

Joint multi-band deconvolution for Euclid and Vera C. Rubin images

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OUTLINE

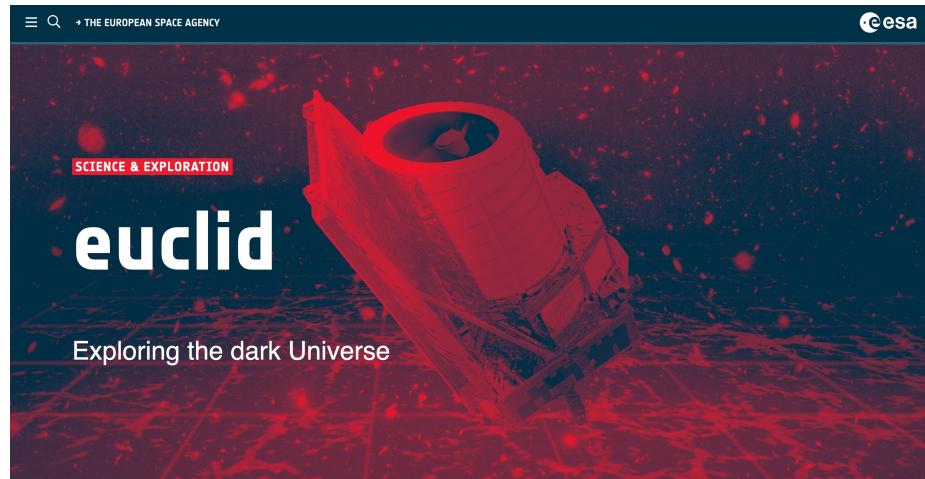
The deconvolution problem

Single band deconvolution

Joint multi-band deconvolution

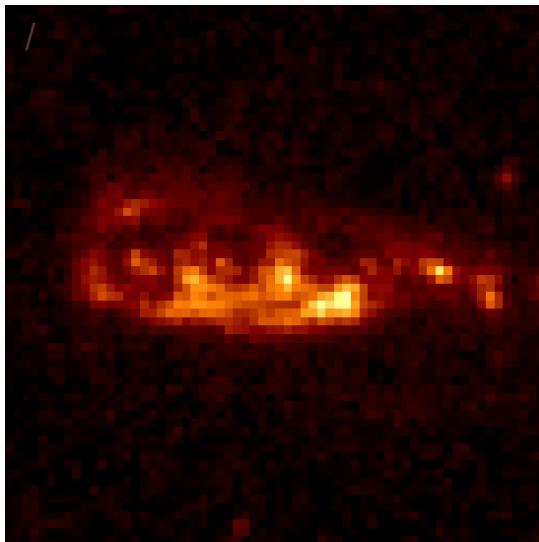


The screenshot shows the official website for the Vera C. Rubin Observatory. The header features the observatory's name and logo. Below the header is a large image of the telescope structure on a hillside under a starry sky. A central text overlay reads "See the Universe in action" and "Rubin Observatory will answer some of our biggest questions about the Universe!". At the bottom, there is a "Press Release" button and a summary text: "Rubin Observatory Will Help Unravel Mysteries of Dark Matter and Dark Energy".

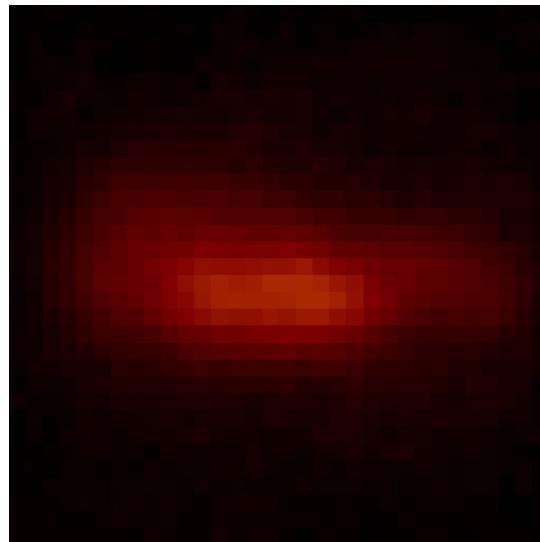


The screenshot shows the official website for the Euclid mission. The header includes the esa logo and the text "+ THE EUROPEAN SPACE AGENCY". Below the header is a large image of the Euclid satellite in space. A central text overlay reads "SCIENCE & EXPLORATION" and "euclid". At the bottom, the text "Exploring the dark Universe" is displayed.

The deconvolution problem



Ground Truth



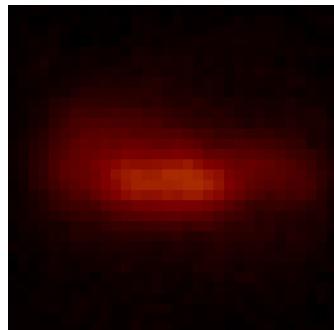
Observed Noisy Image

The deconvolution problem

Forward Model

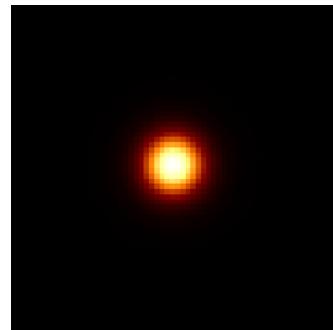
$$\mathbf{y} = \mathbf{h} * \mathbf{x}_t + \eta$$

(ill-posed inverse problem)



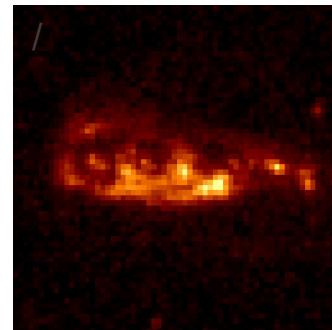
Observed Noisy Image

=



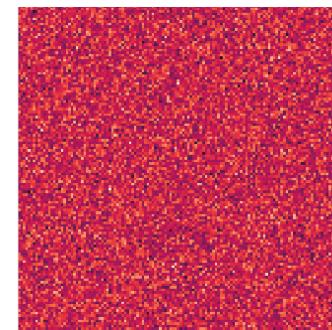
PSF

*



Ground Truth

+



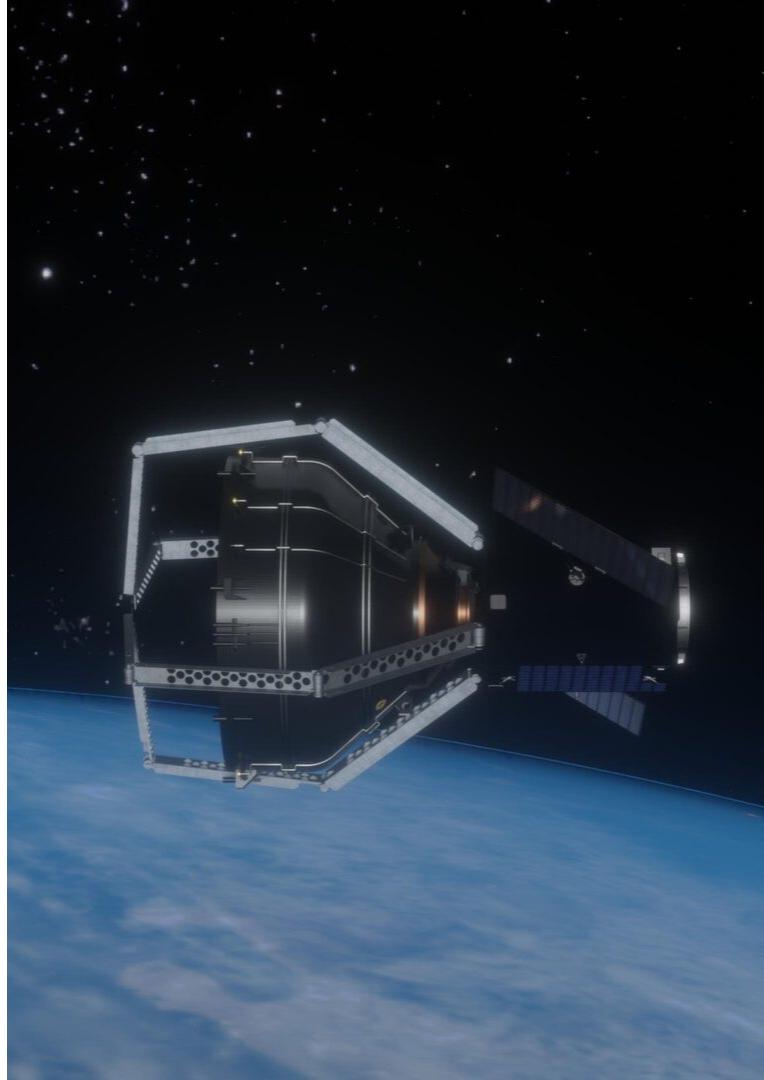
Additive Noise

$$\mathbf{y} \in \mathbb{R}^{n \times n}$$

$$\mathbf{h} \in \mathbb{R}^{n \times n}$$

$$\mathbf{x}_t \in \mathbb{R}^{n \times n}$$

$$\eta \in \mathbb{R}^{n \times n}$$



OUTLINE

The deconvolution problem

Single band deconvolution

Joint multi-band deconvolution

Solution

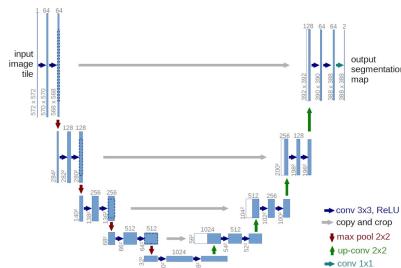
Loss function

$$L(\mathbf{x}) = \frac{1}{2\sigma^2} \| \mathbf{Hx} - \mathbf{y} \|_2^2 + \lambda \| \boldsymbol{\Gamma}\mathbf{x} \|_2^2$$

Analytical solution

$$\hat{\mathbf{x}} = (\mathbf{H}^\top \mathbf{H} + \lambda \boldsymbol{\Gamma}^\top \boldsymbol{\Gamma})^{-1} \mathbf{H}^\top \mathbf{y}$$

Deep learning based denoising



$$\sigma \in \mathbb{R}$$

$$\boldsymbol{\Gamma} \in \mathbb{R}^{n^2 \times n^2}$$

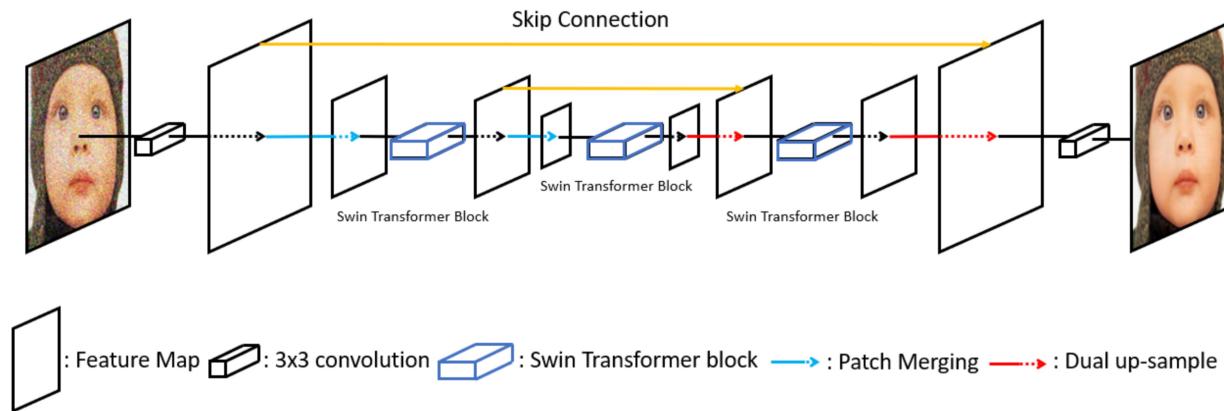
$$\lambda \in \mathbb{R}_+$$

$$\mathbf{H} \in \mathbb{R}^{n^2 \times n^2}$$

- Noise standard deviation
- Laplacian high-pass filter
- Regularization weight
- Block circulant matrix associated with the convolution operator \mathbf{h}

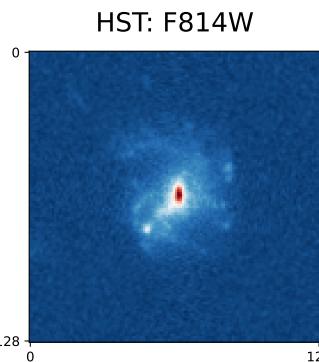
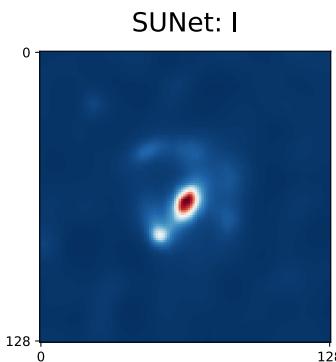
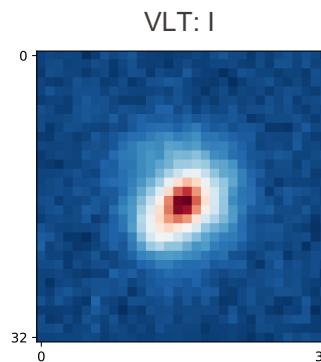
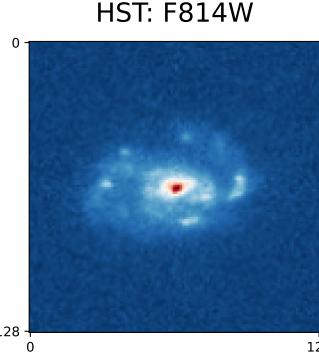
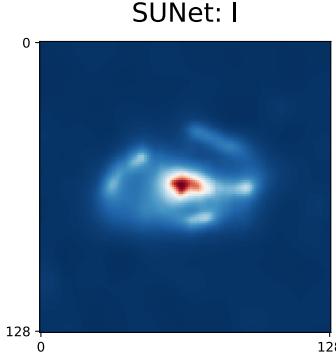
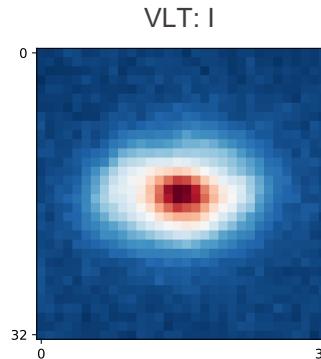
Deep learning-based denoising

SUNet



- A U-net with **Swin Transformer** blocks incorporated in the architecture

Test on real images from VLT, Chile



- **Noisy images:** VLT cutouts of 32×32 pixels in I-band (768nm) with resolution = **0.2''**

- **Ground truth:** HST cutouts of 128×128 pixels in the *F814W* filter with resolution = **0.05''**



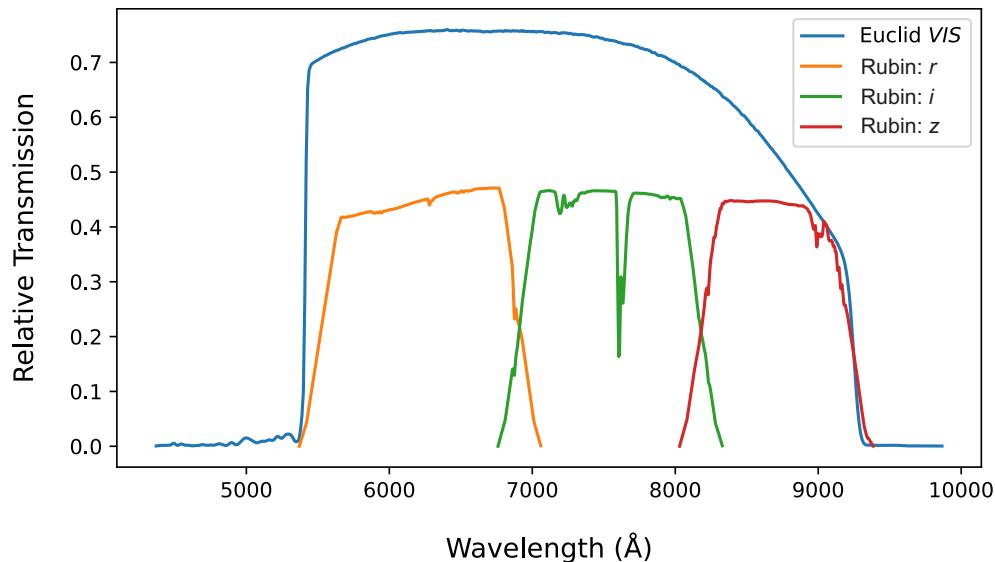
OUTLINE

The deconvolution problem

Single band deconvolution

**Joint multi-band
deconvolution**

The multi-band deconvolution problem



$$\mathbf{x}_{euc} = \alpha_r \mathbf{x}_r + \alpha_i \mathbf{x}_i + \alpha_z \mathbf{x}_z$$

Fractional flux contributions

$$\alpha_r, \alpha_i, \alpha_z \in \mathbb{R}^n$$

The multi-band deconvolution problem

Forward Model

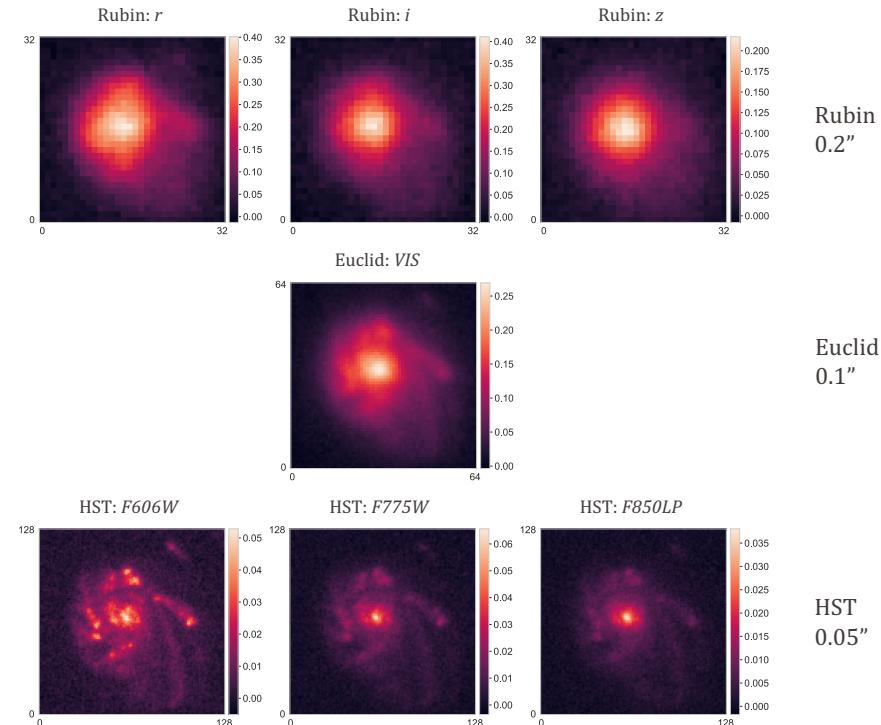
$$\mathbf{y}_r = \mathbf{h}_r * \mathbf{x}_r^t + \eta_r$$

$$\mathbf{y}_i = \mathbf{h}_i * \mathbf{x}_i^t + \eta_i$$

$$\mathbf{y}_z = \mathbf{h}_z * \mathbf{x}_z^t + \eta_z$$

$$\mathbf{x}_{euc}^t = \alpha_r \mathbf{x}_r^t + \alpha_i \mathbf{x}_i^t + \alpha_z \mathbf{x}_z^t$$

$$\mathbf{y}_{euc} = \mathbf{h}_{euc} * \mathbf{x}_{euc}^t + \eta_{euc}$$



- Observed Noisy Images
- PSFs
- Ground Truth Images
- Additive Noise
- Fractional Flux Contributions

$$\mathbf{y}_r, \mathbf{y}_i, \mathbf{y}_z \in \mathbb{R}^{n \times n}$$

$$\mathbf{h}_r, \mathbf{h}_i, \mathbf{h}_z \in \mathbb{R}^{n \times n}$$

$$\mathbf{x}_r^t, \mathbf{x}_i^t, \mathbf{x}_z^t \in \mathbb{R}^{n \times n}$$

$$\eta_r, \eta_i, \eta_z \in \mathbb{R}^{n \times n}$$

$$\alpha_r, \alpha_i, \alpha_z \in \mathbb{R}^n$$

The loss functions

$$L_r(\mathbf{x}_r) = \frac{1}{2} \left\| \frac{\mathbf{h}_r * \mathbf{x}_r - \mathbf{y}_r}{\sigma_r} \right\|_F^2 + \lambda_{constr} \left\| \frac{\mathbf{h}_{euc} * \sum_c \alpha_c \mathbf{x}_c - \mathbf{y}_{euc}}{\sigma_{euc}} \right\|_F^2$$

$$L_i(\mathbf{x}_i) = \frac{1}{2} \left\| \frac{\mathbf{h}_i * \mathbf{x}_i - \mathbf{y}_i}{\sigma_i} \right\|_F^2 + \lambda_{constr} \left\| \frac{\mathbf{h}_{euc} * \sum_c \alpha_c \mathbf{x}_c - \mathbf{y}_{euc}}{\sigma_{euc}} \right\|_F^2$$

$$L_z(\mathbf{x}_z) = \frac{1}{2} \left\| \frac{\mathbf{h}_z * \mathbf{x}_z - \mathbf{y}_z}{\sigma_z} \right\|_F^2 + \lambda_{constr} \left\| \frac{\mathbf{h}_{euc} * \sum_c \alpha_c \mathbf{x}_c - \mathbf{y}_{euc}}{\sigma_{euc}} \right\|_F^2$$

where

$$c \in \{r, i, z\}$$

$$\lambda_{constr} \in \mathbb{R}_+$$

- Fractional flux contributions
- Noisemaps

$$\alpha_r, \alpha_i, \alpha_z \in \mathbb{R}^n$$

$$\sigma_r, \sigma_i, \sigma_z \in \mathbb{R}^{n \times n}$$

Optimization

Loss Functions iteratively minimized using Gradient Descent

Step Sizes

$$\hat{\mathbf{x}}_{\{r,i,z\}} = \operatorname{argmin}_{\mathbf{x}_{\{r,i,z\}}} L_{\{r,i,z\}}(\mathbf{x}_{\{r,i,z\}})$$

$$\mathbf{x}_{\{r,i,z\}}^{[k+1]} = \mathbf{x}_{\{r,i,z\}}^{[k]} - \beta_{\{r,i,z\}} \nabla L_{\{r,i,z\}} \left(\mathbf{x}_{\{r,i,z\}}^{[k]} \right)$$

$$\beta_r, \beta_i, \beta_z \in \mathbb{R}^n$$

Gradients of the Loss Functions

$$\nabla L_r(\mathbf{x}_r) = \frac{\mathbf{h}_r^\top * (\mathbf{h}_r * \mathbf{x}_r - \mathbf{y}_r)}{\|\sigma_r\|_F^2} + 2\lambda_{constr} \alpha_r \mathbf{h}_{euc}^\top * \left[\frac{\mathbf{h}_{euc} * \sum_c \alpha_c \mathbf{x}_c - \mathbf{y}_{euc}}{\|\sigma_{euc}\|_F^2} \right]$$

$$\nabla L_i(\mathbf{x}_i) = \frac{\mathbf{h}_i^\top * (\mathbf{h}_i * \mathbf{x}_i - \mathbf{y}_i)}{\|\sigma_i\|_F^2} + 2\lambda_{constr} \alpha_i \mathbf{h}_{euc}^\top * \left[\frac{\mathbf{h}_{euc} * \sum_c \alpha_c \mathbf{x}_c - \mathbf{y}_{euc}}{\|\sigma_{euc}\|_F^2} \right]$$

$$\nabla L_z(\mathbf{x}_z) = \frac{\mathbf{h}_z^\top * (\mathbf{h}_z * \mathbf{x}_z - \mathbf{y}_z)}{\|\sigma_z\|_F^2} + 2\lambda_{constr} \alpha_z \mathbf{h}_{euc}^\top * \left[\frac{\mathbf{h}_{euc} * \sum_c \alpha_c \mathbf{x}_c - \mathbf{y}_{euc}}{\|\sigma_{euc}\|_F^2} \right]$$

Convergence guarantee & Optimal step size

A function's gradient is Lipschitz continuous if

$$\|\nabla f(\mathbf{x}') - \nabla f(\mathbf{x})\| \leq C\|\mathbf{x}' - \mathbf{x}\|$$

where C is the Lipschitz constant

In our case

$$\|\nabla L_{\{r,i,z\}}(\mathbf{x}'_{\{r,i,z\}}) - \nabla L_{\{r,i,z\}}(\mathbf{x}_{\{r,i,z\}})\| \leq C_{\{r,i,z\}}\|\mathbf{x}'_{\{r,i,z\}} - \mathbf{x}_{\{r,i,z\}}\|$$

Substituting the individual loss functions, we get

$$C_{\{r,i,z\}} \geq \frac{\mathbf{h}_{\{r,i,z\}}^\top * \mathbf{h}_{\{r,i,z\}}}{\|\sigma_{\{r,i,z\}}\|_F^2} + \frac{2\lambda_{constr}\alpha_{\{r,i,z\}}^2 \mathbf{h}_{euc}^\top * \mathbf{h}_{euc}}{\|\sigma_{euc}\|_F^2}$$

The Optimal Condition for Convergence

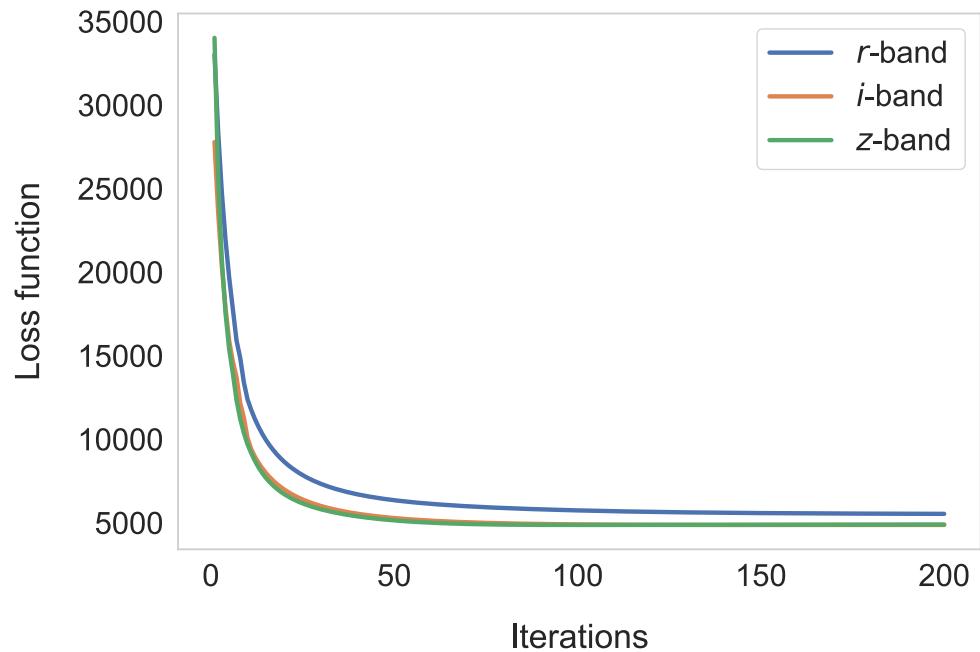
$$\beta_{\{r,i,z\}} \leq \frac{1}{C_{\{r,i,z\}}}$$

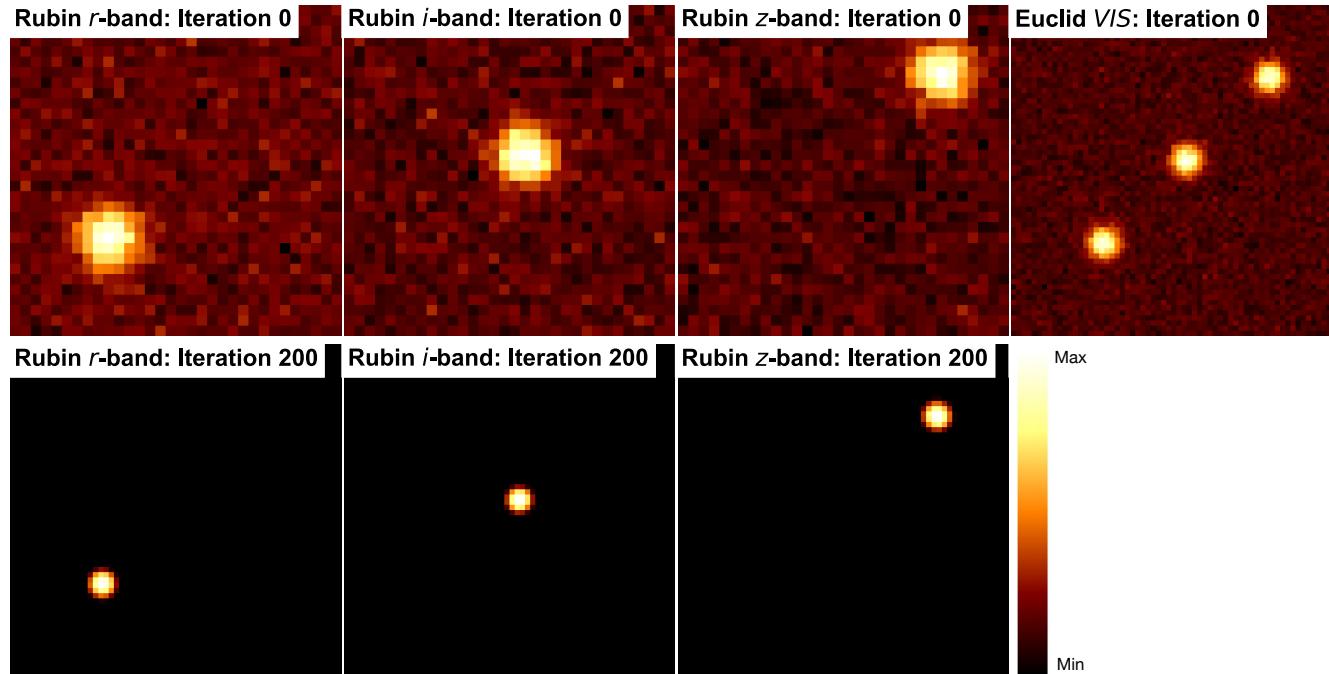
Hence, we choose

$$\beta_{\{r,i,z\}} = \frac{1}{(1 + 10^{-5})C_{\{r,i,z\}}}$$

Convergence

- Algorithm run for 200 iterations
- Convergegence within 50-100 iterations

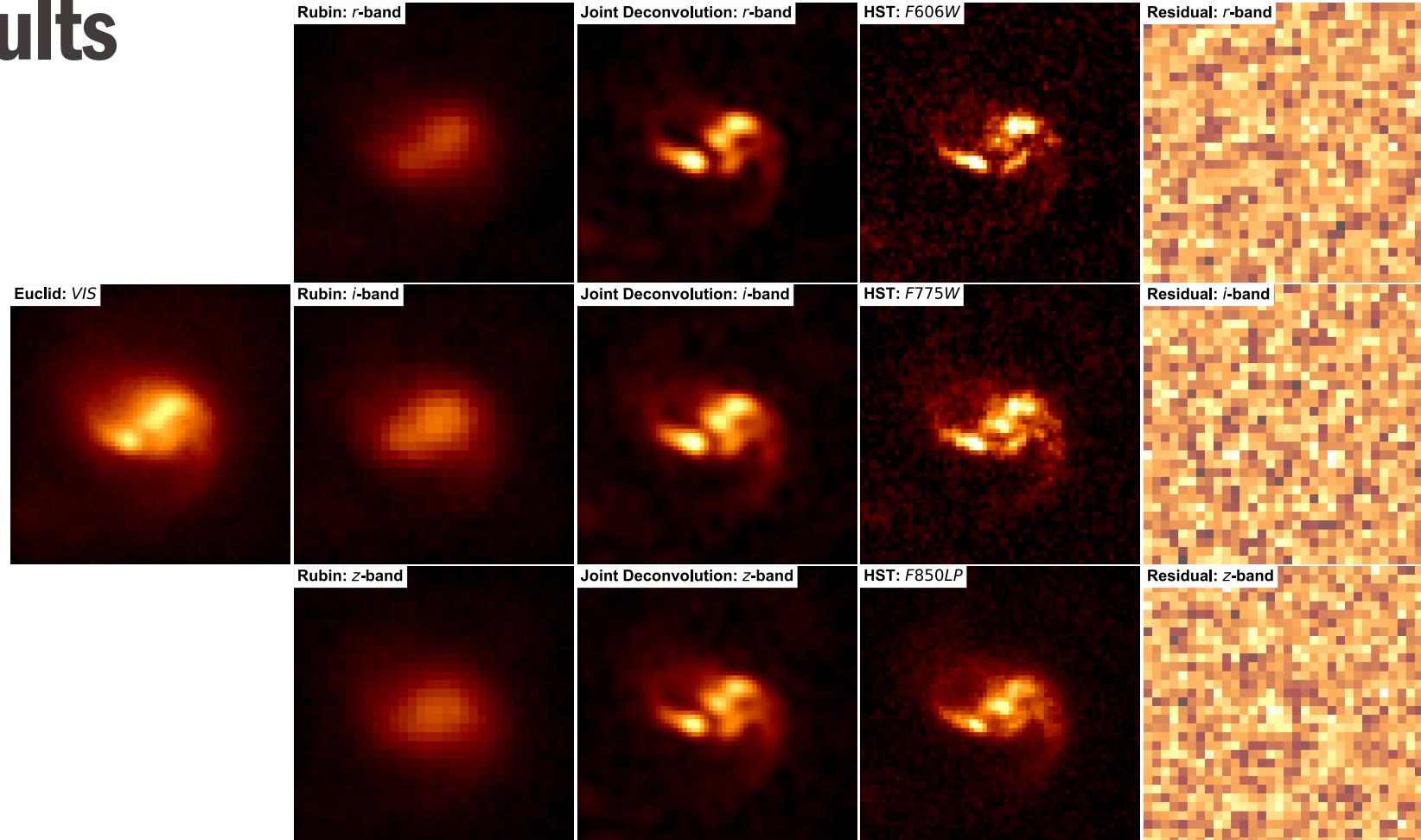




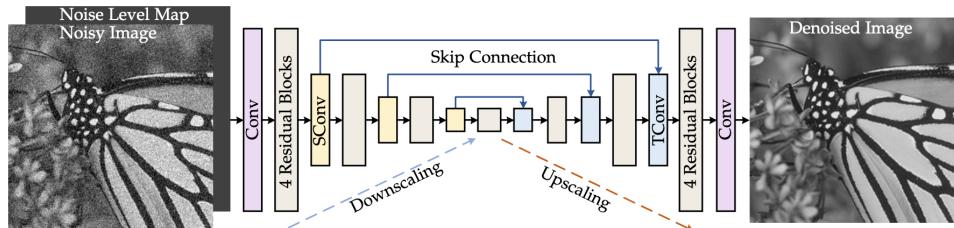
Flux Leakage Test

- Assume 3 separately placed Gaussians in each channel (corresponding to LSST channels)
- The joint image (Euclid) is a linear sum of these channels
- No Flux Leakage from one channel to another

Results

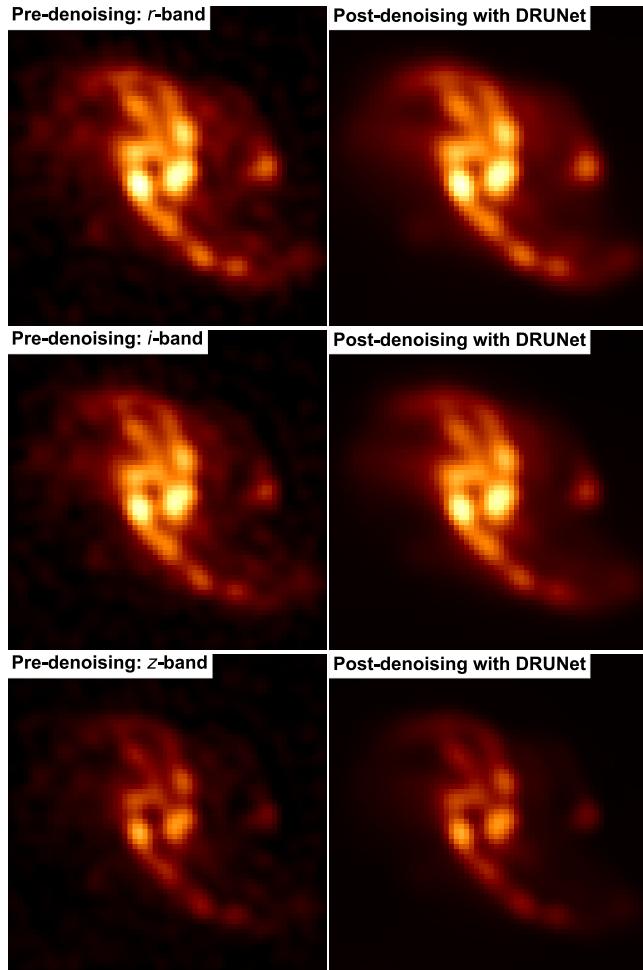


DRUNet denoising

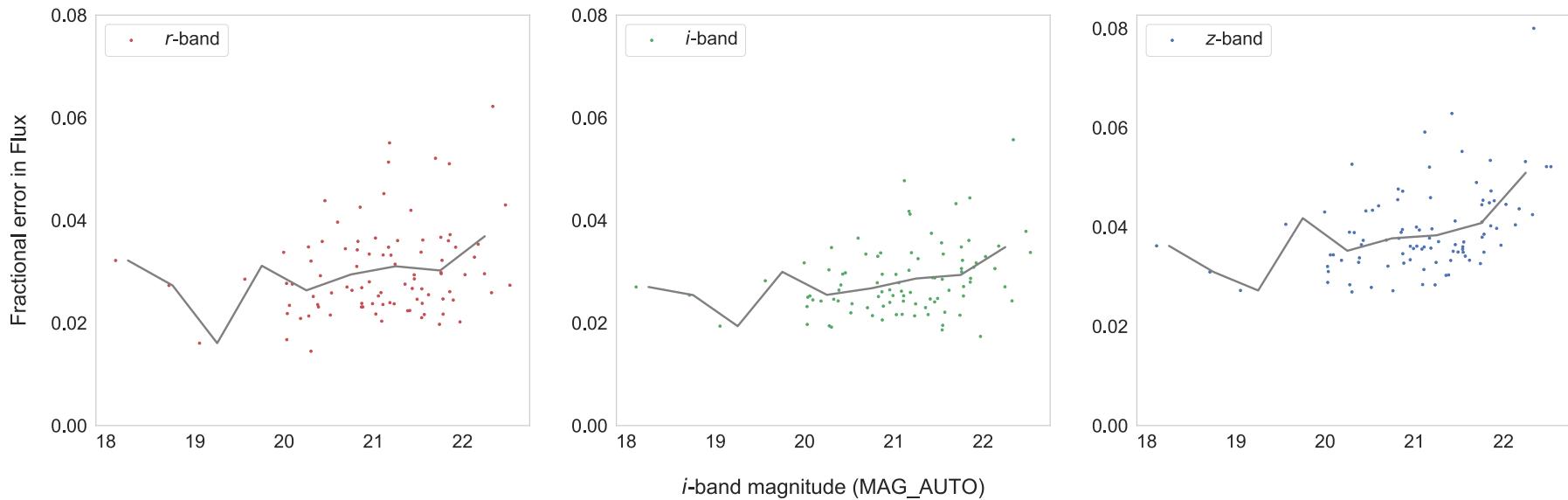


Plug-and-Play Image Restoration with
Deep Denoiser Prior, Zhang et al., 2021

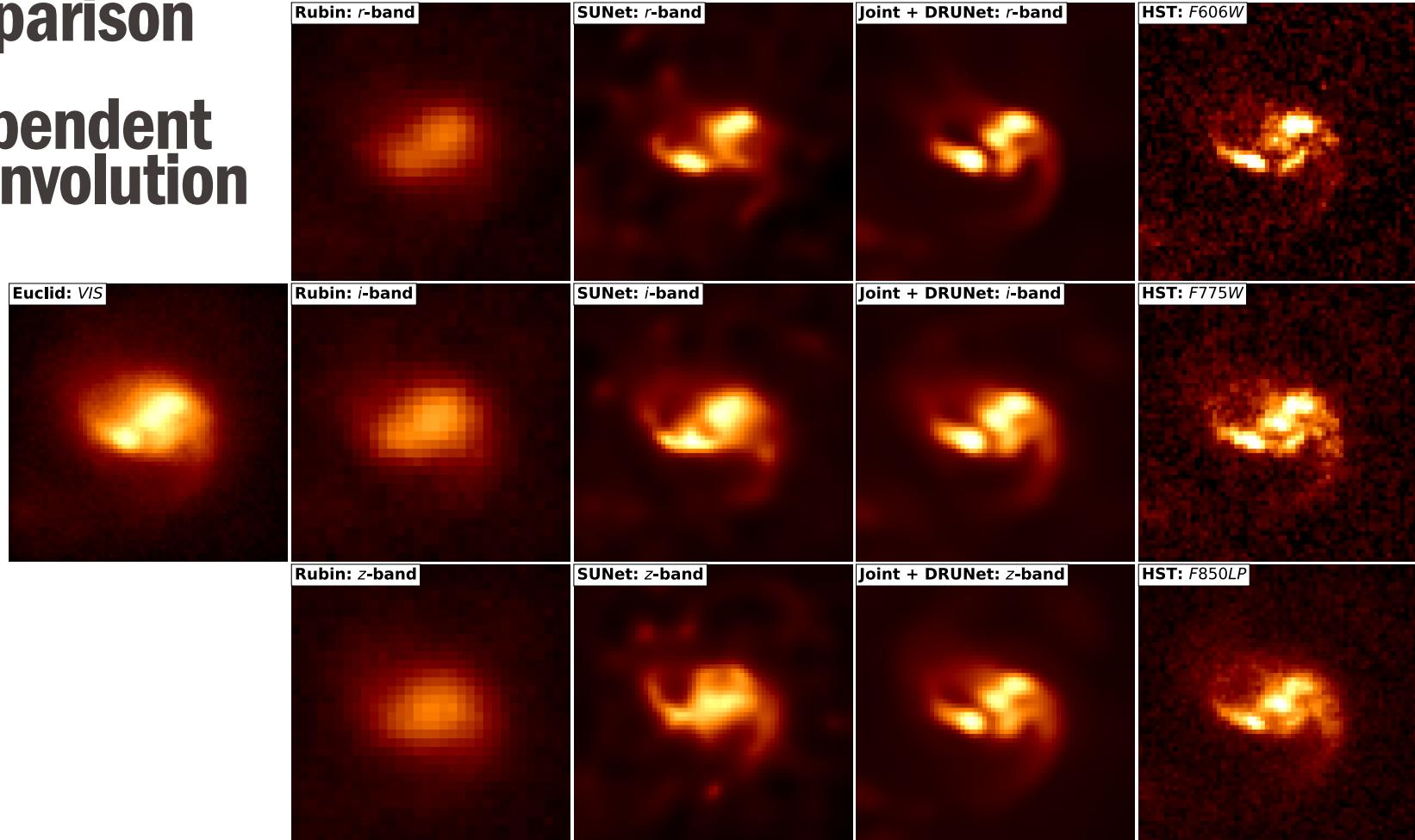
NMSE	<i>r</i> -band	<i>i</i> -band	<i>z</i> -band
Pre-denoising	0.059	0.041	0.053
Post-denoising	0.058	0.038	0.038
% improvement	1.69%	7.32%	28.3%



Flux recovery

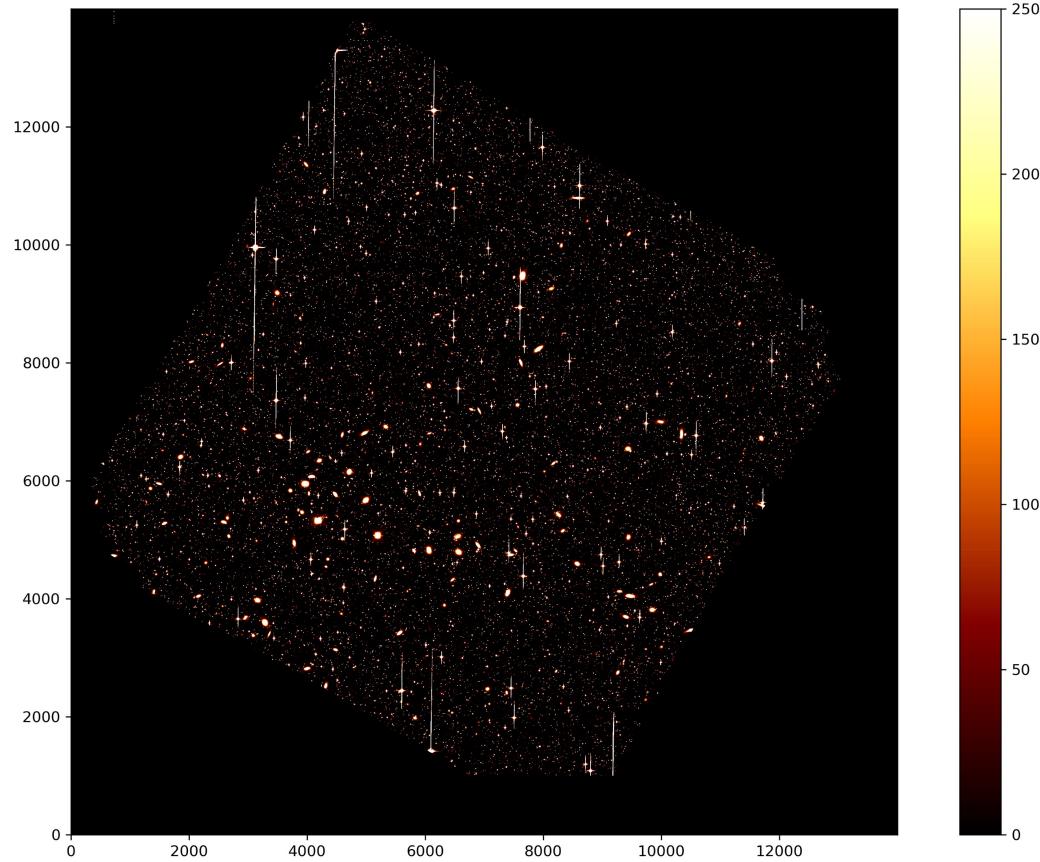


Comparison with independent deconvolution



Test on real data

Perseus Cluster



Test on real data

Euclid Images

- Euclid Early Release Observation (ERO)
- Perseus Cluster [ERO-10] - <https://euclid.esac.esa.int/dr/ero/ERO-Perseus>
- VIS Band
- Pixel scale = 0.1"



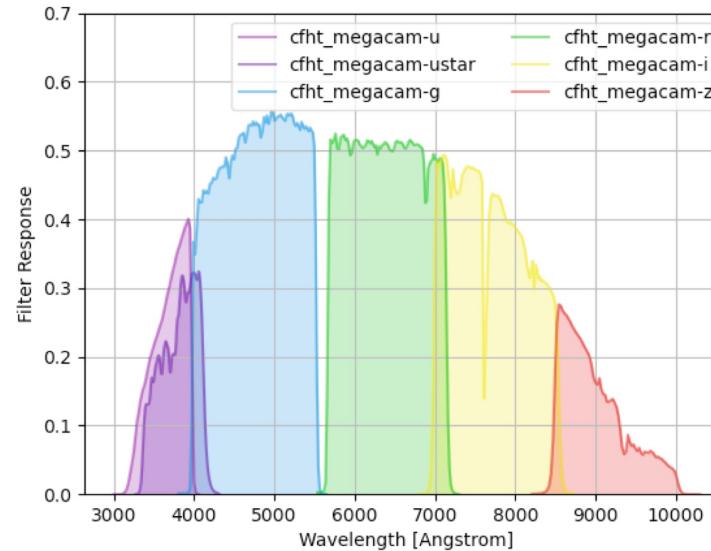
Name	ERO-Perseus	
DOI	https://doi.org/10.57780/esa-qmocze3	
Data	Image	Catalog
	VIS	Euclid-VIS-Stack-ERO-Perseus.DR3.tar Euclid-VIS-Catalog-ERO-Perseus.DR3.tar.gz
	NISP	Euclid-NISP-Stack-ERO-Perseus.DR3.tar Euclid-NISP-Catalog-ERO-Perseus.DR3.tar.gz
Version	V3.0	
Credit guidelines	Please refer to the credits page for instructions on how to acknowledge the use of this data.	
Acknowledgement	Euclid is a fully European mission, built and operated by ESA, with contributions from NASA. The Euclid Consortium is responsible for providing the scientific instruments and scientific data analysis. ESA selected Thales Alenia Space as prime contractor for the construction of the satellite and its Service Module, with Airbus Defence and Space chosen to develop the Payload Module, including the telescope. NASA provided the near-infrared detectors of the NISP instrument. Euclid is a medium-class mission in ESA's Cosmic Vision Programme.	

The Euclid Data Service is managed by the [Euclid Science Operations Centre](#)

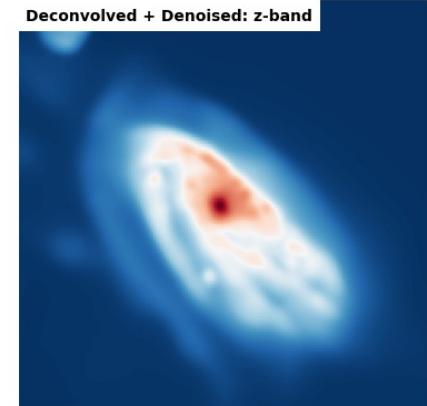
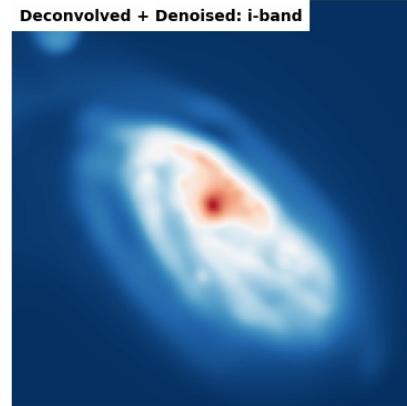
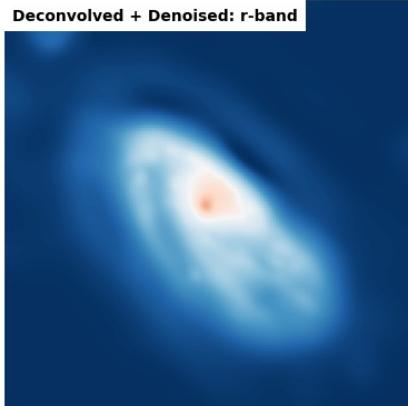
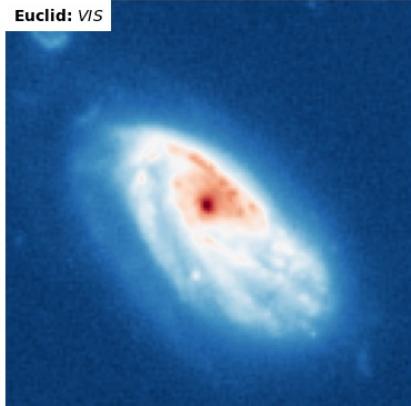
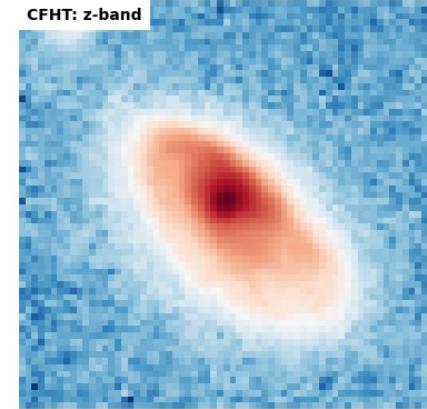
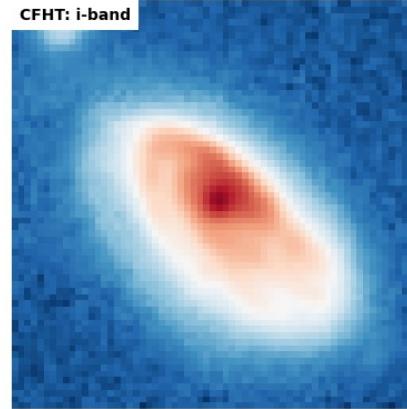
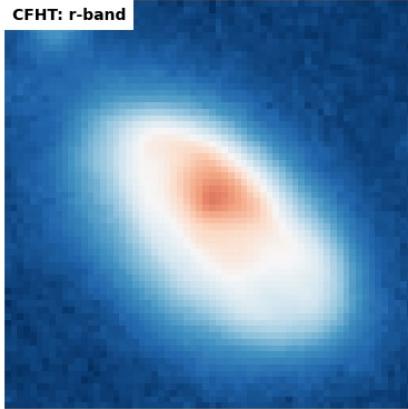
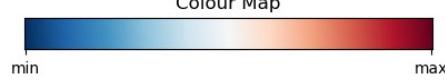
Test on real data

CFHT Images

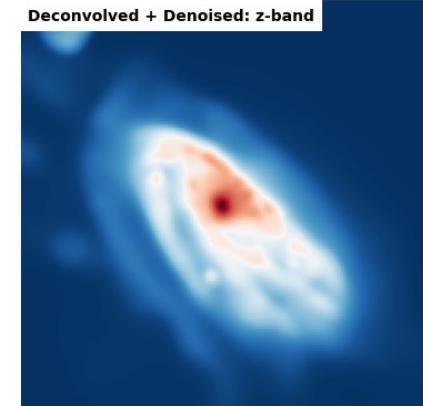
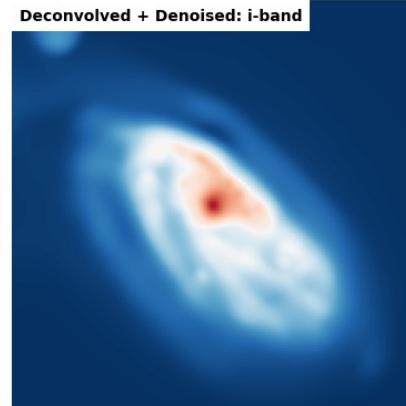
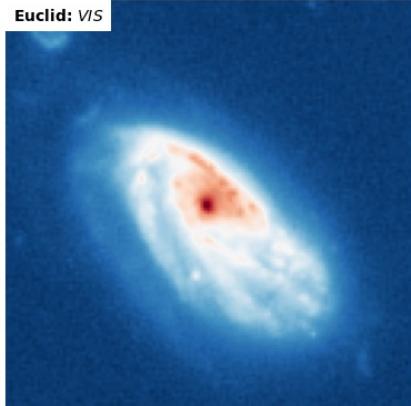
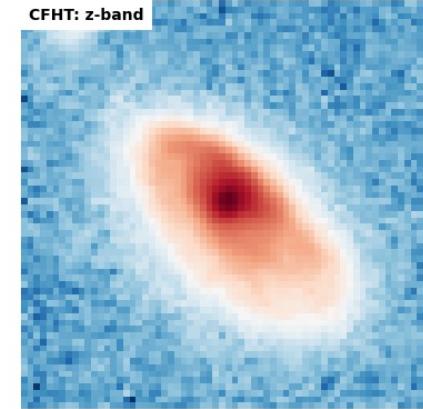
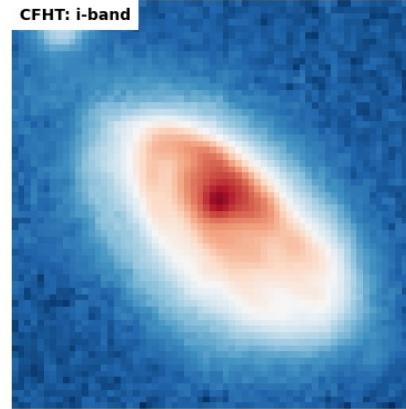
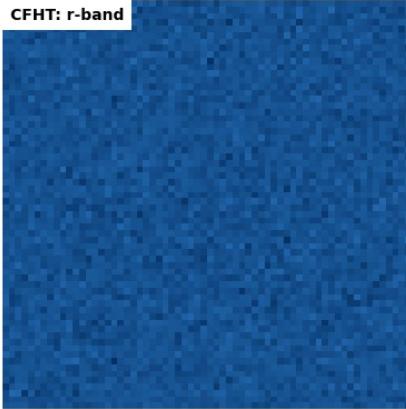
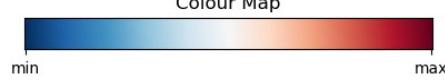
- MegaCam - wide-field optical imaging facility at CFHT that covers a 1 x 1 square degree FOV
- Region of interest - Perseus Cluster (covereing the same portion of the sky as the Euclid ERO FOV)
- Pixel scale = $0.187''$ (median seeing = $0.7''$ at Mauna Kea)
- Images rebinned to $0.2''$ (to obtain an intergral ratio with respect to the Euclid pixel scale of $0.1''$)
- r, i, z bands



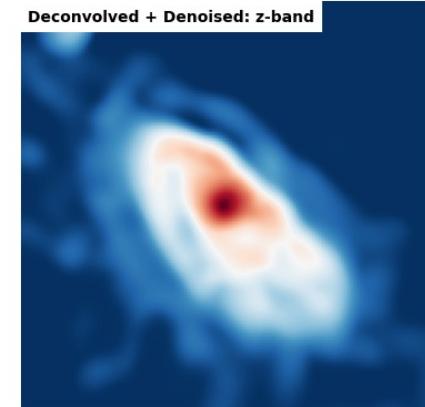
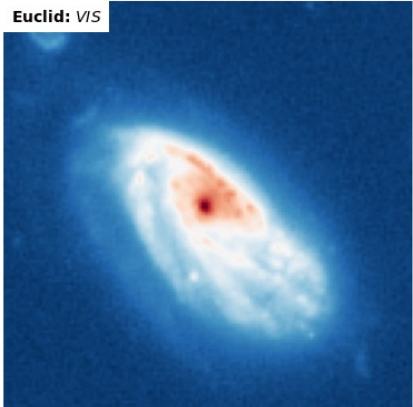
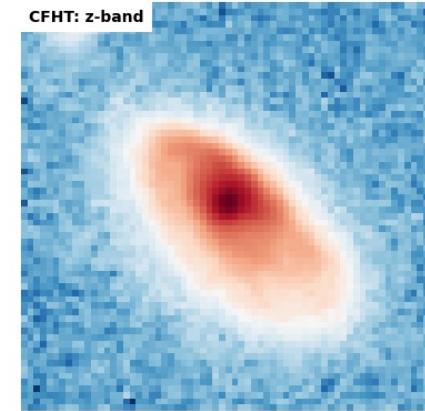
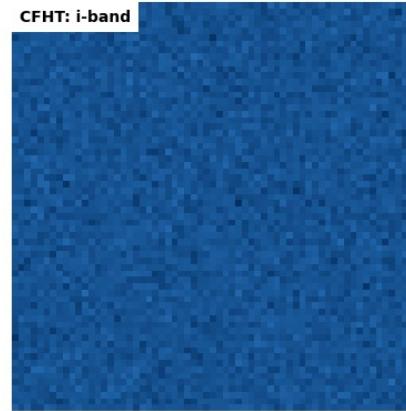
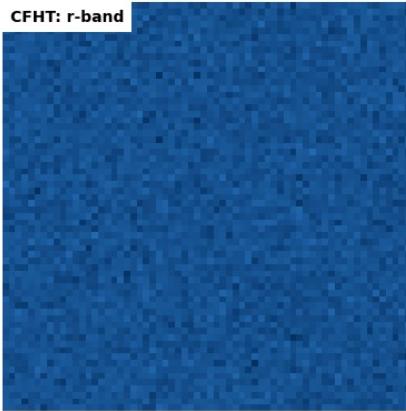
Test on real data



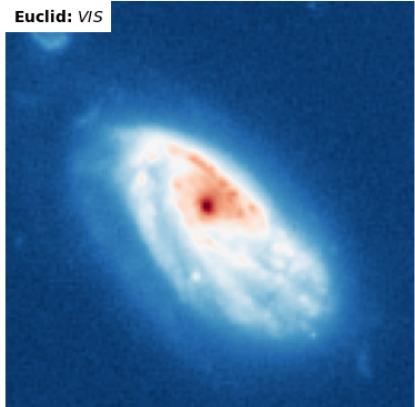
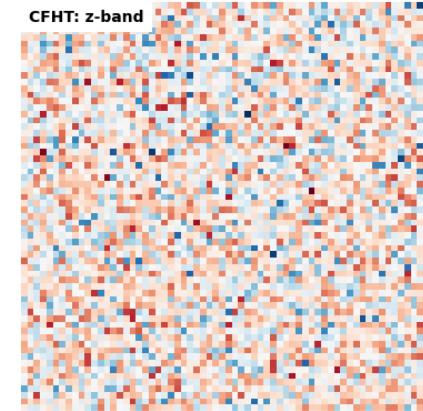
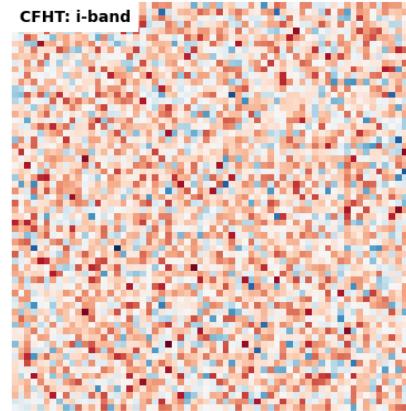
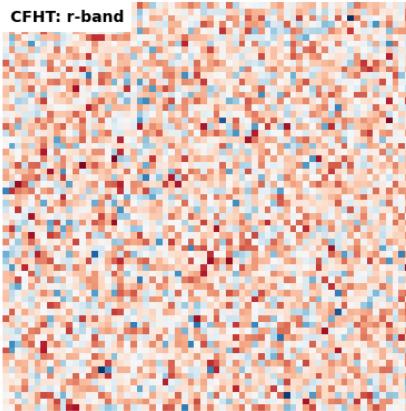
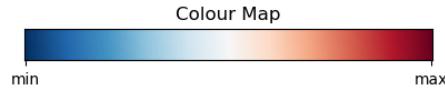
Leakage test



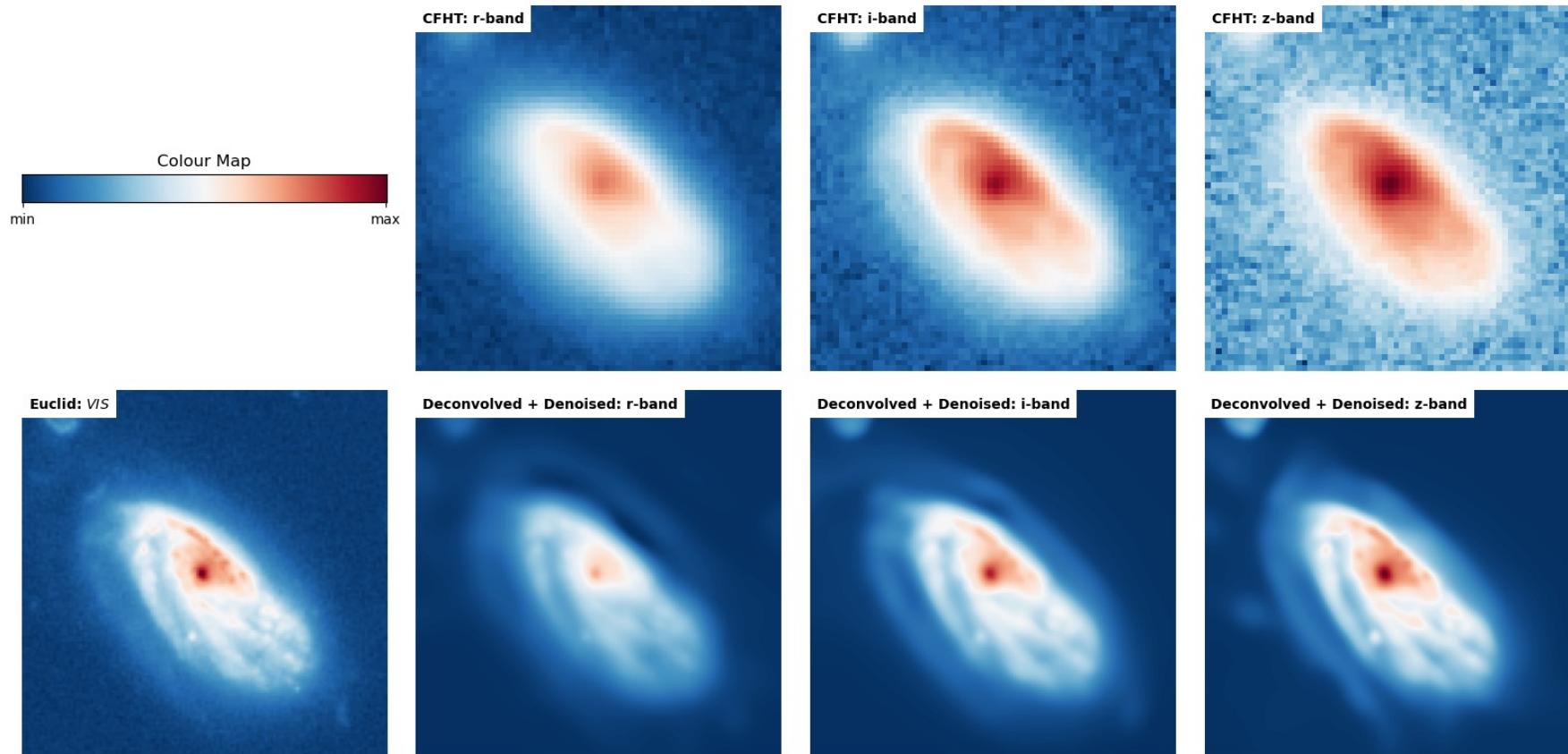
Leakage test



Leakage test



Conclusion





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