



AI, Astrostat and Big Data Challenges for the ALMA2040 Era

Fabrizia Guglielmetti (ESO)



From ALMA WSU to ALMA 2040: Telescope capability increases



Data Volume



Scientific and Computing complexity



More users



Current reduction paradigm : PL and/or manual QA + CLEAN + inspection



Bayesian inference + AI : BRAIN framework

	ALMA TODAY	ALMA2030 (WSU)	ALMA2040
Bandwidth	■	■	■
Data volume	■	■	■
Cube size	■	■	■
Spectral richness	■	■	■
Calibration	■	■	■
Imaging cost	■	■	■
Human effort	■	■	■
AI contribution	■	■	■



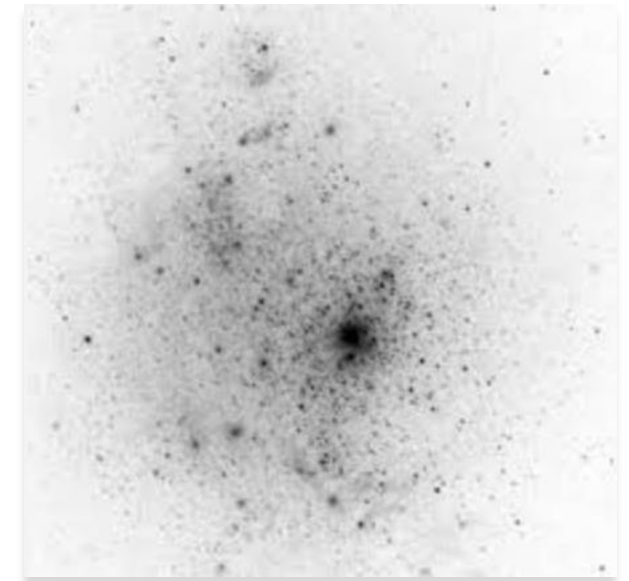
Is Big Data a 21st-century problem?

Big Data challenges triggered modern data analysis

A brief astronomical computing history

STScI foundation planning work :

- Newsletters and technical memos from late 1970s
 - Early introduction of **AI, Machine Learning, and Bayesian methods** centers around a massive problem: **How to guide the Hubble Space Telescope.**
 - No digital map of the sky dense enough to guide HST
 - Digitizing thousands of photographic plates (20×10^6 sources)
 - **Massive catalogue (10^9 objects) was produced**
- (Barry M. Lasker, **FG**, *et al*, 2008, *AJ*, **136** 735)



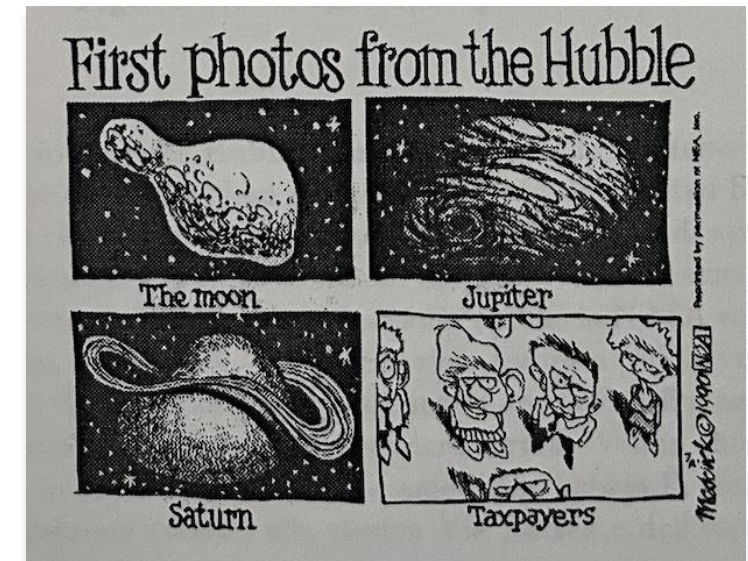
Schmidt plate, Coma cluster

STScI, AI and Bayesian Connections

A brief astronomical computing history

Dealing with a data problem decades ahead of its time, its early newsletters and technical memos became a hotbed for discussing advanced computational techniques:

- **Automated Classification (The Roots of ML)**
 - **Bayesian approach to pattern recognition and classification complexities**
 - **First ML applications** “*Automated Star-Galaxy Discrimination with Neural Networks*”
Odewahn, S. C., et al. 1992, **AJ**, 103, 318
- **Artificial Intelligence for Scheduling (SPIKE)**
 - Based on NN (1980s)
- **The Bayesian Image Restoration Shock (1990)**
 - Lucy-Richardson, MEM



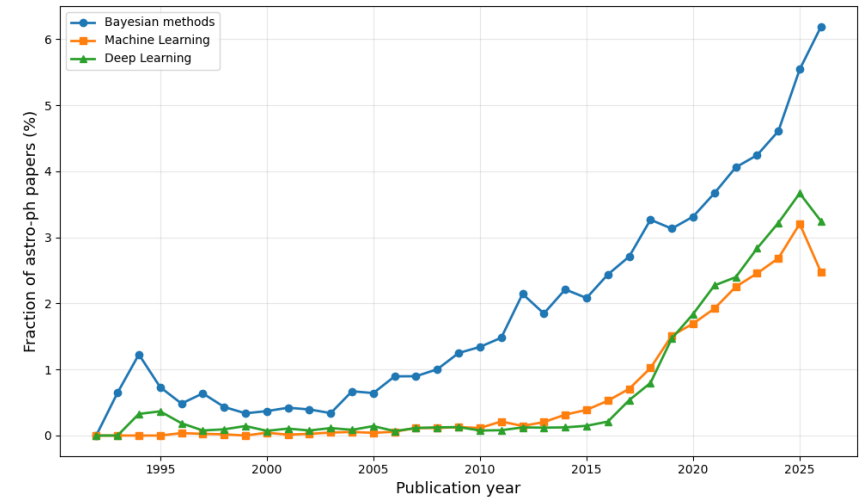
Bayesian Inference in astronomy

Gold standard for interpreting the cosmos

- Earliest roots (1960s-1970s)
 - The 1968 Foundation (Ed Jaynes)
 - The Lucy-Richardson Deconvolution (1974)
- Modern Revolution: MCMC (1990s)

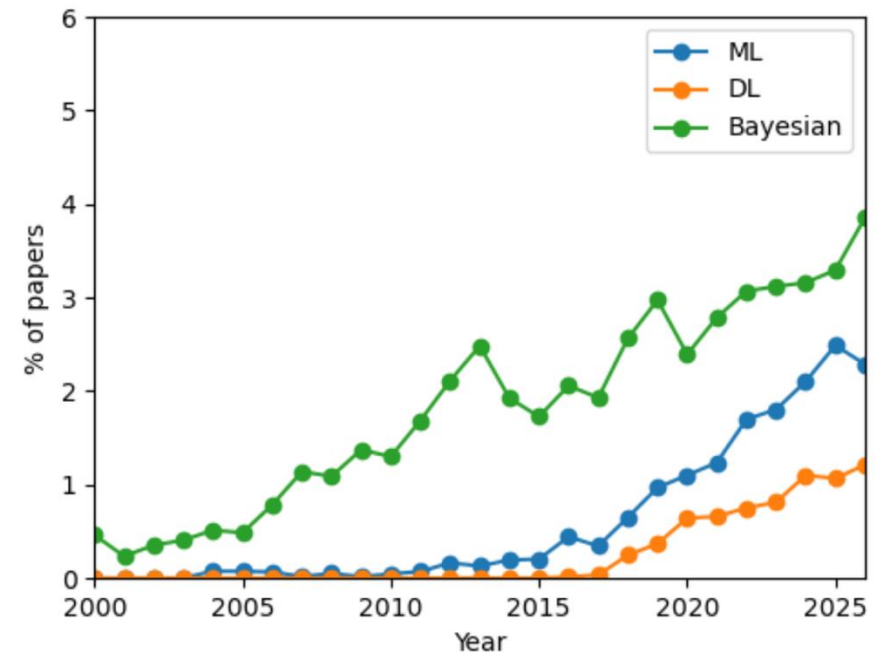
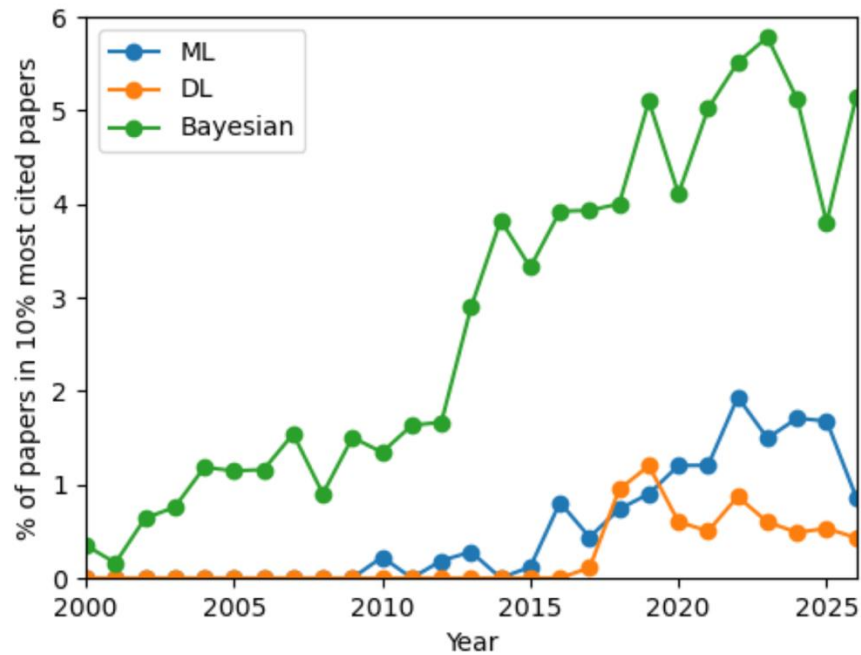
$$P(\theta|D) = \frac{P(D|\theta)P(\theta)}{P(D)}$$

- *"From Laplace to supernova SN 1987A: Bayesian inference in astrophysics"*, Tom Loredo (1990)
- **Today : treated as a core branch of probabilistic machine learning**
 - Exoplanet Detection, Cosmological Parameters, Gravitational Wave Analysis



Data taken from arXiv: astro-ph – statistics based on titles and abstracts

Most influential methods



(papers in A&A, ApJ, AJ & MNRAS – statistics based on titles & abstracts only!)



Bayesian techniques allow astronomers to bake the laws of physics directly into the model as "Priors" (pdf), meaning they can extract highly accurate, mathematically sound discoveries from even the faintest, most isolated signals in the sky

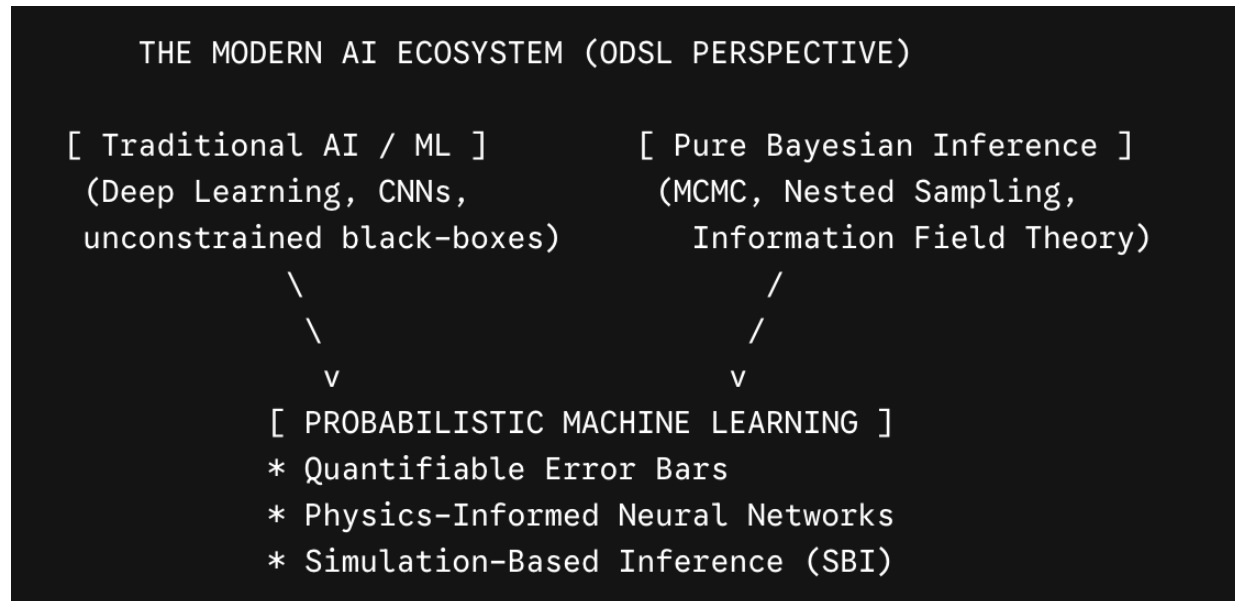


What is modern AI?

The emergence of modern AI

Major paradigm shift at the Garching Hub: Intersection of Deep Learning, Bayesian Inference, and Physical Laws

- From AI treated as a “black-box” (simply looking at patterns) to **Physics-Informed and Risk-Aware AI**
- Systems that understand **uncertainty** and the **laws of physics**





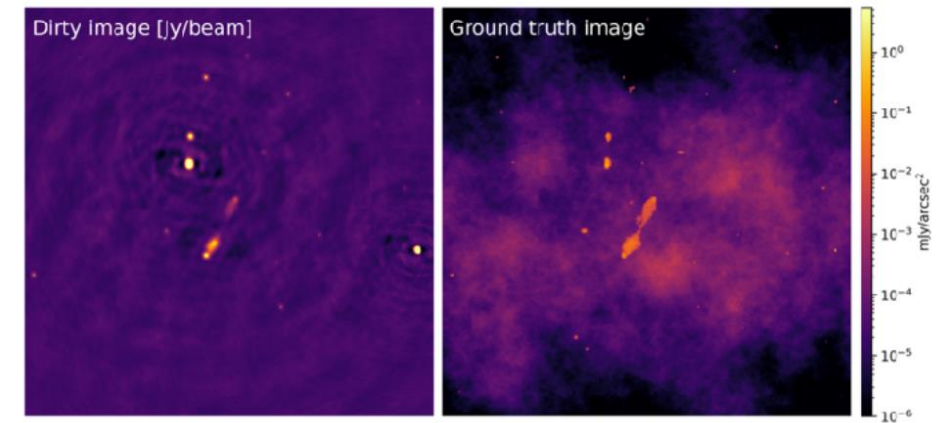
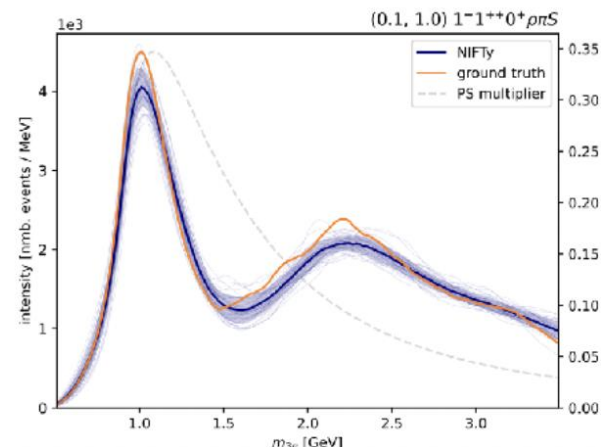
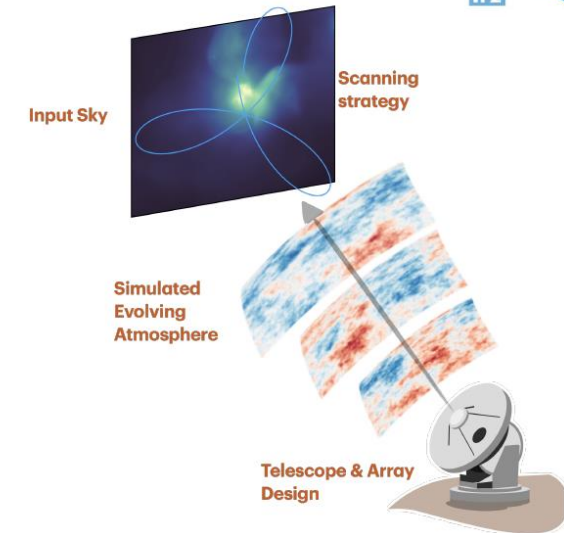
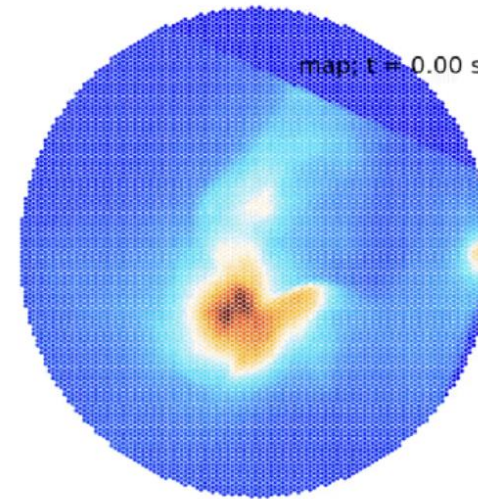
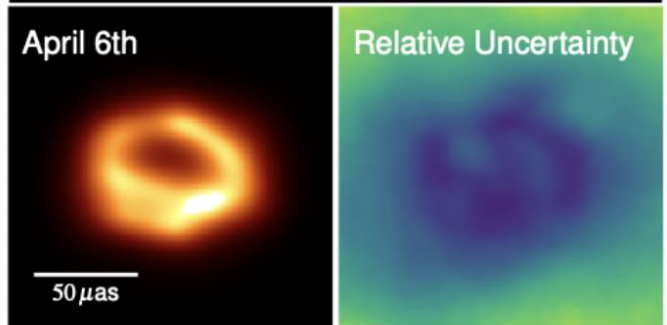
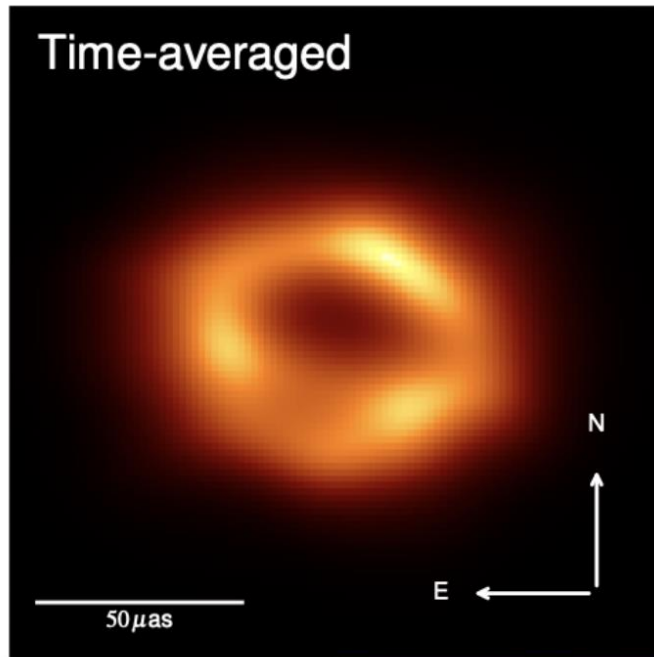
The emergence of modern AI

Major paradigm shift at the Garching Hub: Intersection of Deep Learning, Bayesian Inference, and Physical Laws

- **Origins Data Science Lab (ODSL)**
 - A group of interdisciplinary minded people interested in Statistics, AI/ML, Data Science, Open Data / Open Science, Distributed Computing
 - Bridging the two historically competing philosophies of data science:
 - **The Frequentist/Deep Learning camp:** "Give me massive data, and the neural network will figure out the rules."
 - **The Bayesian camp:** "Give me a physical model, and I will compute the exact probabilities."
 - Demonstrating to the broader AI community how to build **trustworthy AI** for the natural sciences.
 - Mapping out the future blueprint of how AI operates under strict physical constraints.

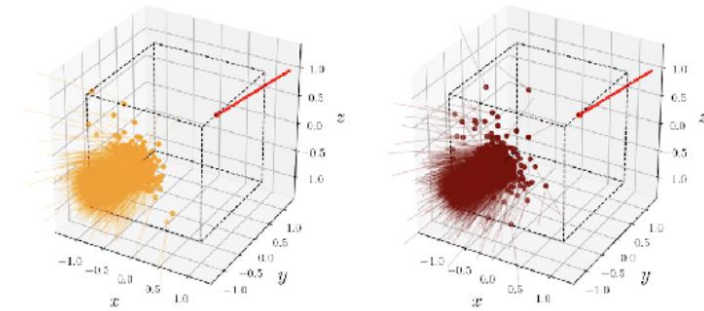
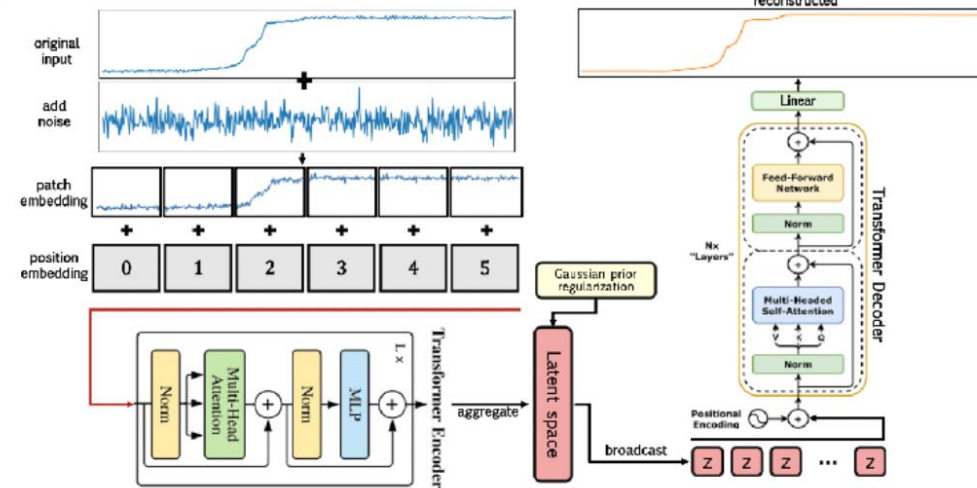
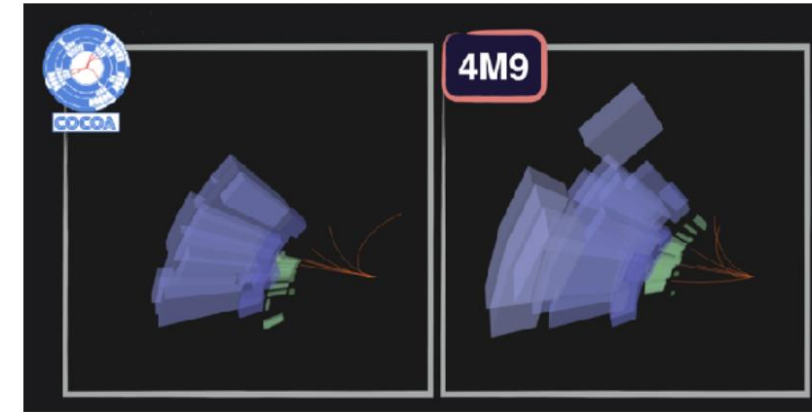
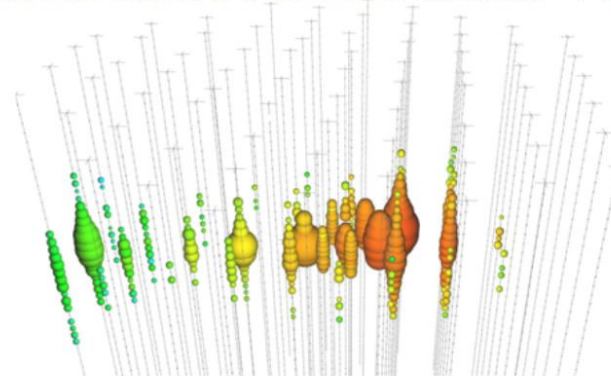
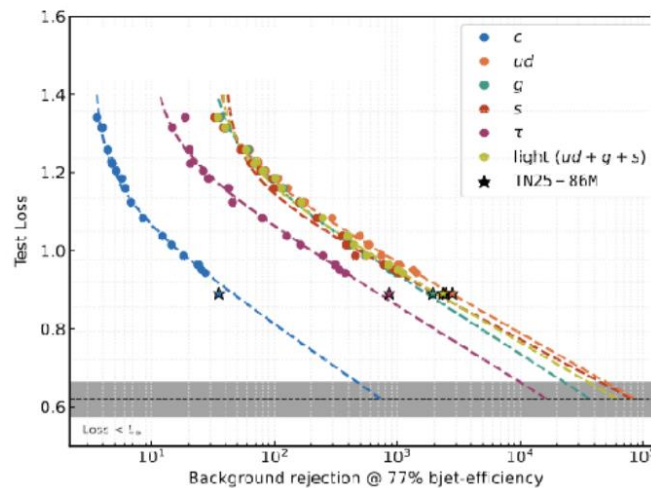
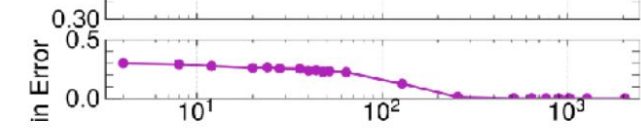
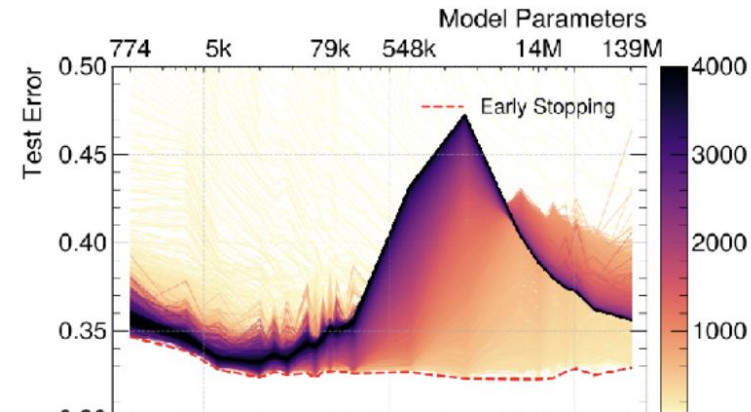
Bayesian Inverse Problems

Transfer over “multiple generations” from Astro to Particle Physics



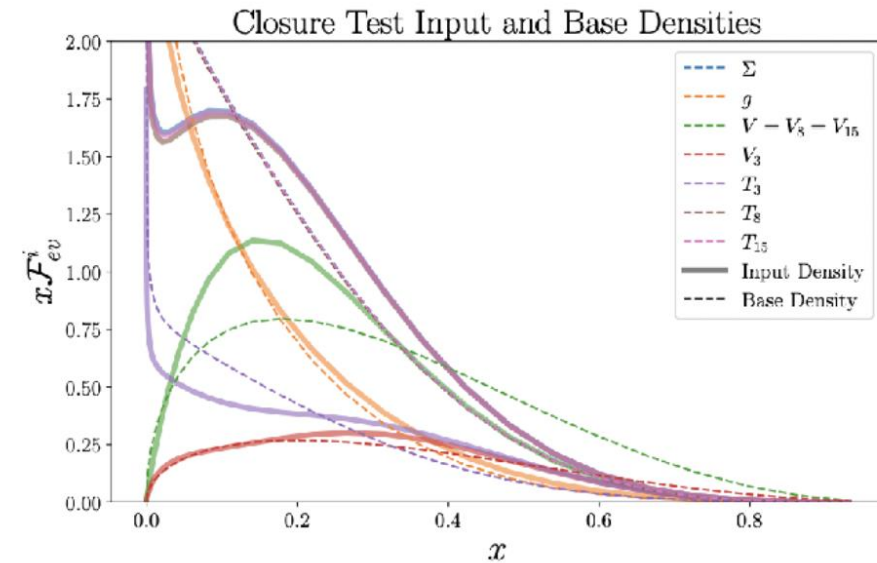
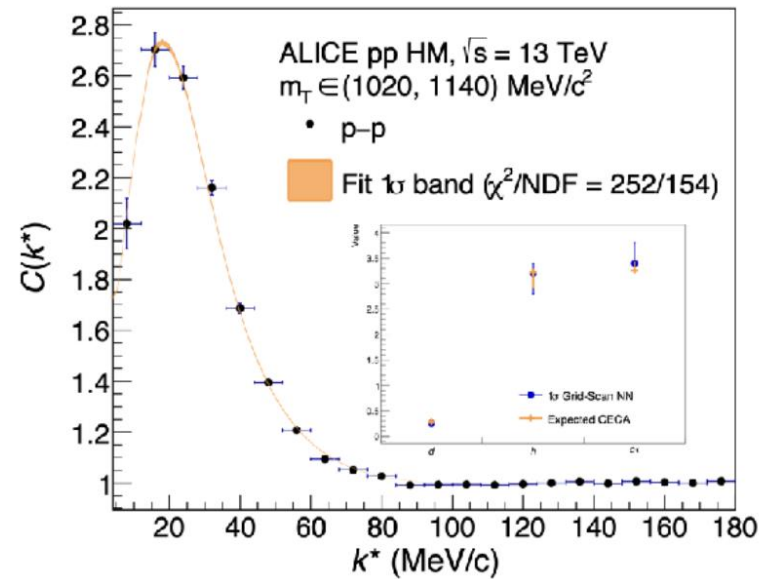
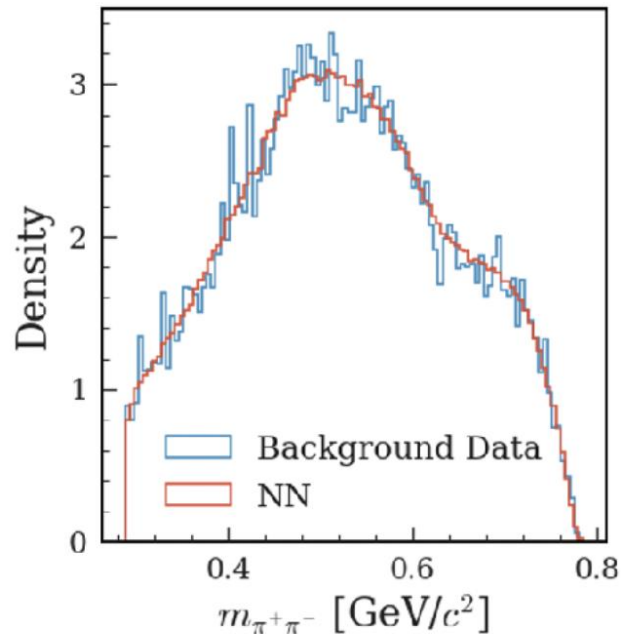
Foundation Models:

Big Trend: from specialists to generalist AI models. We can train O(1-10 B) models!



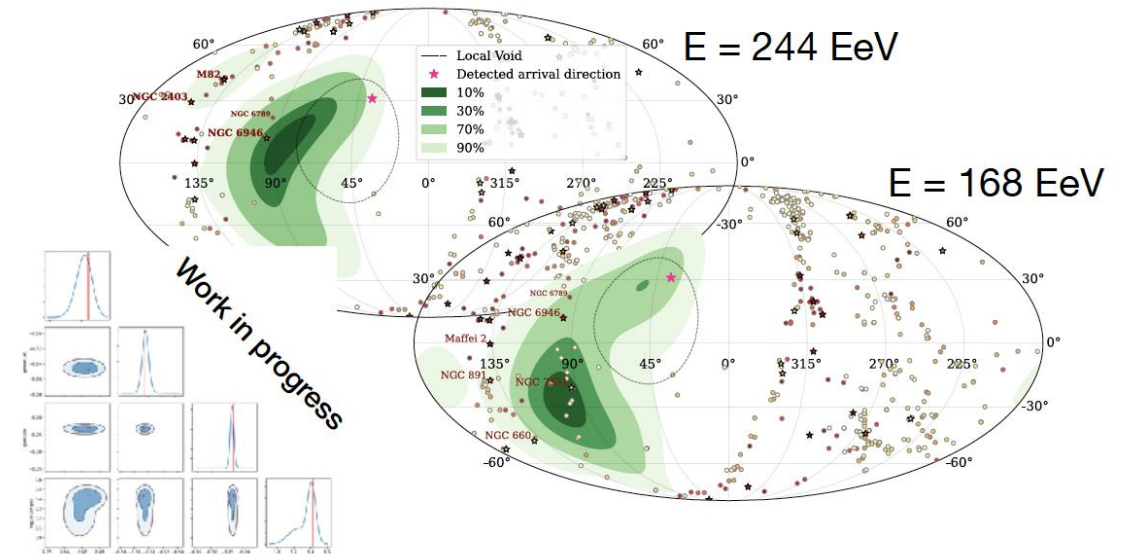
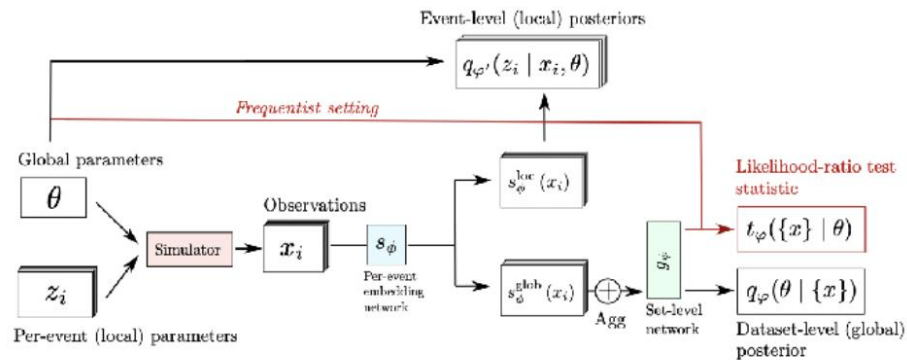
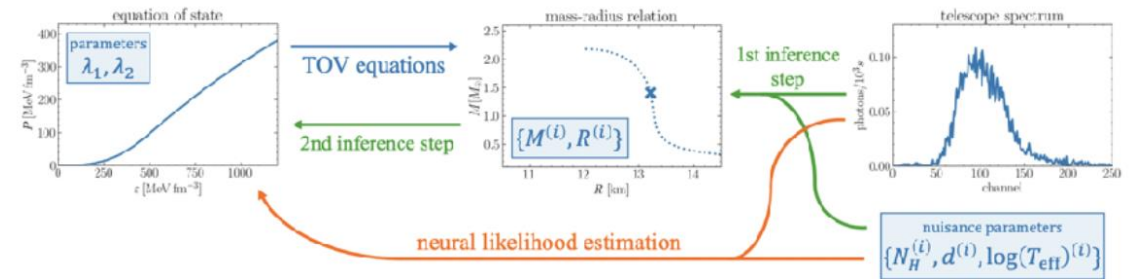
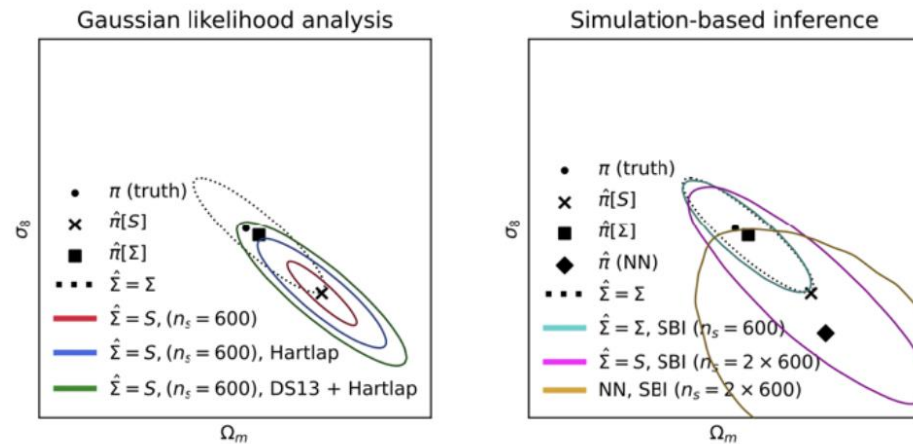
Neural Density Modelling

AI methods are excellent as a non-parametric density modelling tool. Multiple applications across ORIGINS sciences e.g. for background modelling



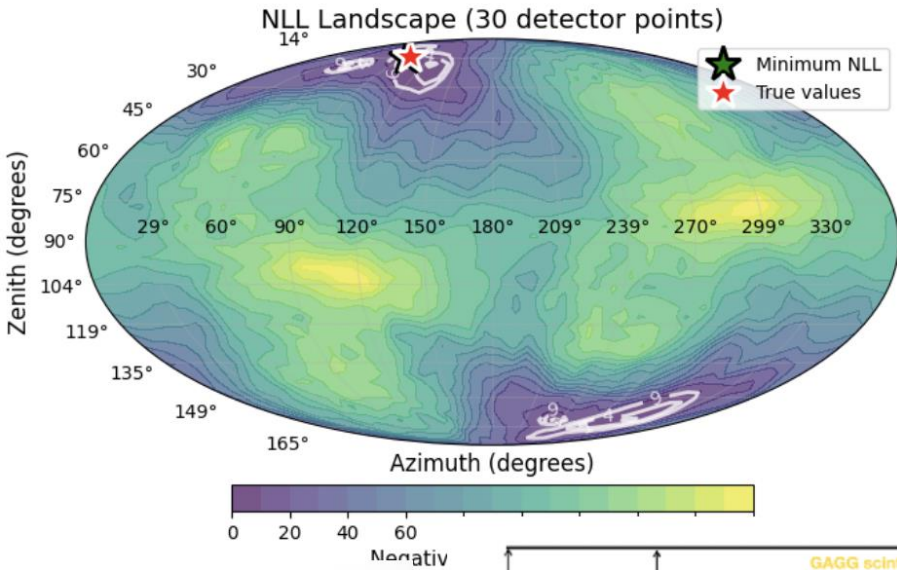
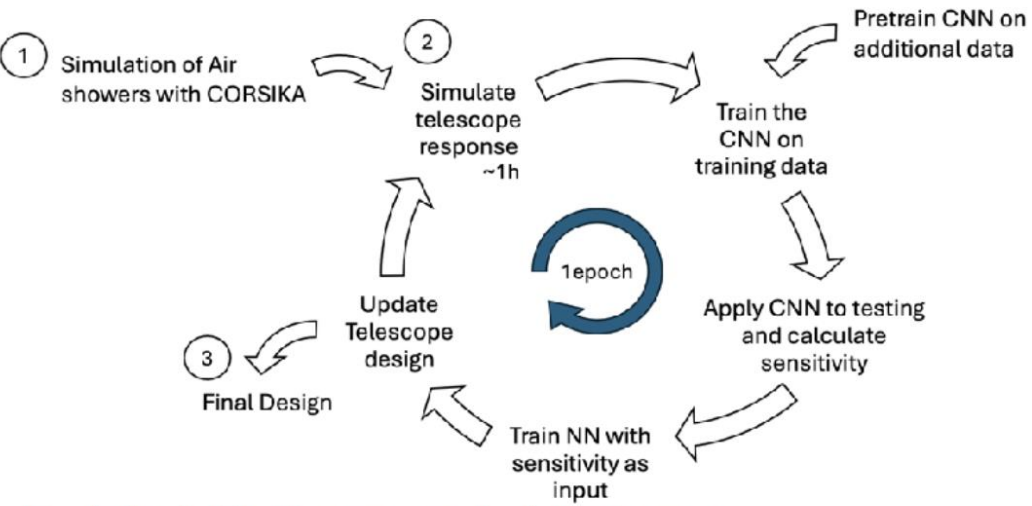
Simulation-based Inference

A updated, cross-disciplinary approach on statistical inference for big science

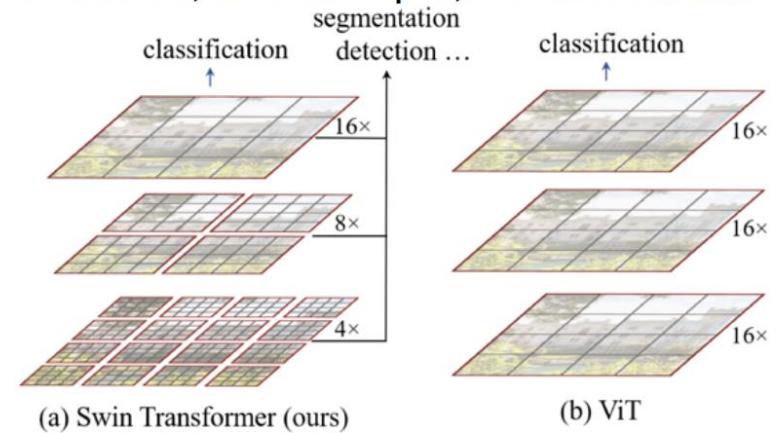


AI-driven Design

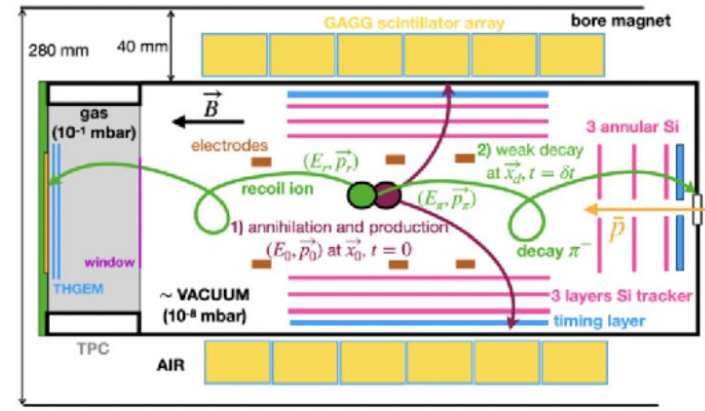
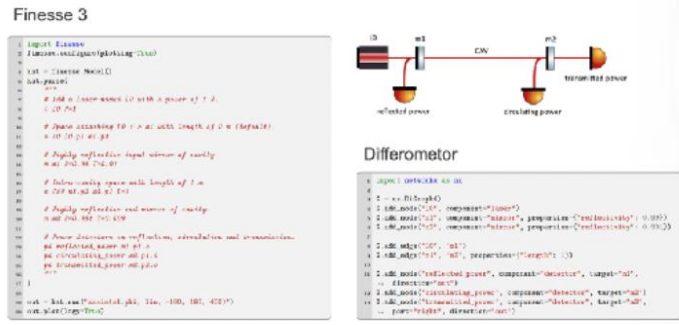
The Framework



E. Jobst, D. Paneque, L. Hackmann



differometer - A Differentiable Interferometer Simulator



Credit Lukas Heinrich (TUM/MPP)
ODSL Days July 8, 2026

Data driven approaches

The ESO internal ALMA Development Study
Bayesian Reconstruction through Adaptive Image
Notion (2019-2024), *PI Fabrizia Guglielmetti*

ALMA Development Studies: Motivation and Context



Why ALMA Development studies?

- Enable community-led innovation beyond standard operations
- Prepare for future scientific and technical challenges
- Test high-risk/high-gain ideas before operational adoption

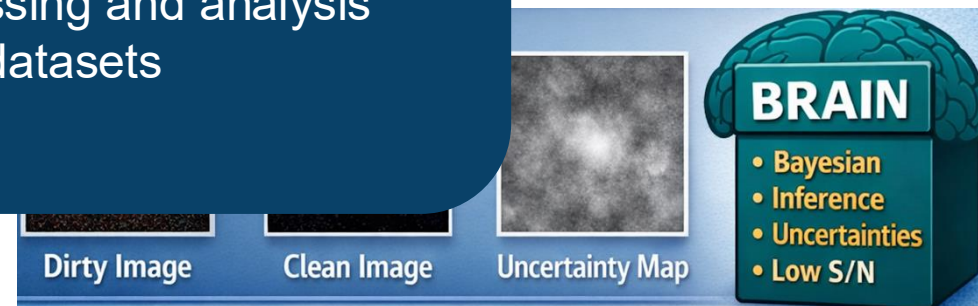
Challenges motivating new approaches

- Increasing sensitivity exposes faint, extended emission
- Complex source morphologies are difficult to image robustly
- Standard techniques may rely on manual tuning
- Limited propagation of uncertainties to science products

The BRAIN Study

- Enhance image analysis using data-driven algorithms
 - Prioritize objective insights over subjective interpretations
 - Achieve more precise and efficient image processing and analysis
- Improve computational efficiency for handling large datasets
- Automate data processing to elevate image quality

Guglielmetti+2019,
Guglielmetti+2022,
Guglielmetti+2023

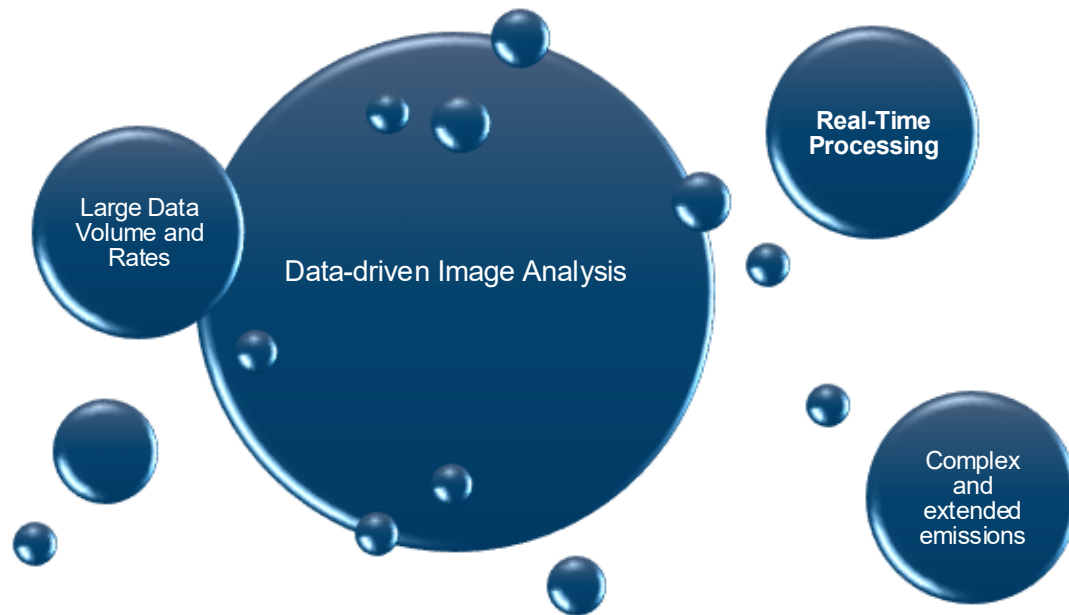


Enhancing Data Processing in Astrophysics: The Brain Study



ALMA2030 Operational Challenges

(Gonzalez, A. et al., 13102, SPIE, 2024; Carpenter, J. et al., ALMA memo 621, 2022)



Solution Pathways:

Cutting-edge Resources and AI Innovation:

- **Expertise in AI and Astronomy:** Engage teams with dual expertise to create advanced algorithms tailored for astronomical data.
- **Enhanced Source Detection Accuracy:** AI-based approaches can refine source detection, leveraging knowledge embedded in Point Spread Functions (PSF), noise patterns, source distribution...

ALMA Simulation Tools:

- **Purpose:** Develop an ALMA simulator designed to generate datasets for AI training, testing and validation.
- **Applications:** This simulator is supported by empirical noise studies and used to validate AI models, ensuring alignment with real-world conditions in the ALMA2030 environment.



Resolve $\subset$ NIFTy $\subset$ UBIK

RESOLVE:

➤ Methodology:

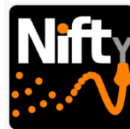
- ✓ Bayesian approach within IFT
- ✓ Synthesizing radio apertures
- ✓ Enabling precise image reconstruction

➤ Key Capabilities:

- ✓ Low signal-to-noise diffuse emissions detection
- ✓ Data combination
- ✓ Self-calibration, Full Polarization imaging

➤ Results:

- ✓ High accuracy in identifying and preserving weak celestial signals, even in noisy observational environments



<https://github.com/information-field-theory/nifty>
<https://ift.pages.mpcdf.de/nifty/index.html>

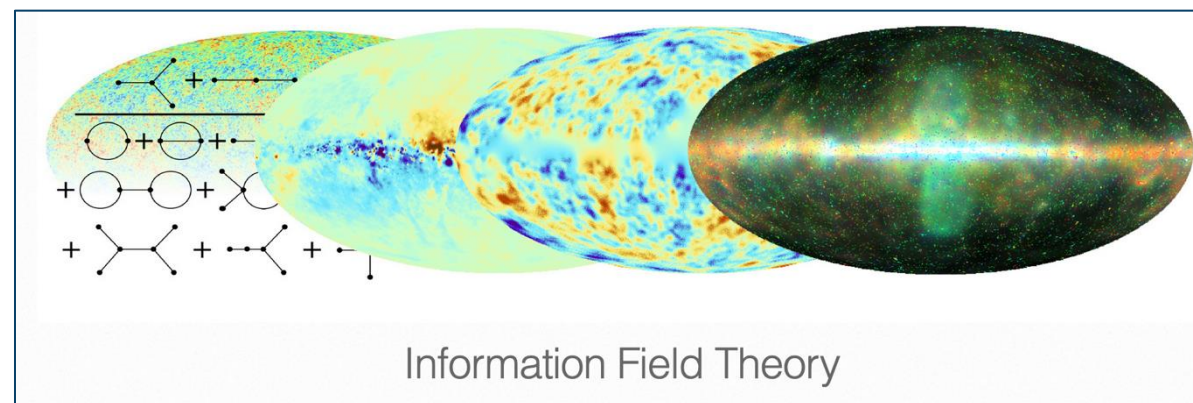


<https://github.com/NIFTy-PPL/J-UBIK>
<https://ift.pages.mpcdf.de/j-ubik/>

<https://gitlab.mpcdf.mpg.de/ift/resolve>
<https://ift.pages.mpcdf.de/resolve/>

Resolve user guide

This guide is an overview and explains the main idea behind resolve. More details are found in the [API reference](#).

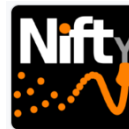


Research, Applications, Software and Publications of the IFT group

RESOLVE:

➤ Methodology:

- ✓ Bayesian approach within IFT
- ✓ Synthesizing radio apertures



<https://github.com/information-field-theory/nifty>
<https://ift.pages.mpcdf.de/nifty/index.html>



<https://github.com/NIFTy-PPL/J-UBIK>
<https://ift.pages.mpcdf.de/j-ubik/>

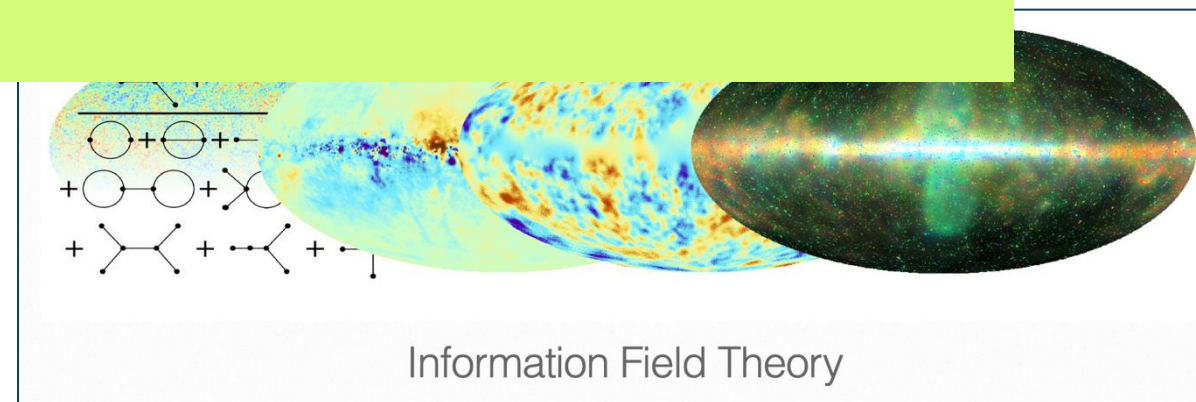
<https://gitlab.mpcdf.mpg.de/ift/resolve>

➤ Key

**Enßlin+2009, Junklewitz+2016, Tychoniec+2022, Roth+2024,
Johnson+2026, Popp+2026, Rüstig+2026**

➤ Results:

- ✓ High accuracy in identifying and preserving weak celestial signals, even in noisy observational environments



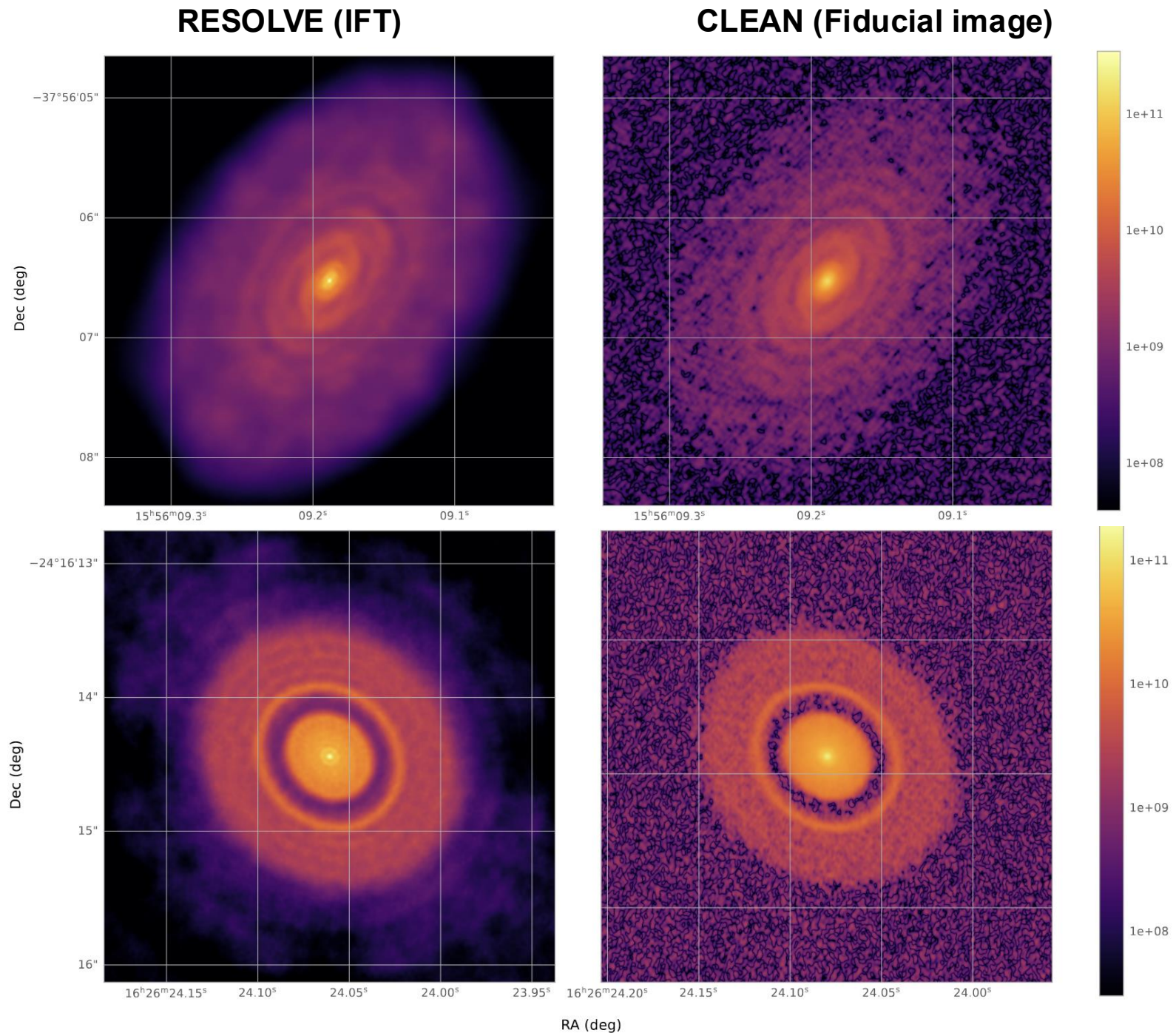
Research, Applications, Software and Publications of the IFT group

Outcome from the BRAIN Study
PI: F. Guglielmetti (ESO)

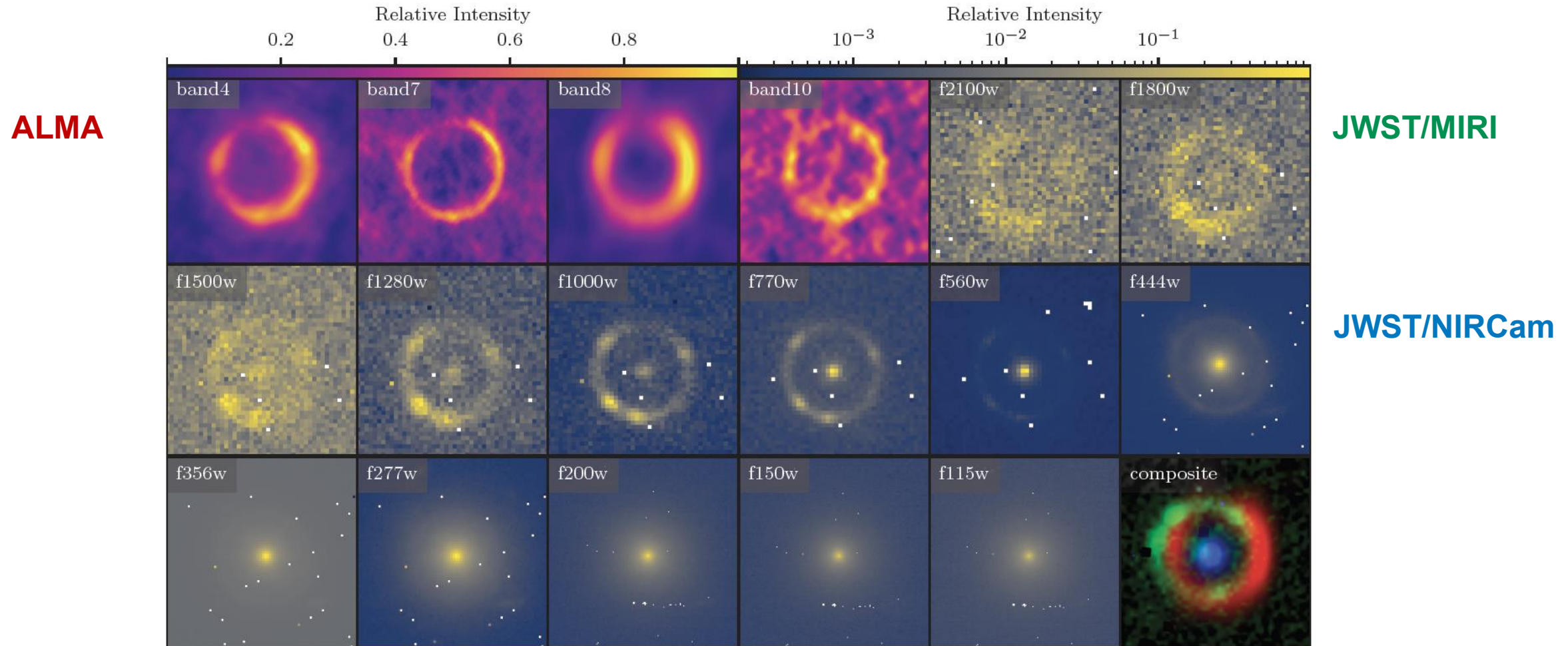
Re-imaging ALMA DSHARP
Large Program (2016.1.00484.L)
with resolve

Johnson et al, in prep

(left) Reconstructed images of **IM Lup** (top) and **Elias 24** (bottom) with **resolve** (IFT algorithm), **revealing extended emission and new features, all consistent with the LAS and not previously detected;**
(right) Corresponding **fiducial images** from the DSHARP team, obtained employing the traditional **CLEAN algorithm**



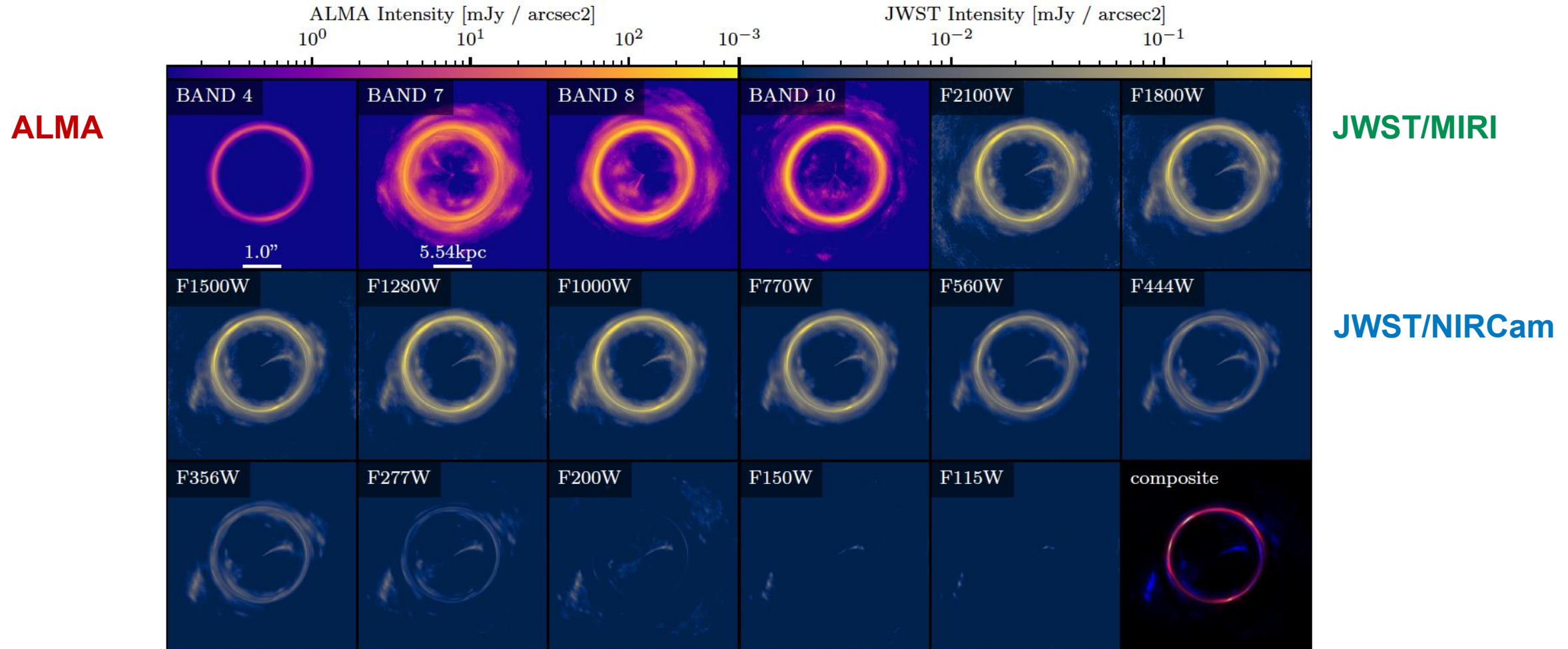
Combining ALMA and JWST data for source reconstruction ($z \sim 4.2$) with Information Field Theory



Ruestig et al, submitted Outcome from the Early-Career Scientific Visitor Programme, 2025

IFT is capable to perform analyses on ALMA multi-bands and JWST multi-filters analyses and to provide a composite image.

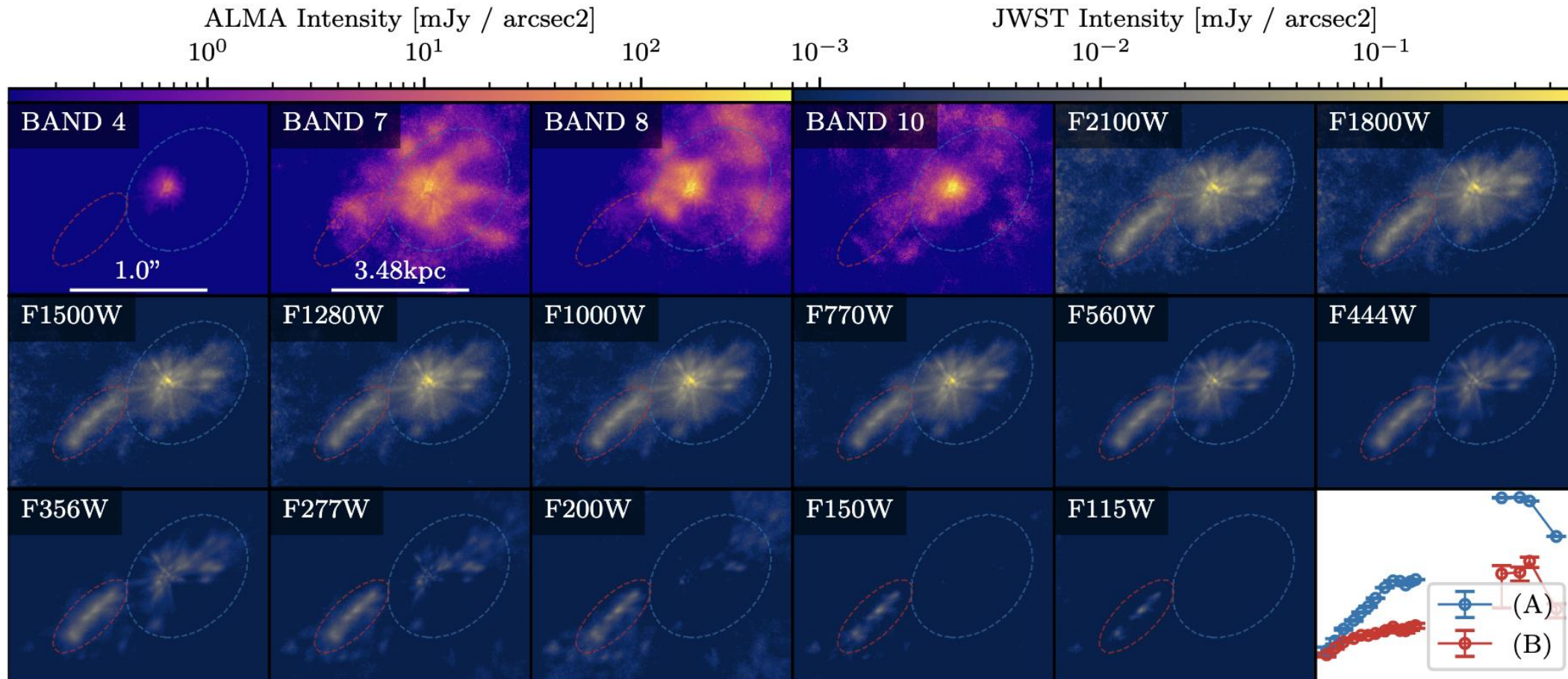
Combining ALMA and JWST data for source reconstruction ($z \sim 4.2$) with Information Field Theory



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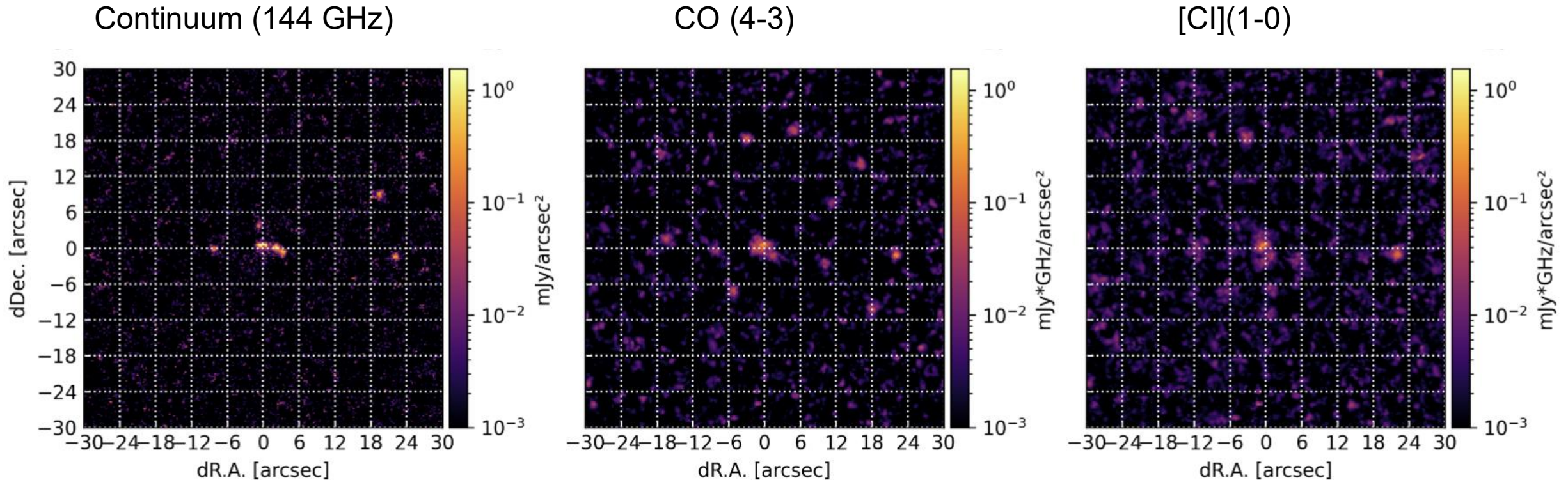
IFT is capable to provide Posterior-mean lensed-light reconstruction across all 17 bands.

Combining ALMA and JWST data for source reconstruction ($z \sim 4.2$) with Information Field Theory



Ruestig et al, submitted Outcome from the Early-Career Scientific Visitor Programme, 2025
IFT is capable to perform joint analyses and to provide the information of 2 merging galaxies.

Targeting the Spiderweb ($z \sim 2.16$) with ALMA in Band 4 with resolve



Popp et al, in prep Outcome from the SSDF project, 2025

Scale independent detection allows us to reveal more sources than previous analyses and to detect faint galaxies connected to the spiderweb.

- Employing 3 ALMA projects

Universal Bayesian Imaging Kit

UBIK

- unifies imaging for a growing number of instruments
- is multi-frequency, -instrument, & -component data analysis
- provides Bayesian uncertainty quantification
- is artificial intelligence without training
- is based on information field theory, NIFTy, & JAX
- is open source



Credit T. Ensslin (MPA/DZA)
ESO/DZA partnership

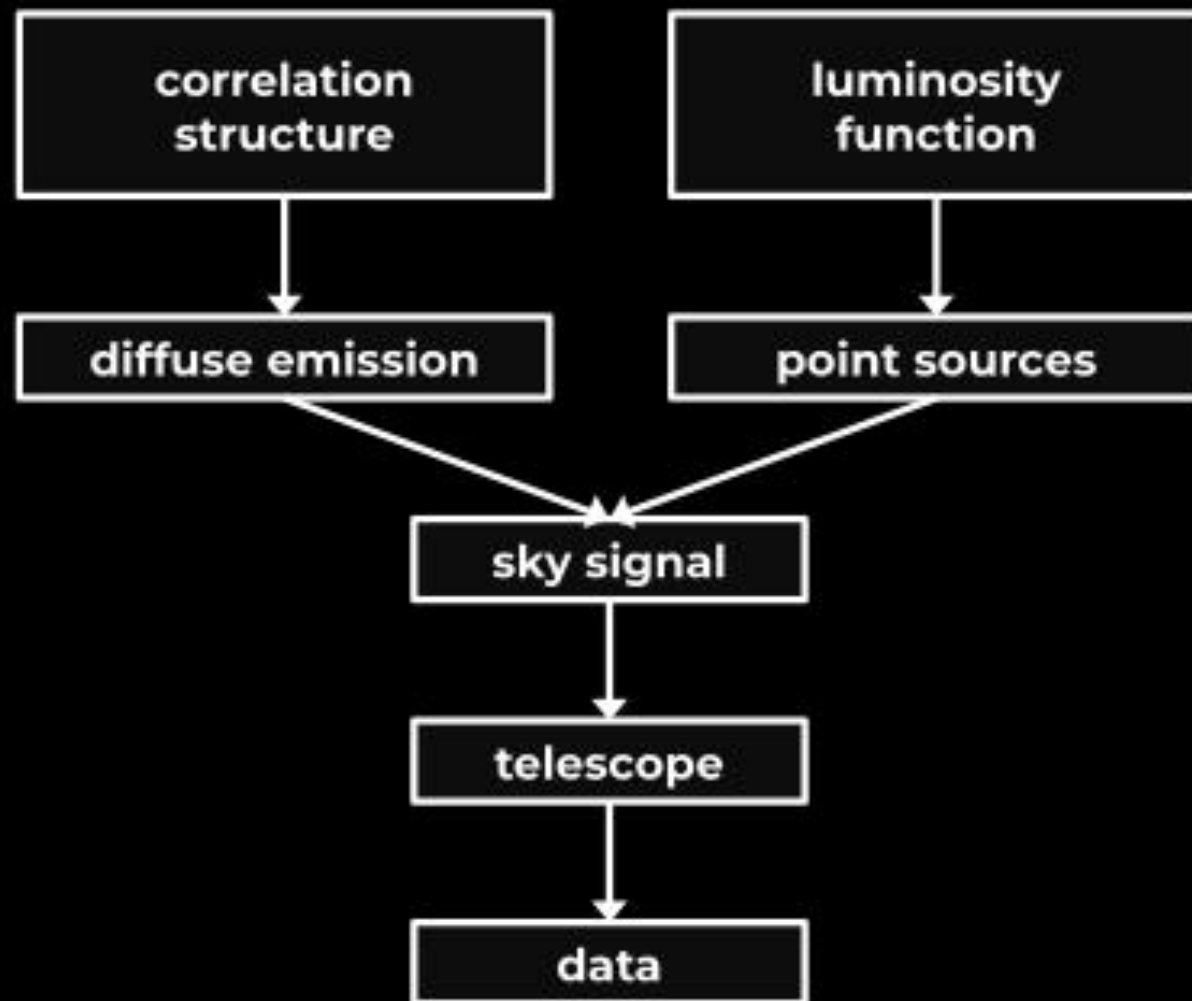


Digital Perception of Reality

Credit T. Ensslin (MPA/DZA)

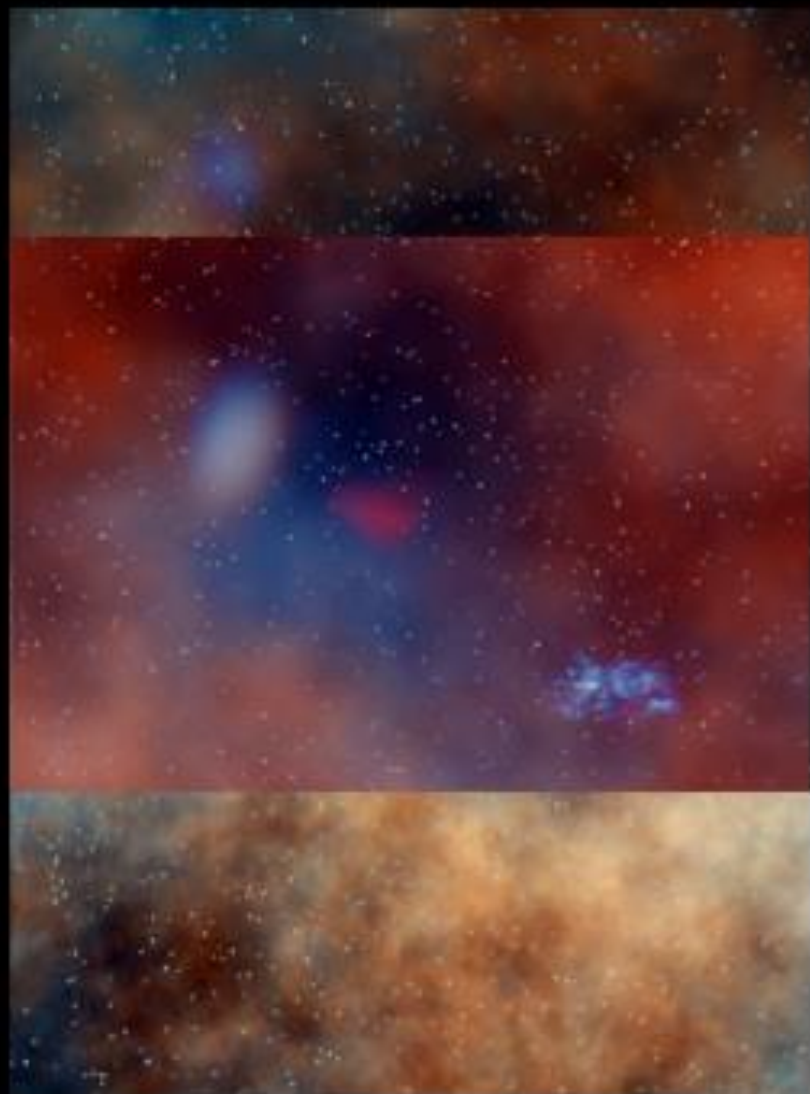


generative model
= simulation

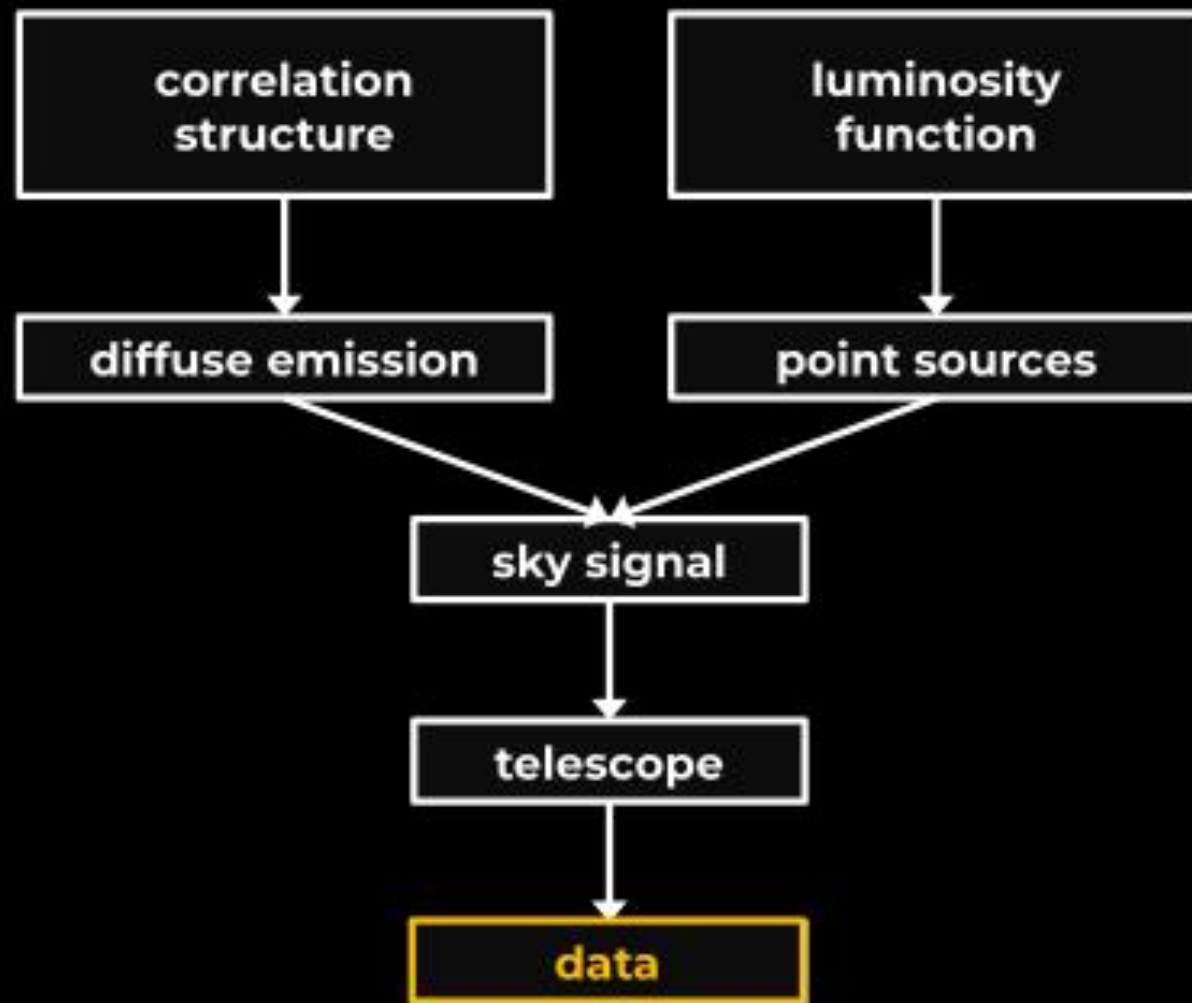


Digital Perception of Reality

Credit T. Ensslin (MPA/DZA)



back propagation
= inference



LIKELIHOOD MODEL

PRIOR MODEL

Space

Dimensions

Time

Energy

Polarization

...

Sources

Diffuse emission

Point sources

Emission lines

...

Signal

Noise

Instrument

Models



Numerics

Inference

MAP

VI

HMC

...

geoVI

MGVI

NIFTY

Minimizer

Autodiff

...

Credit T. Ensslin (MPA/DZA)



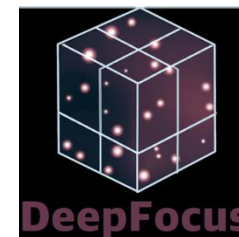
Deep Learning

DeepFocus:



➤ Overview:

- ✓ **Deep Learning** architecture
- ✓ **Meta-learner:** Taxonomy with Bayesian Surrogates
- ✓ Designed to analyze and process **large** ALMA data cubes



<https://github.com/MicheleDelliVeneri/DeepFocus>
Delli Veneri +2023

<https://github.com/MicheleDelliVeneri/ALMASim>

<https://github.com/lbaronch/NOISEMPIRE>. Baronchelli, I.

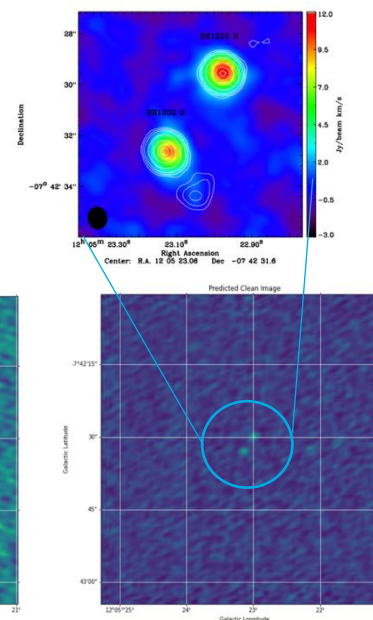
➤ Enhanced by Simulation:

- ✓ Utilizes **ALMASIM** for realistic data generation and **NOISEMPIRE** for accurate noise representation, ensuring fidelity to real ALMA observations

➤ Outcomes:

- ✓ Real-time image analysis
- ✓ Catalogues of extracted sources

Carniani et al.+2013



Simulated data

Real data

INPUT DIRTY CUBE



DEEP FOCUS



TAXONOMY



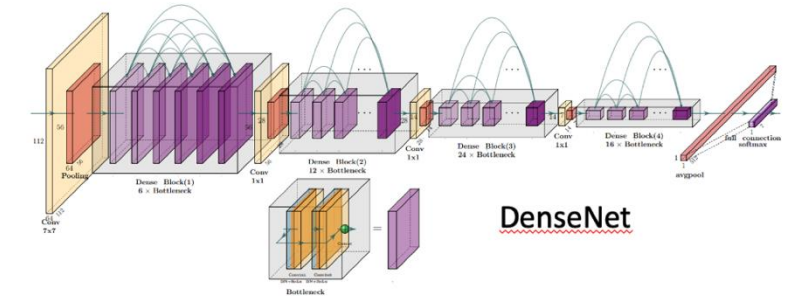
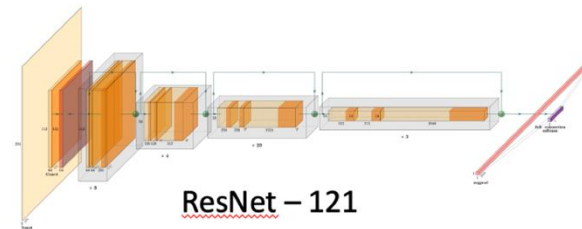
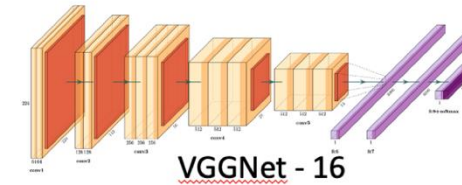
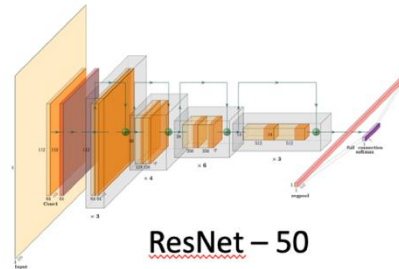
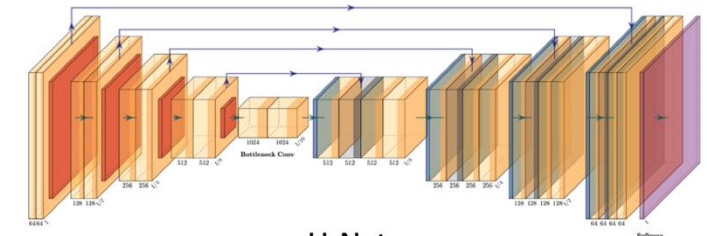
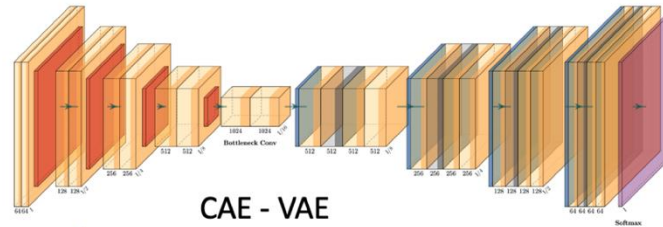
DECONVOLUTION
AND DENOISING



FOCUSING



REGRESSION AND
CLASSIFICATION



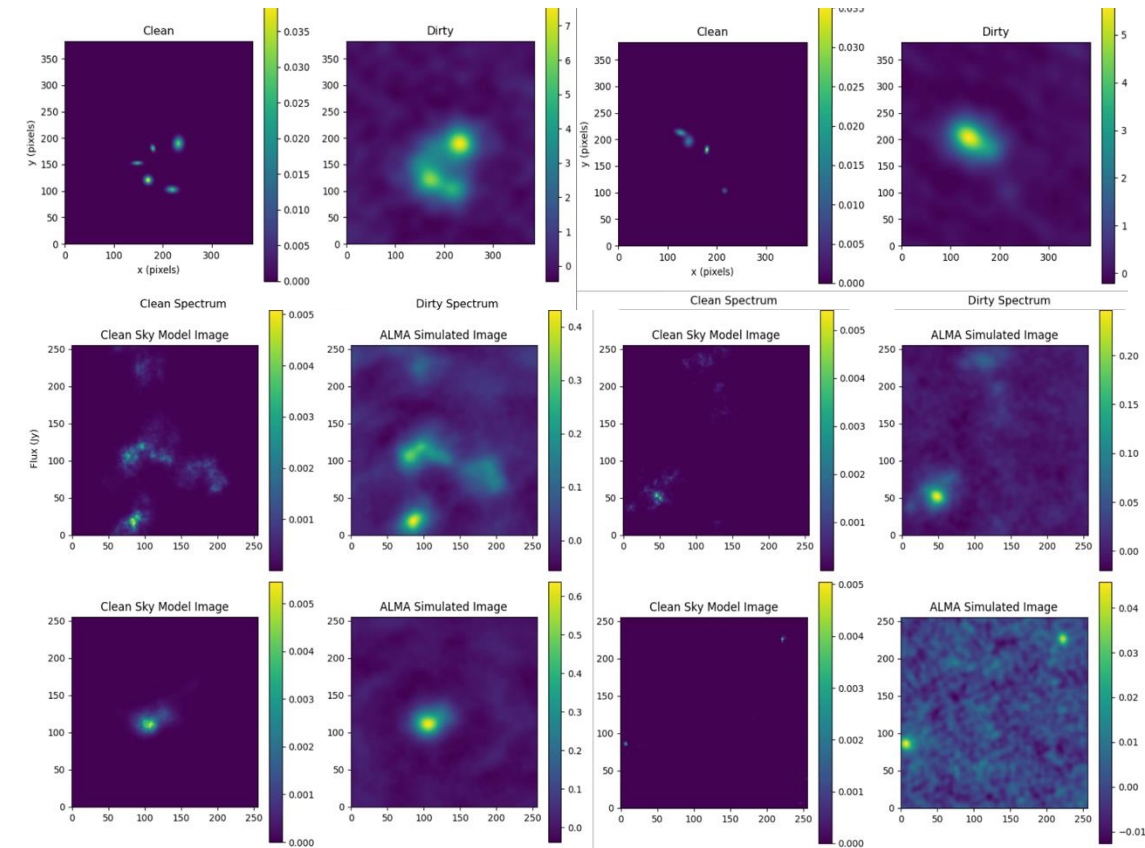
<https://github.com/lbaronch/NOISEMPIRE>

<https://github.com/MicheleDelliVeneri/ALMASim.git>

<https://github.com/MicheleDelliVeneri/DeepFocus>

ALMASim (Simulator)

- **Metadata retrieval** : query the ALMA Archive
- **Source reasoning** :
 - Extract fitting lines for the observation window
 - Calculate continuum and line fluxes
- **Sky model generation** :
 - Serendipitous sources
 - Storing source properties and observations on disk
- **Mock ALMA Observations** :
 - Accounts for atmospheric noise and gain errors
 - Save data products to disk



ALMASim: the ALMA Simulator for AI Developments



ALMASim: set up your simulation parameters

Output Directory: /Volumes/Storage/almasim-test Browse

TNG Directory: /Users/michele/TNGData Browse

Galaxy Zoo Directory: /Users/michele/GalaxyZoo Browse ☐ Get Galaxy Zoo Data

Hubble Top 100 Directory: /Users/michele/HubbleData Browse ☐ Get Hubble 100 Data

Email:

Project Name: test-simulation

Number of Simulations: 10

N. CPUs / Processes: 100

Save Format: fits

Computation Mode: parallel

Local or Remote: local

☒ Line Mode Select Line Indices (space-separated): 21

Min/Max Line Widths in Km/s: 365 km/s 503 km/s

Briggs Robustness: 0.5

☒ Set SNR 1.3 ☒ Set IR Luminosity 1e10 ☐ Fix Spatial Dim ☐ Fix Spectral Dim ☒ Inject Serendipitous

Select Model: Point

Metadata Retrieval Mode: get

Metadata Path: /Users/michele/GitHub/ALMASim/almasim/metadata/tester.csv Browse

Reset Start Simulation Stop

60: HCN(0-0) 963.62 - 9.32e+02 GHz
61: HCN(7-6) 483.3 - 6.20e+02 GHz
62: HCN(8-7) 422.91 - 7.09e+02 GHz
63: HCO+(3-2) 1120.48 - 2.68e+02 GHz
64: HCO+(4-3) 840.38 - 3.57e+02 GHz
65: HCO+(5-4) 672.33 - 4.46e+02 GHz
66: HCO+(6-5) 560.3 - 5.35e+02 GHz
67: HCO+(7-6) 480.28 - 6.24e+02 GHz
68: HCO+(8-7) 420.27 - 7.13e+02 GHz
69: H2 12.37 - 2.42e+04 GHz
70: HNC(3-2) 1102.25 - 2.72e+02 GHz
71: HNC(4-3) 826.72 - 3.63e+02 GHz
72: HNC(5-4) 661.4 - 4.53e+02 GHz
73: HNC(6-5) 551.19 - 5.44e+02 GHz
74: H2 12.28 - 2.44e+04 GHz
75: H2 17.03 - 1.76e+04 GHz
76: H2 6.91 - 4.34e+04 GHz
77: H2 9.66 - 3.10e+04 GHz
78: LiH(1-0) 675.28 - 4.44e+02 GHz
79: N2H+(5-4) 643.57 - 4.66e+02 GHz
80: N2H+(6-5) 536.33 - 5.59e+02 GHz
81: N2H+(8-7) 402.29 - 7.45e+02 GHz
82: NH3(10-00) 523.66 - 5.72e+02 GHz
83: NII 121.90 - 2.46e+03 GHz
84: NII 205.18 - 1.46e+03 GHz
85: NIII 57.32 - 5.23e+03 GHz
86: NeII 12.81 - 2.34e+04 GHz
87: NeIII 15.55 - 1.93e+04 GHz
88: OH 446 - 4.46e+02 GHz
89: OI 145.52 - 2.06e+03 GHz
90: OI 63.18 - 4.75e+03 GHz
91: OIII 51.81 - 5.79e+03 GHz
92: OIII 88.36 - 3.39e+03 GHz
93: PAH 11.25 - 2.66e+04 GHz
94: PAH 12.7 - 2.36e+04 GHz
95: PAH 3.3 - 9.08e+04 GHz
96: PAH 6.2 - 4.84e+04 GHz
97: PAH 7.7 - 3.89e+04 GHz
98: PAH 8.6 - 3.49e+04 GHz
99: SiII 18.71 - 1.60e+04 GHz
100: SiII 33.48 - 8.95e+03 GHz
101: SiV 10.49 - 2.86e+04 GHz
102: SO(6-5) 1363.01 - 2.20e+02 GHz
103: SO(7-6) 1174.34 - 2.55e+02 GHz
104: SiII 34.82 - 8.61e+03 GHz
105: SiO(12-11) 575.55 - 5.21e+02 GHz
106: SiO(14-13) 493.4 - 6.08e+02 GHz
107: SiO(16-15) 431.81 - 6.94e+02 GHz
108: [C](1-0) 609.14 - 4.92e+02 GHz
Loading metadata from /Users/michele/GitHub/ALMASim/almasim/metadata/tester.csv
Metadata contains 1 samples
Setting file path is /Users/michele/Library/Preferences/com.infn-section-of-naples.ALMASim.plist

Figure 33: The ALMASim Graphical User Interface.

Deep Learning Models: Future extensions

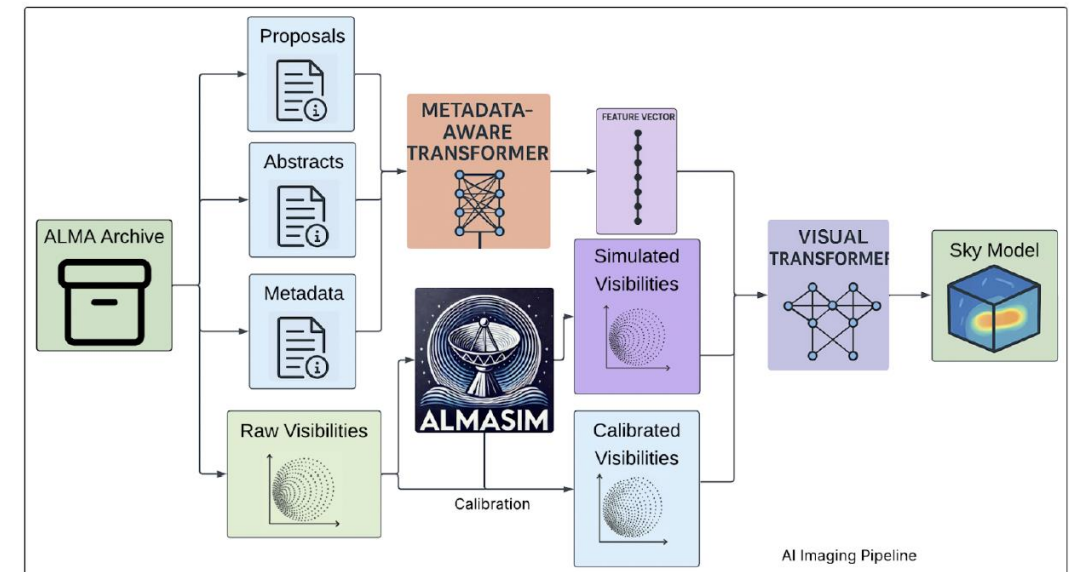


MADIV: a Metadata Aware transformer for Direct Imaging from Visibilities (*P.I. Michele Delli Veneri*)

Approved ESO external ALMA Development Study

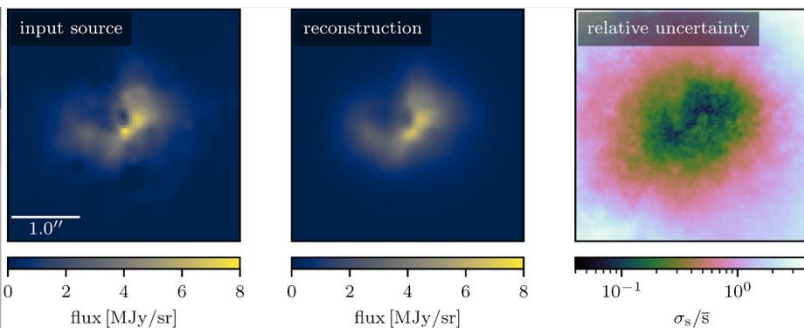
ESO/SKA partnership

- *Visual Transformer architecture as deconvolver/imager operating on visibility domain*
- *Metadata-Aware Transformer as prior knowledge on signal distribution*
- *ALMASim will learn from ALMA Archival data*
- *Diffusion models will be used to create simulated data*
- **Applications to ALMA Archival data**

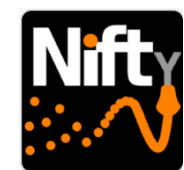




Potential AI applications for ALMA

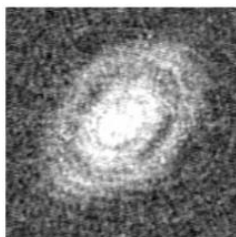


Rüstig+2024



RESOLVE
IFT + UQ
Improves
accuracy in
ALMA imaging

Baronchelli+2024



**Simulated
observation**

NOISEMPIRE 1.0.3 Noce

Empirical noise simulator

Simulates pure noise 2D images given a real ALMA cube or image in input.
A real sky image (same size as reference image) can also be added.



Real-time **ALARM**
Systems for
Observatory's
health?

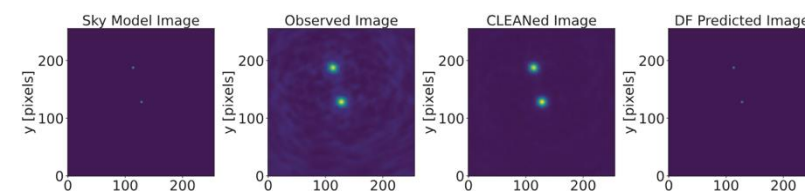
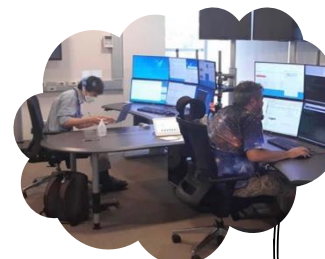


Fig. 1. – From left to right: Sky Model Image, Observed Image, CLEAN image, DF Prediction of the Sky Model.

Delli Veneri+2024



DeepFocus

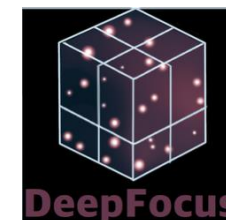
Meta Learner

Enables

Real-time

ALMA imaging

+ cataloguing



ALMASim

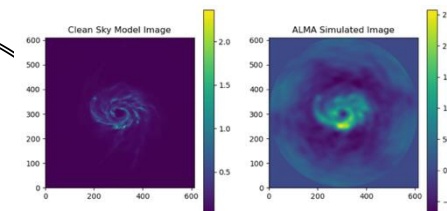
Encodes +100

Molecular

transitions at

several z, **HPC**

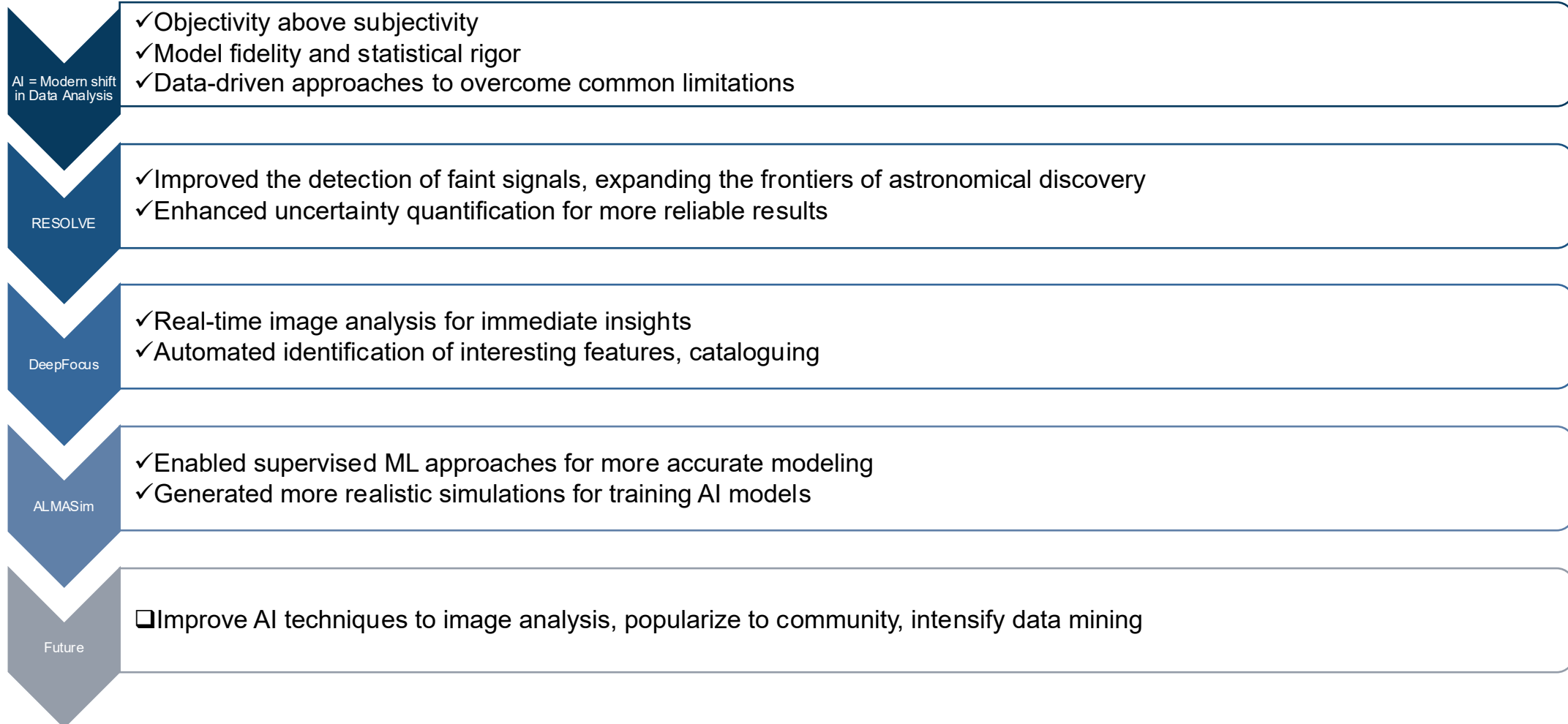
enabled



Delli Veneri, ALMASim
documentation



Conclusion on Imaging beyond CLEAN



Thank you!



the BRAIN Study Team

fgugliel@eso.org

<https://gitlab.mpcdf.mpg.de/ift/resolve>

<https://github.com/NIFTy-PPL>

<https://github.com/MicheleDelliVeneri/DeepFocus>

<https://pypi.org/project/ALMASim>

<https://github.com/lbaronch/NOISEMPIRE>



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