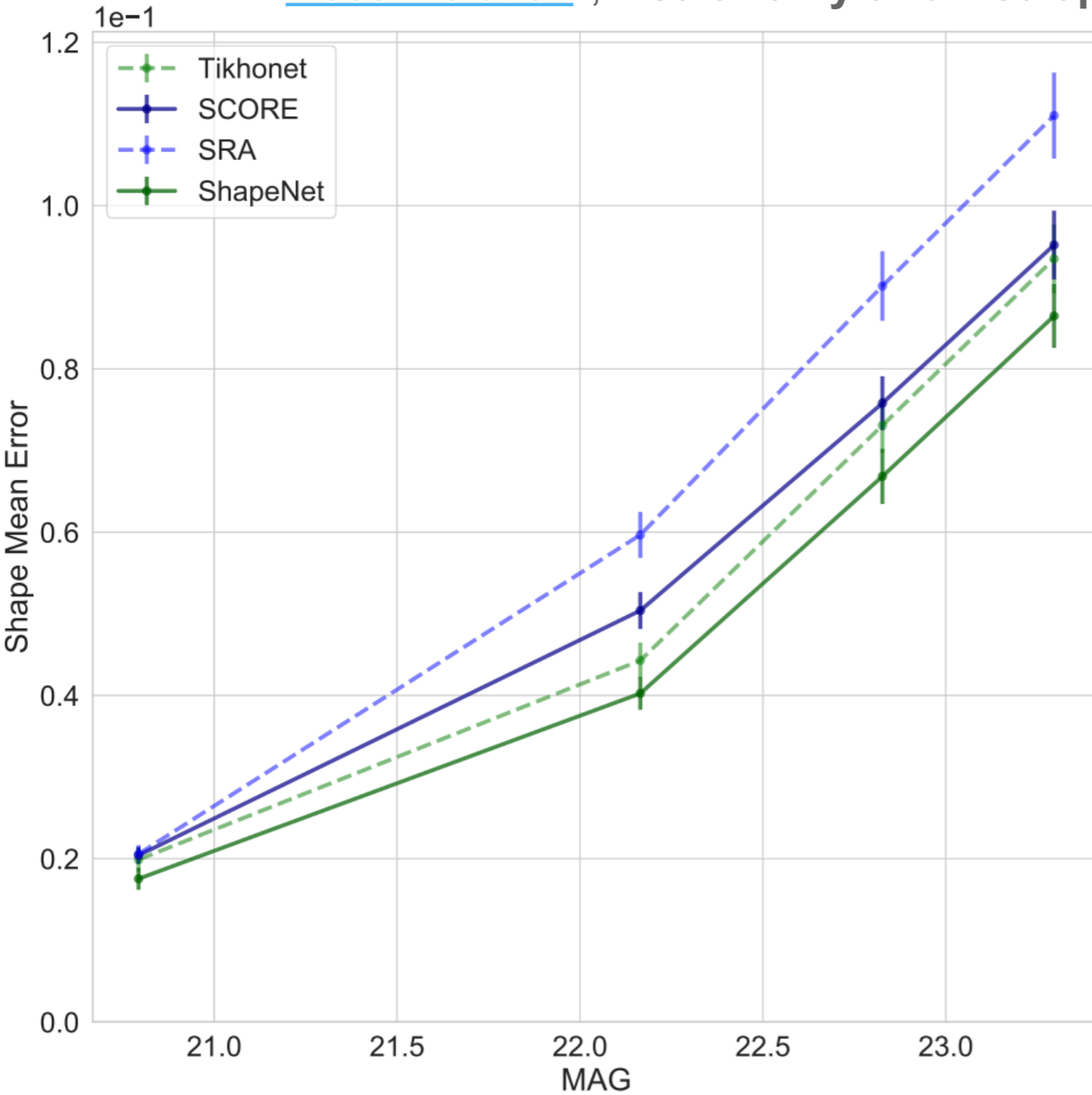
$$M(X) = \sum_{i=1}^6 \mu_i \langle X * H - Y, U_i \rangle^2$$



Deep Deconvolution



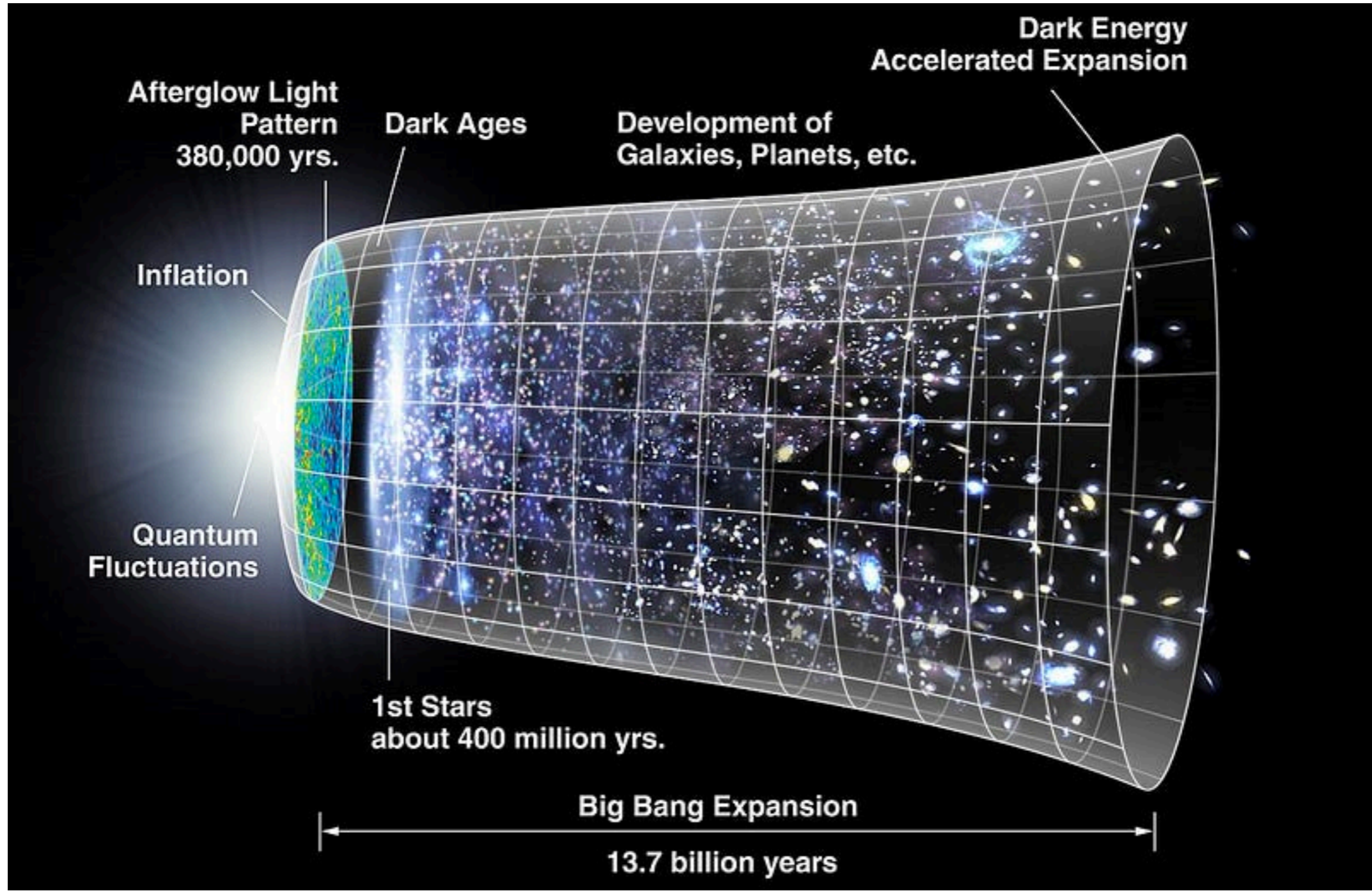
- F. Nammour et al, "[ShapeNet: Shape Constraint for Galaxy Image Deconvolution](#)", *Astronomy and Astrophysics*, 663, id.A692022, 2022.



(a) Mean relative error of flux.



Epoch of Reionization (EoR)

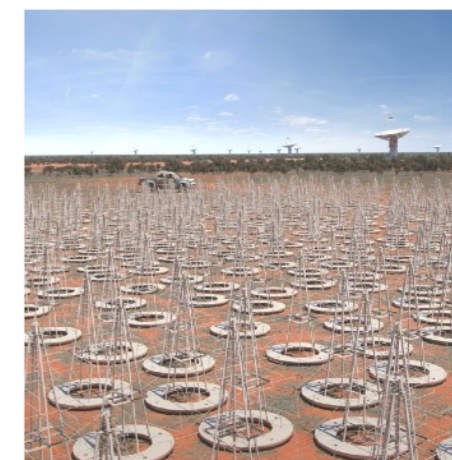
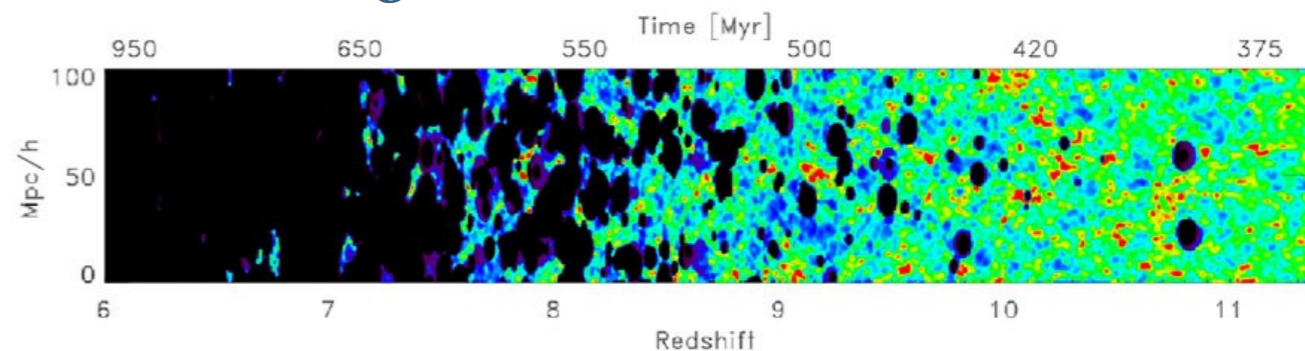
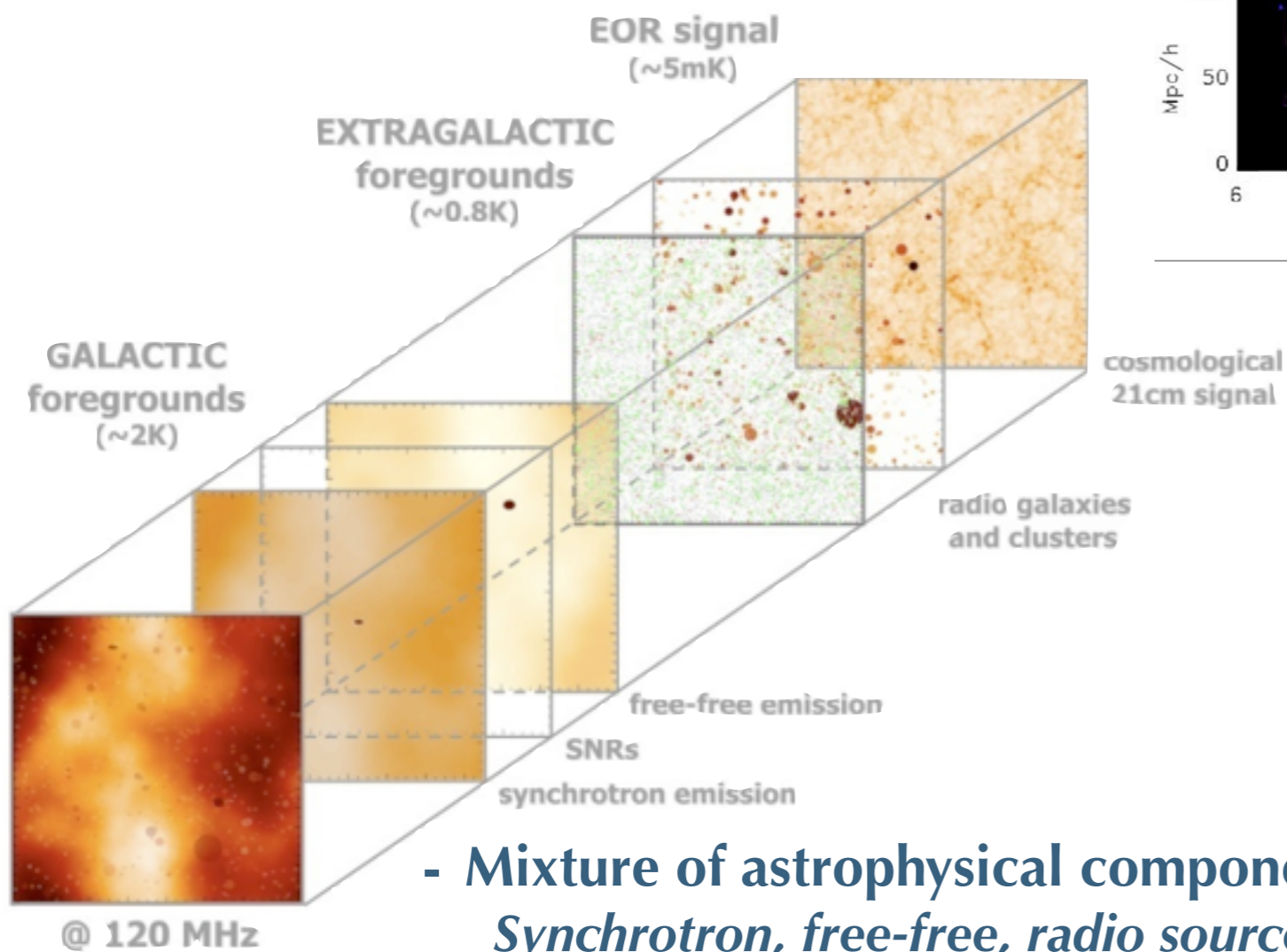




Epoch of Reionization (EoR)



- Imaging the dawn of the Universe at the **epoch of reionisation** ($6 < z < 12$)
- Mapping out the 21cm emission integrated over unresolved galaxies



SKA-low

Interferometric observations

- Mixture of astrophysical components:
Synchrotron, free-free, radio sources

- **Very low-level signal**

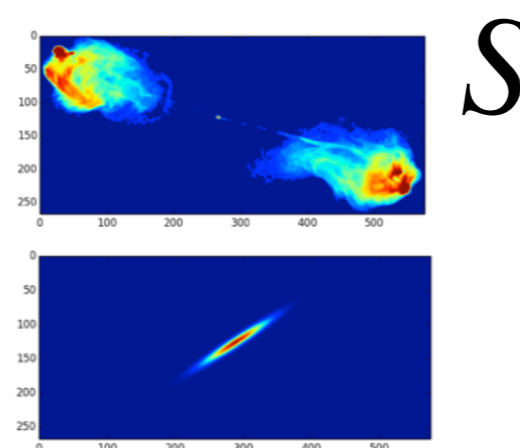
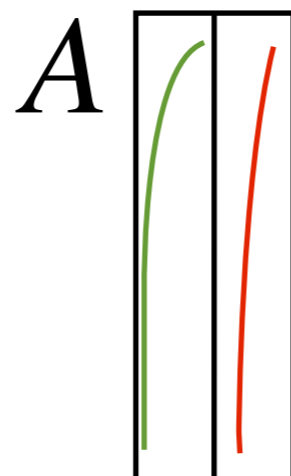
Foregrounds	>	Noise	>	EoR signal
$\sim 1 - 10^2$				$\sim 10^{-3}$



A Mixture & Convolution

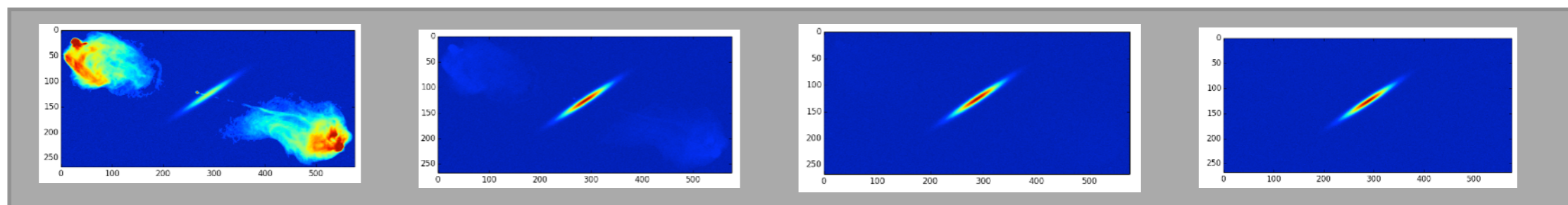


Ground Truth



Mixtures

$$X = AS$$



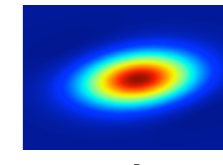
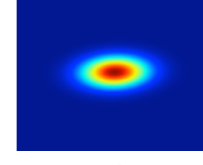
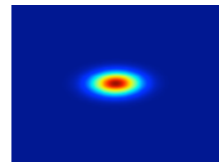
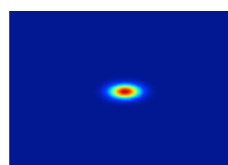
chan 1

chan 4

chan 7

chan 10

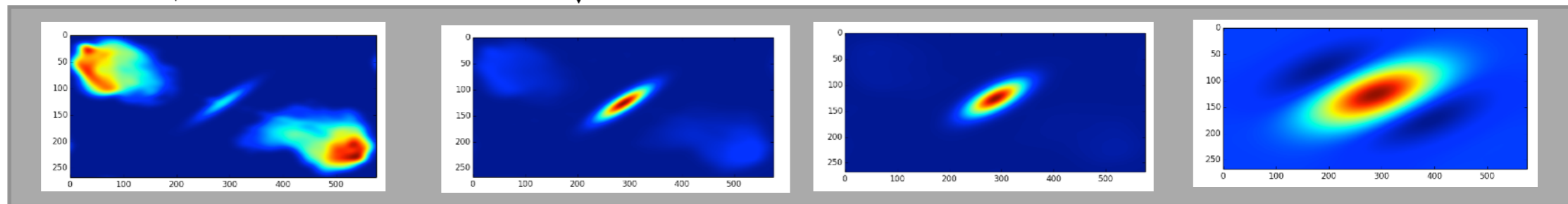
Beam H



*

Data

$$Y = HX + N$$



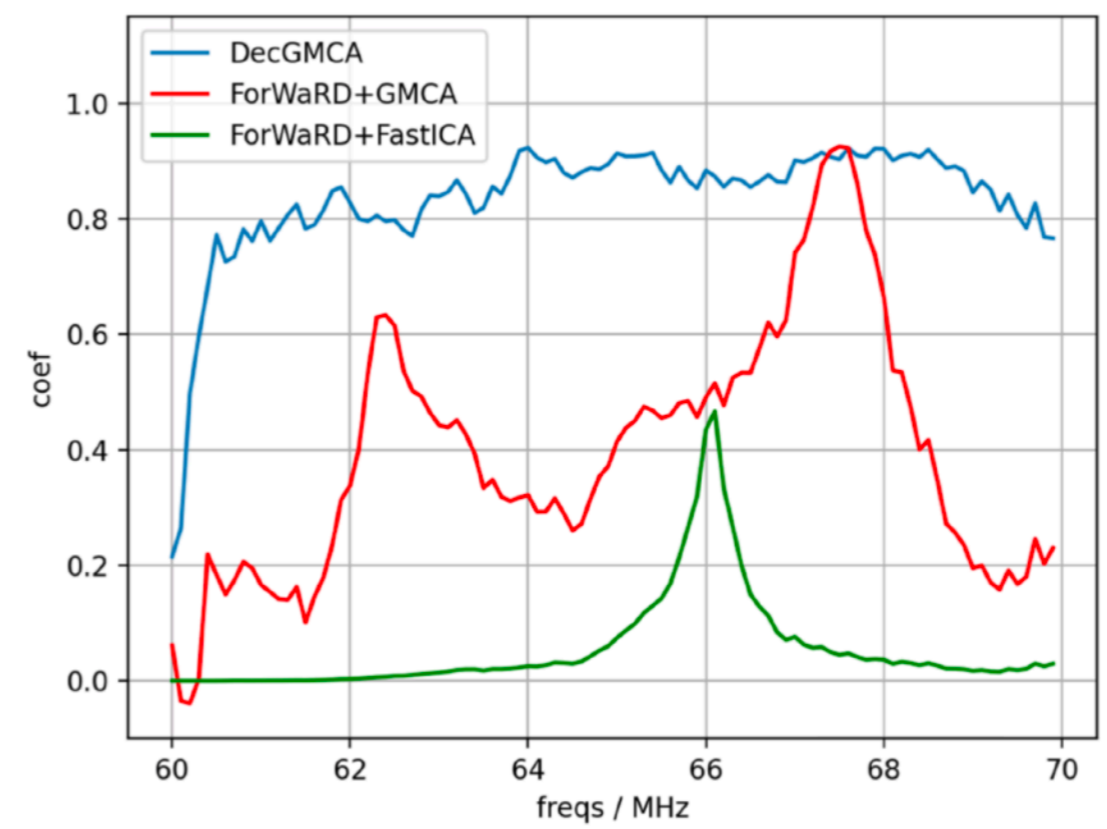
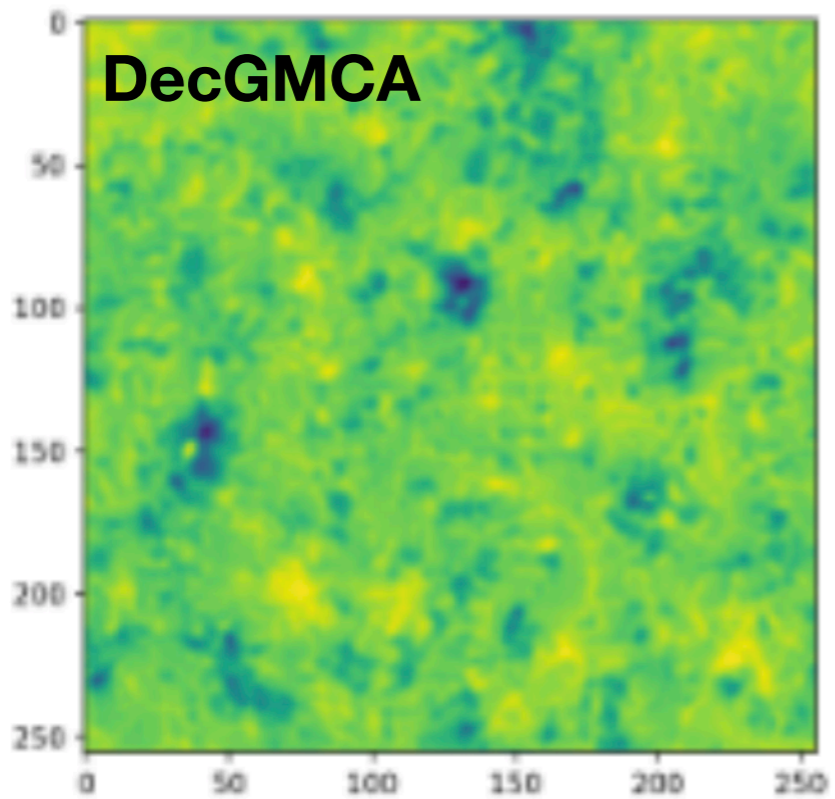
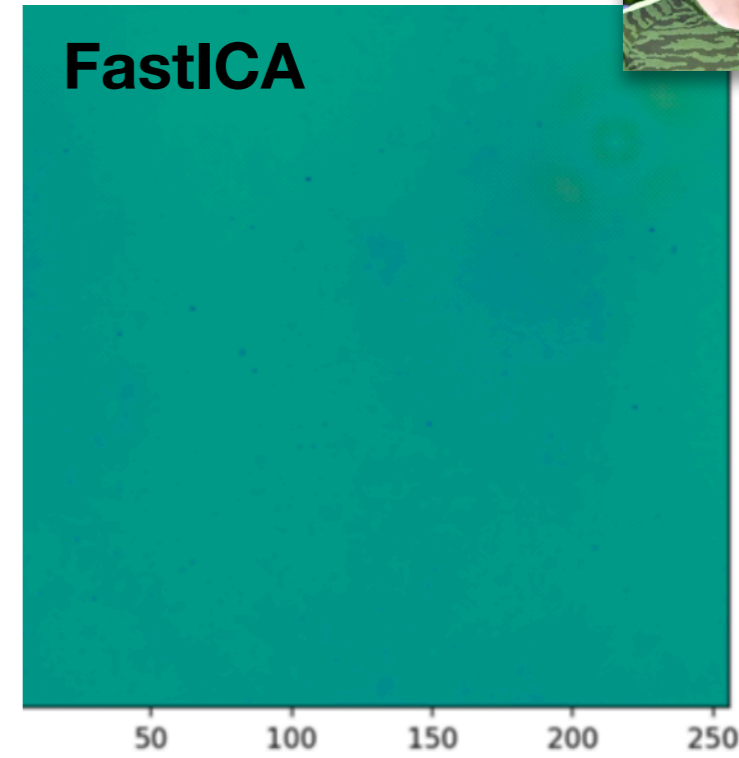
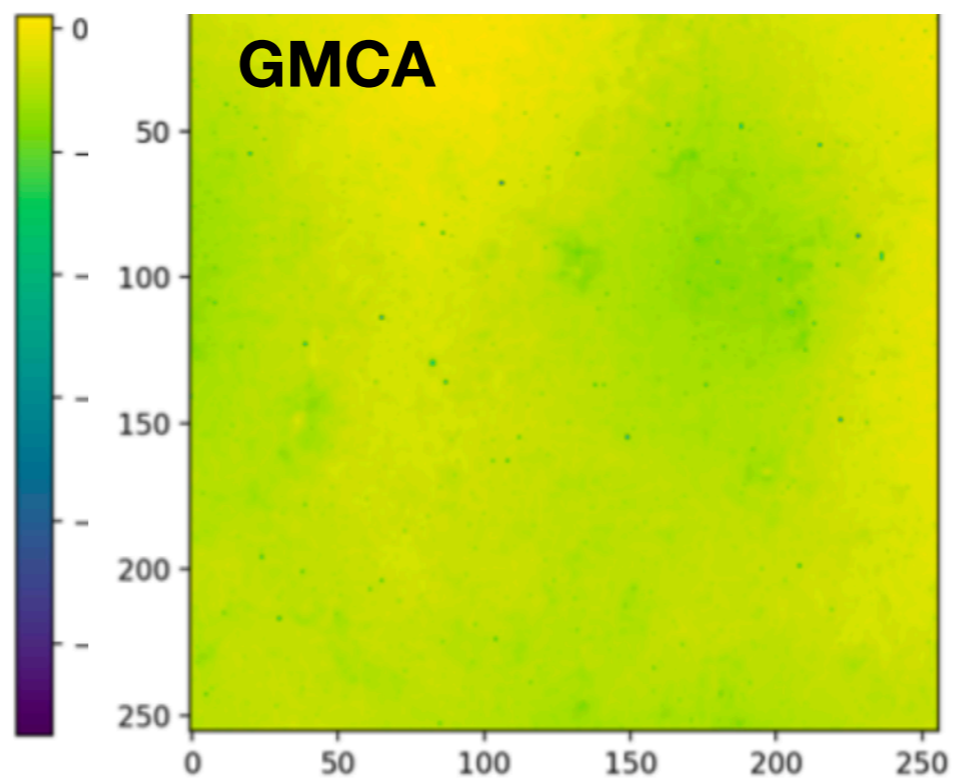
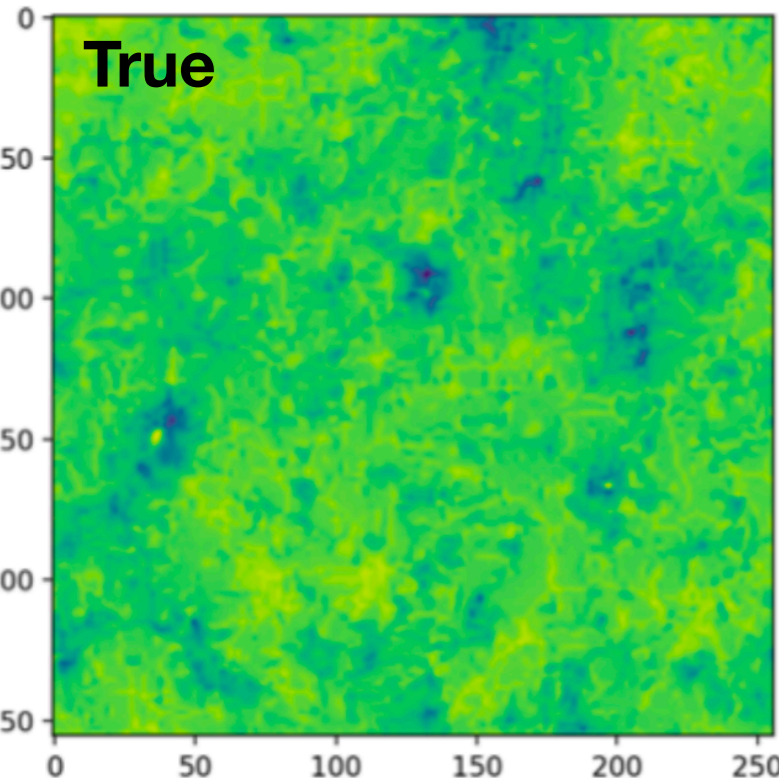
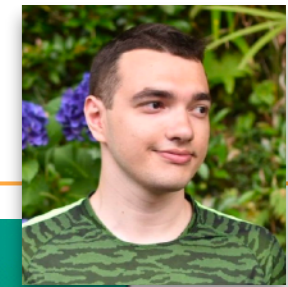
chan 1

chan 4

chan 7

chan 10

$$\forall i; \hat{y}_i = \hat{h}_i \left(\sum_j a_{ij} \hat{x}_j \right) + \hat{n}_i$$





Our SKA research Impact Outside Astrophysics



- Our Astronomical Radio Image Reconstruction paper has led to many collaborations outside astrophysics at CEA:

S. Farrens et al, "[PySAP: Python Sparse Data Analysis Package for Multidisciplinary Image Processing](#)", *Astronomy and Computing*, 32, pp 100402, 2020. DOI



Sam Farrens, DAp



Philippe Ciuciu,
Neurospin, CEA

- DRF Impulsion **COSMIC** (DRF, PI: Ph. Ciuciu, 200 k€) → **Neurospin MRI Imaging**
- DRF Impulsion **Fast FIB-SEM** (DRF Inac, PI: PH Jouneau, 70 k€) → **Electronic Microscopy**
- PTC **ComSET**: (DRT/LETI, PI: Zineb Saghi, 70 k€) → **Spectroscopic Electron Tomography**
- PTC **SILICOSMIC** (DRF, PI: Ph. Ciuciu, 60 k€) → COSMIC's Follow'up
 - › NeuroSpin
 - › CosmoStat
 - › Maison de la Simulation (Pierre Kestener)
- PTC Instrumentation et Détection **CROCUS** (DRF, PI: Ph. Ciuciu, 100 k€) → capteur ultrasons multiélément pour des applications de Contrôle Non Destructif (CND)



SKA Data Analysis will push knowledge on

- **Big Data, HPC, Optimization techniques, etc**
- **3D Hyperspectral + Time (flux video hyperspectral massif)**
- **Machine Learning in high Dimension**
- **Compressed Sensing applications**

SKA Data Interpretation will push knowledge on

- **EoR, Large scale structures**
- **Dark Matter & Dark Energy**
- **Violent Universe**

Many ongoing and futur scientific projects at CosmoStat