

Statistical inference in Python

the **NIFTY** way



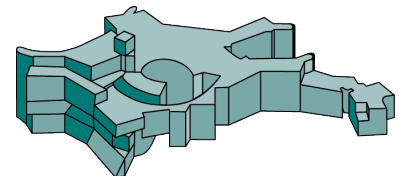
Michael R. Bell
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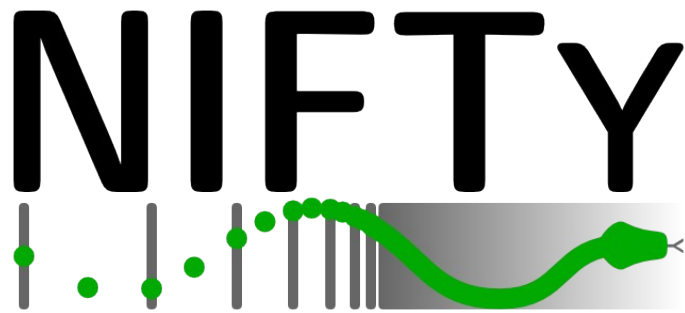


Marco Selig, Henrik Junklewitz, Niels Oppermann,
Martin Reinecke, Maksim Greiner, Carlos Pachajoa, Torsten A. Enßlin



*Max Planck Institute for Astrophysics,
Munich, Germany*





is a Python framework for developing statistical inference algorithms independent of the underlying geometry or resolution.

Follow along at: <http://tinyurl.com/nifty-demo-slides>

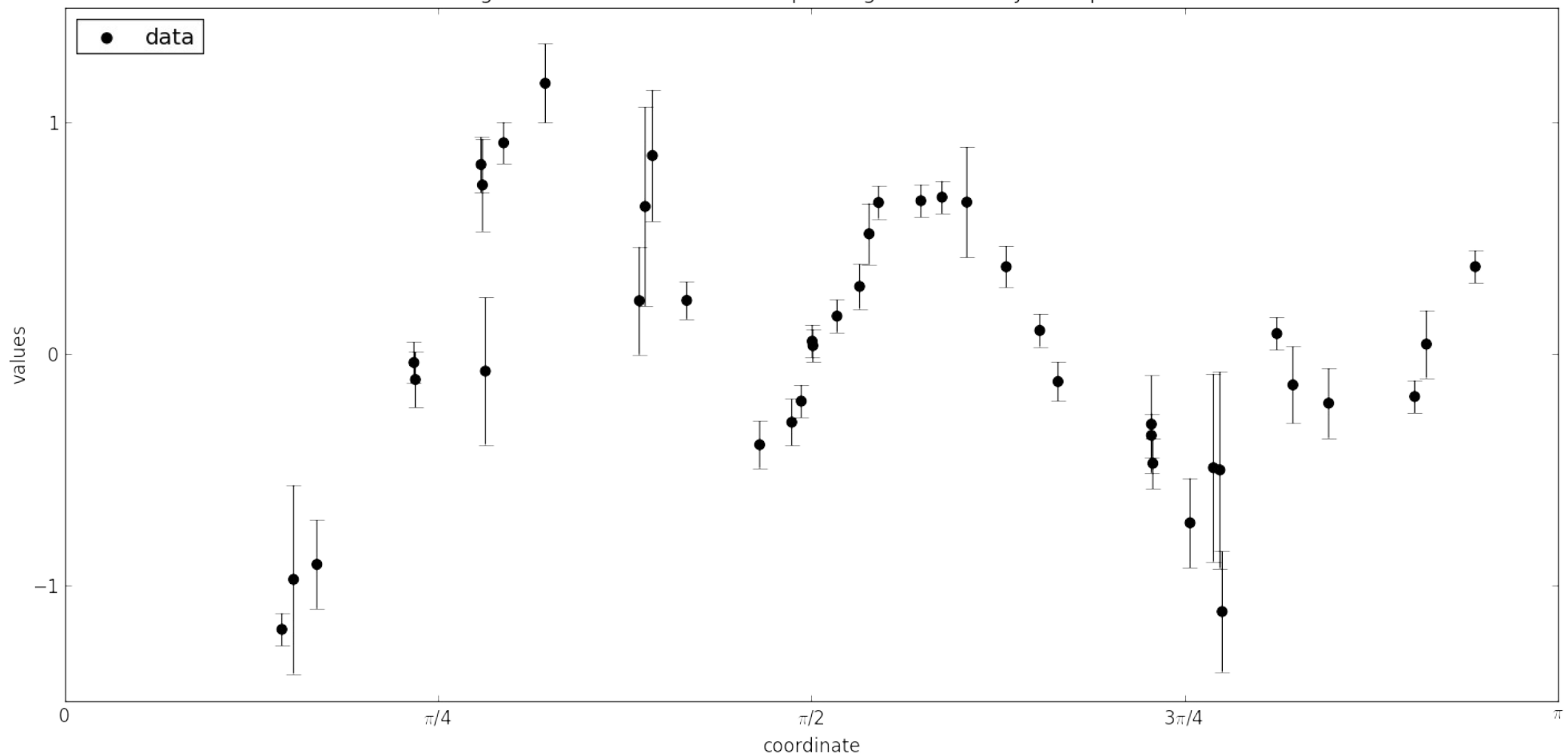
Data & Signal

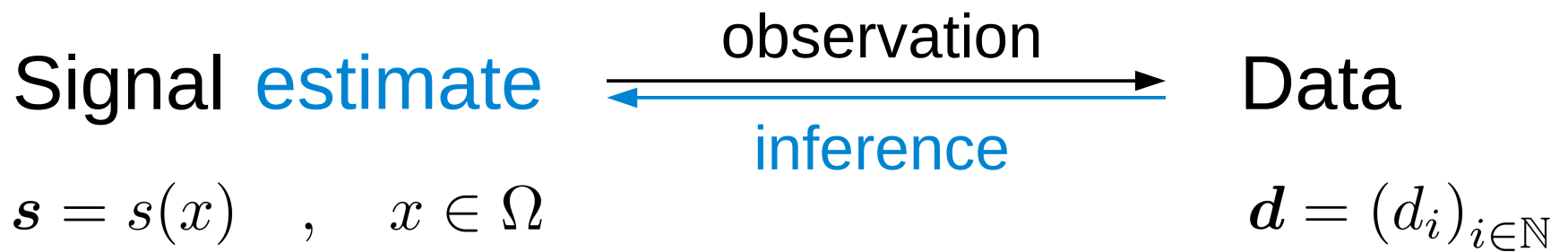
Signal $\xrightarrow{\text{observation}}$ Data

$$s = s(x) \quad , \quad x \in \Omega$$

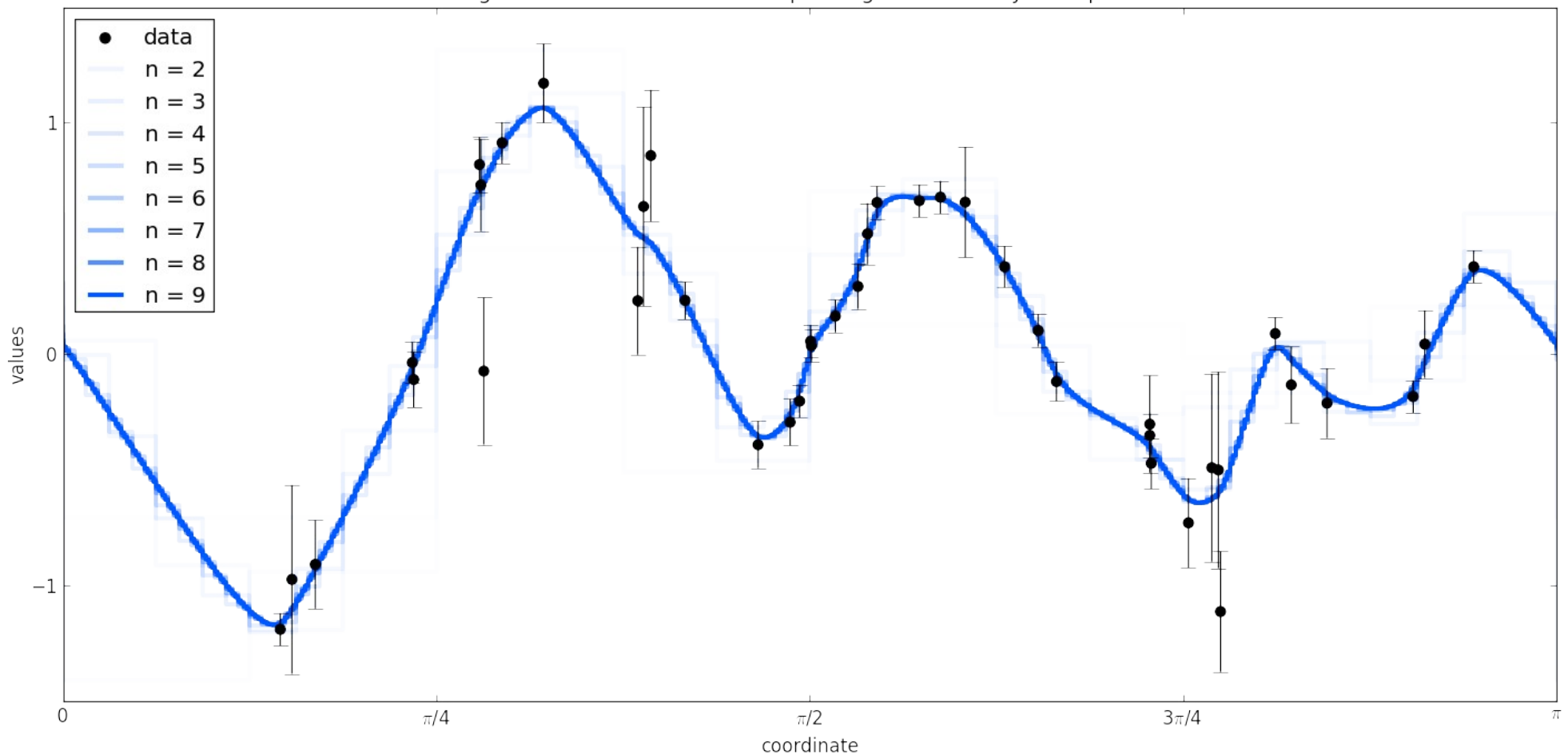
$$d = (d_i)_{i \in \mathbb{N}}$$

signal reconstruction with 2^n pixels given 42 noisy data points

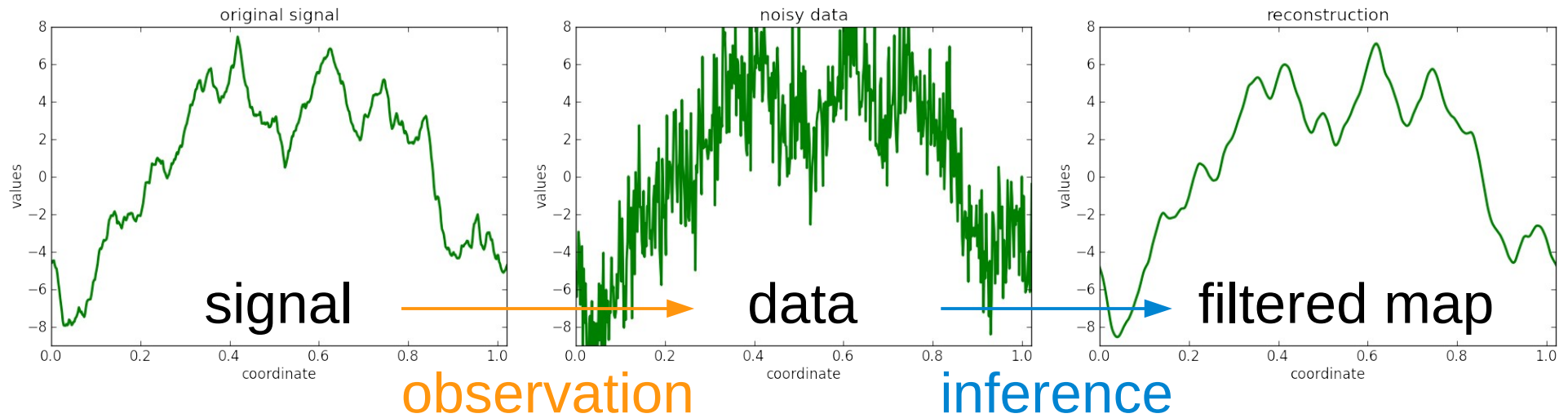




signal reconstruction with 2^n pixels given 42 noisy data points



Wiener filtering



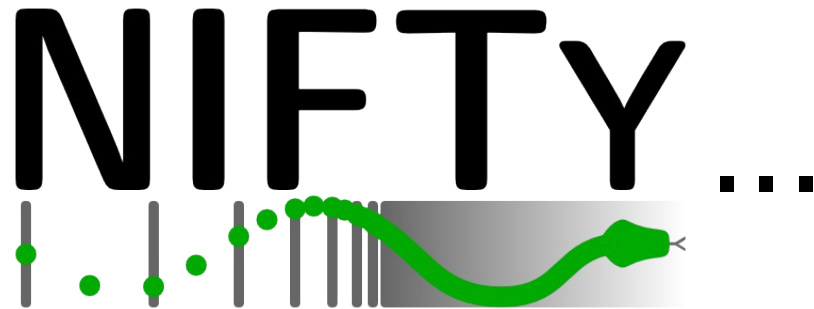
$$s \sim \mathcal{G}(s, S)$$

$$n \sim \mathcal{G}(n, N)$$

$$d = Rs + n$$

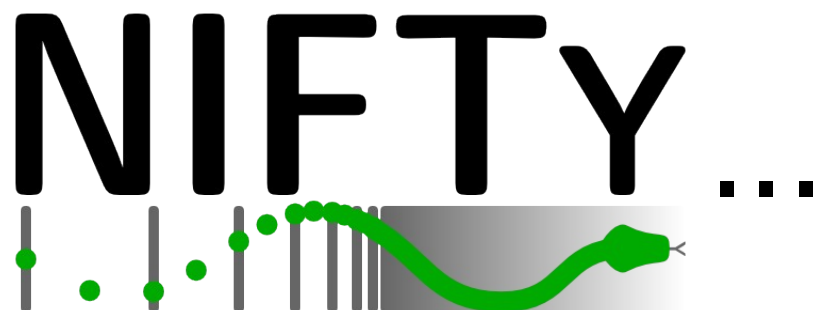
$$m = \underbrace{\left(S^{-1} + R^{\dagger} N^{-1} R \right)^{-1}}_D \underbrace{\left(R^{\dagger} N^{-1} d \right)}_j$$





Selig et al. (2013)
arXiv: [1301.4499](https://arxiv.org/abs/1301.4499)

- is a versatile PYTHON library incorporating CYTHON, C++, and C libraries for performance

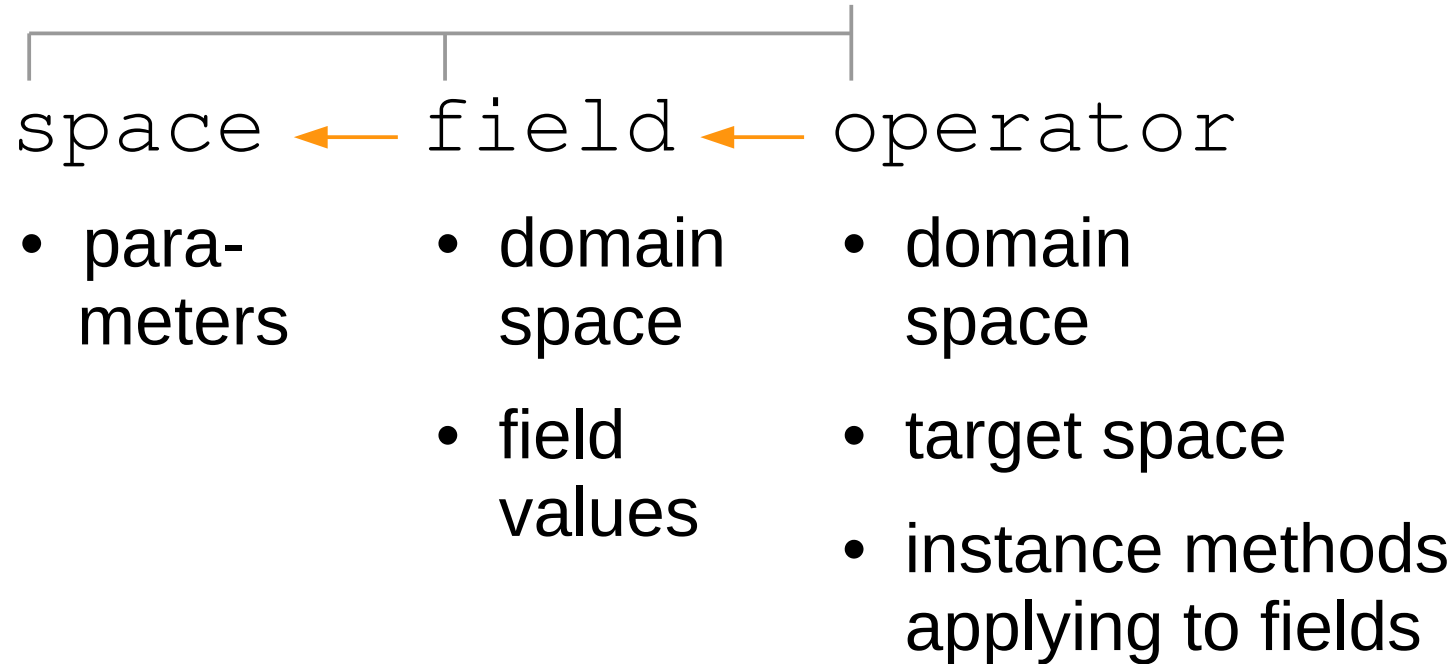


Selig et al. (2013)
arXiv: [1301.4499](https://arxiv.org/abs/1301.4499)

- is a versatile PYTHON library incorporating CYTHON, C++, and C libraries for performance
- **abstracts spaces, fields, and operators into an object-orientated framework**

NIFTY classes

object

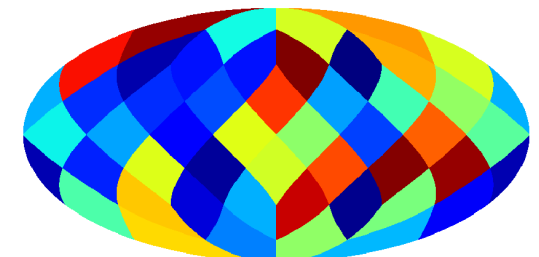
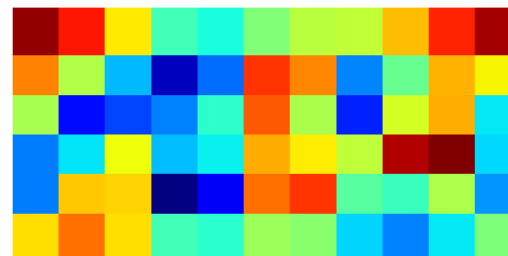
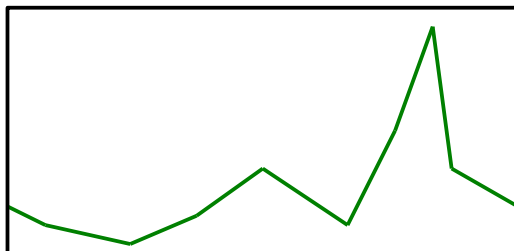


NIFTY classes

object



point_space	• unstructured list of points
rg_space	• n-dimensional regular grid
lm_space	• spherical harmonics
hp_space	• HEALPix grid
gl_space	• Gauss-Legendre grid
nested_space	• (arbitrary product of grids)



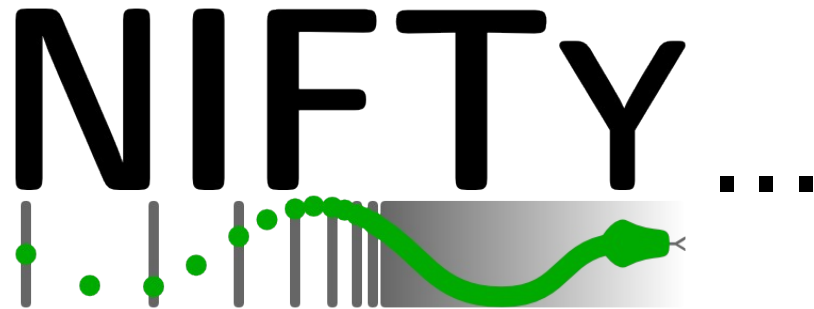
NIFTY classes

object

space ← field ← operator ← probing

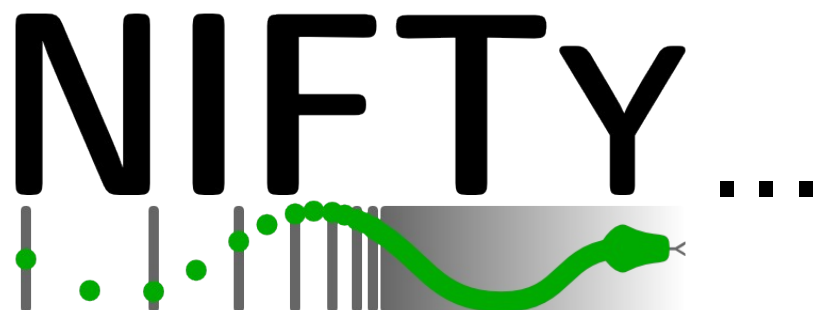
- point_space
- rg_space
- lm_space
- hp_space
- gl_space
- nested_space

- diagonal_operator
 - power_operator
- projection_operator
- vecvec_operator
- response_operator
- invertible_operator
- propagator_operator
- explicit_operator



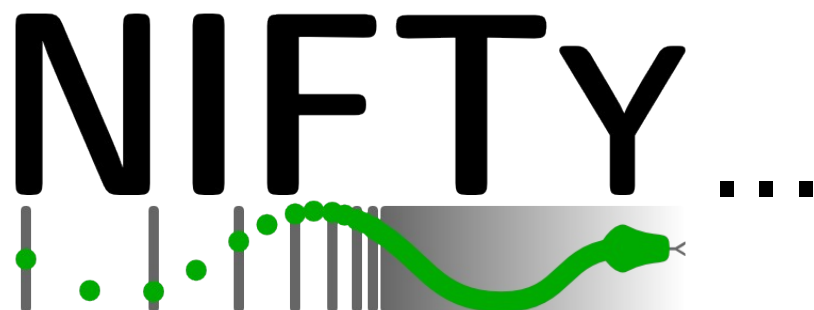
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- abstracts spaces, fields, and operators into an object-orientated framework
- **operates regardless of the underlying spatial grid and its resolution**



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arXiv: [1301.4499](https://arxiv.org/abs/1301.4499)

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- operates regardless of the underlying spatial grid and its resolution
- **provides useful tools for the development of statistical inference algorithms (optimization, operator probing, visualization, etc.)**




Selig et al. (2013)
arXiv: [1301.4499](https://arxiv.org/abs/1301.4499)

- is a versatile PYTHON library incorporating CYTHON, C++, and C libraries for performance
- abstracts spaces, fields, and operators into an object-orientated framework
- operates regardless of the underlying spatial grid and its resolution
- provides useful tools for the development of statistical inference algorithms (transformations, operator probing, visualization, etc.)
- **includes extensive on-line documentation**

NIFTY project homepage: <http://www.mpa-garching.mpg.de/ift/nifty/>

NIFTY 0.2.0 documentation »

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NIFTY - Numerical Information Field Theory

NIFTY [1], “Numerical Information Field Theory”, is a versatile library designed to enable the development of signal inference algorithms that operate regardless of the underlying spatial grid and its resolution. Its object-oriented framework is written in Python, although it accesses libraries written in Cython, C++, and C for efficiency.

NIFTY offers a toolkit that abstracts discretized representations of continuous spaces, fields in these spaces, and operators acting on fields into classes. Thereby, the correct normalization of operations on fields is taken care of automatically without concerning the user. This allows for an abstract formulation and programming of inference algorithms, including those derived within information field theory. Thus, NIFTY permits its user to rapidly prototype algorithms in 1D and then apply the developed code in higher-dimensional settings of real world problems. The set of spaces on which NIFTY operates comprises point sets, n -dimensional regular grids, spherical spaces, their harmonic counterparts, and product spaces constructed as combinations of those.

References

[1] M. Selig et al., “NIFTY – Numerical Information Field Theory – a versatile Python library for signal inference”, submitted to IEEE, 2013; [arXiv:1301.4499](https://arxiv.org/abs/1301.4499)

Note: Parts of this publication can with or without modification be found within the source code and this online documentation for obvious reasons, and they are not explicitly marked as quotations.

Documentation

Welcome to NIFTY’s documentation!

Tip: For a quickstart, [download](#) NIFTY and browse through the *informal introduction*.

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- IFT – Information Field Theory
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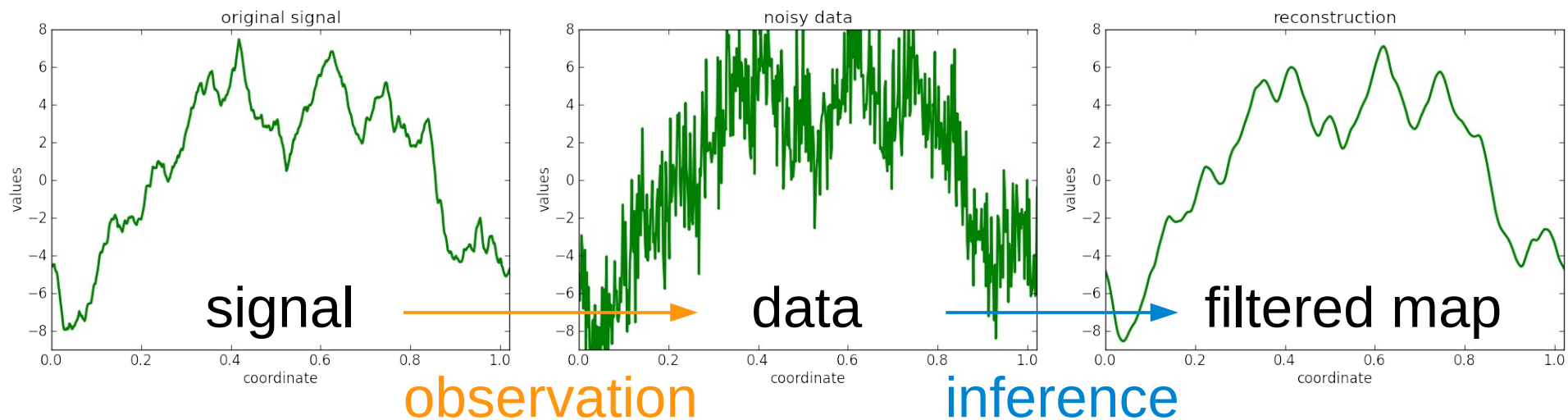
NIFTY 0.2.0 documentation »

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Demo

Wiener filtering, revisited



$$d = Rs + n \quad s \sim \mathcal{G}(s, S) \quad n \sim \mathcal{G}(n, N)$$

$$m = \langle s \rangle_{(s|d)} = \underbrace{\left(S^{-1} + R^{\dagger} N^{-1} R \right)^{-1}}_D \underbrace{\left(R^{\dagger} N^{-1} d \right)}_j$$

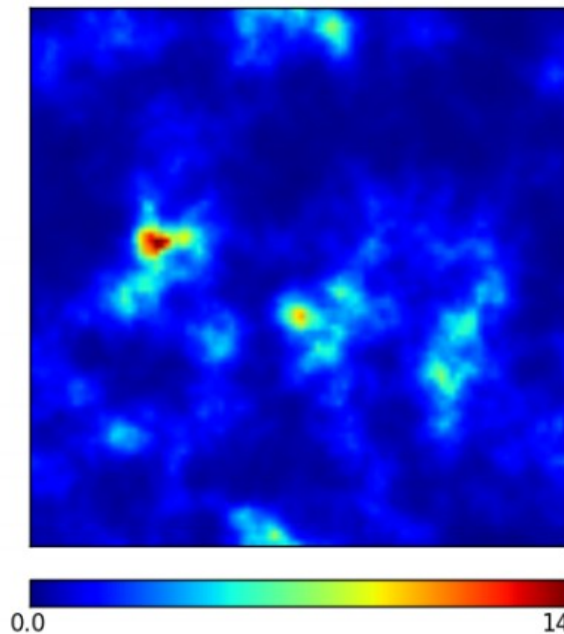
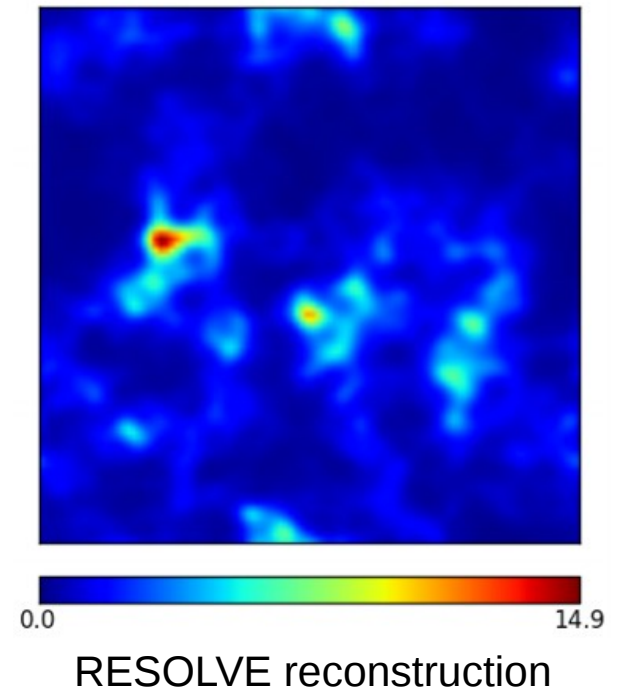
Go here to follow along: <http://tinyurl.com/nifty-demo-notebook>

Applications

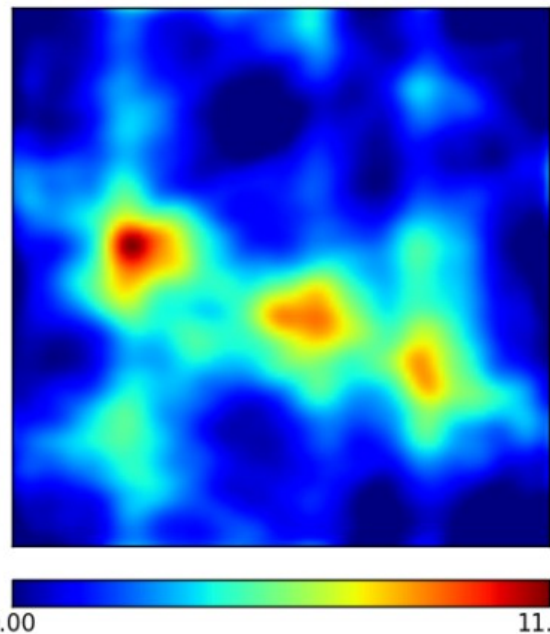
The RESOLVE algorithm:
Improved image reconstruction
for radio astronomy.

Narrow bandwidth image
reconstruction.

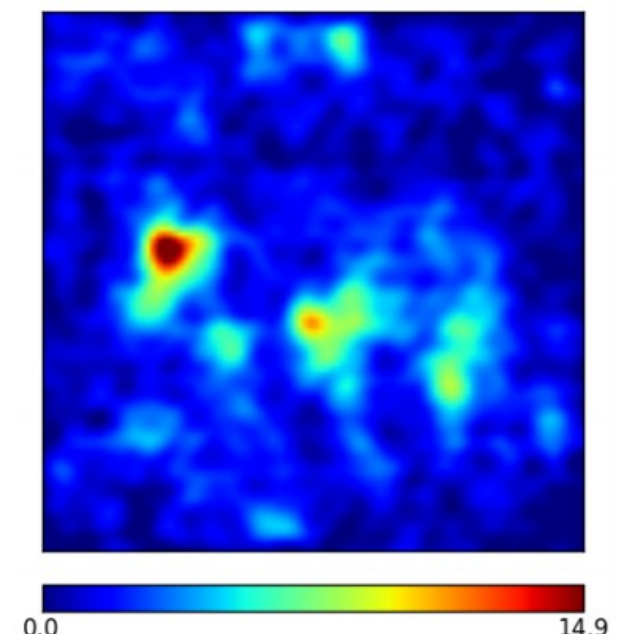
Junklewitz, Bell, Selig, Enßlin (2014)
<http://arxiv.org/abs/1401.4711>



Original signal



Naive reconstruction



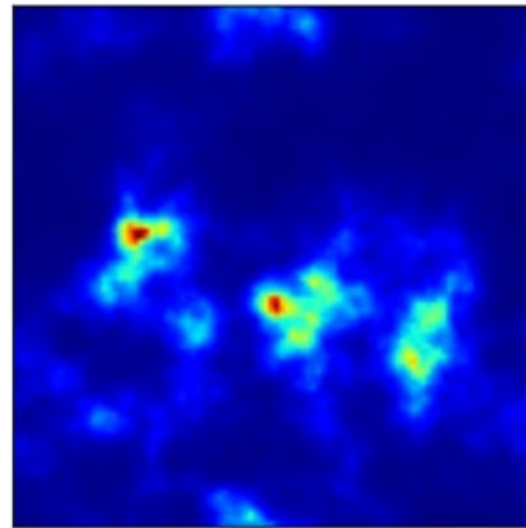
Typical reconstruction

The RESOLVE algorithm: Improved image reconstruction for radio astronomy.

Broad bandwidth image
reconstruction.

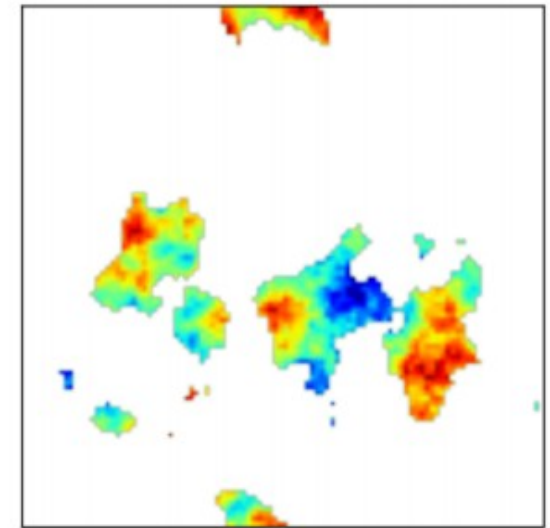
Junklewitz, Bell, Enßlin (2014)
<http://arxiv.org/abs/1311.5282>

Original signal



0.00

27.14



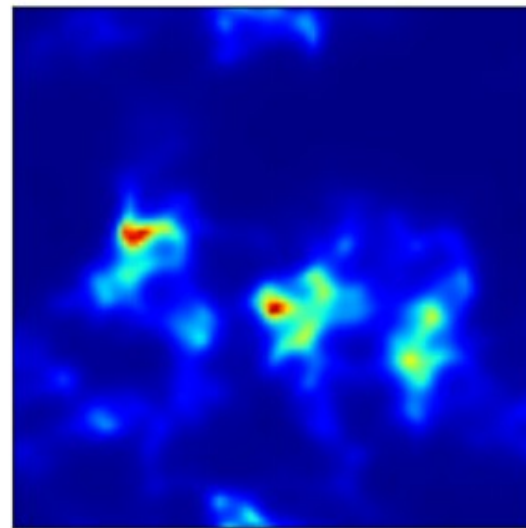
-2.23

0.89

Brightness at a ref. frequency

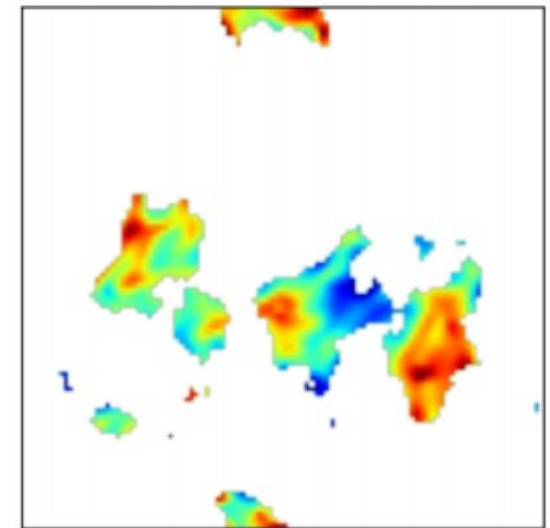
Slope of power law spectra

RESOLVE reconstruction



0.00

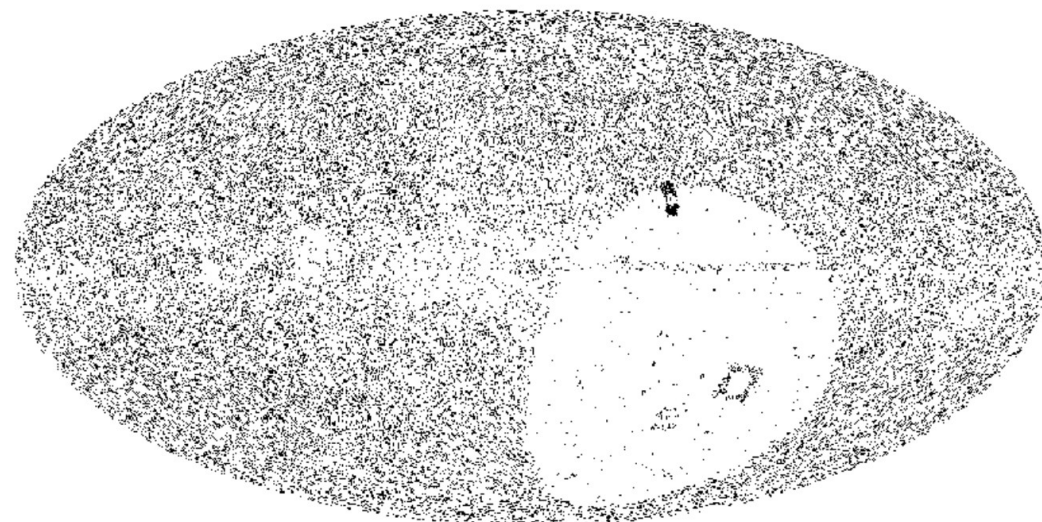
27.14



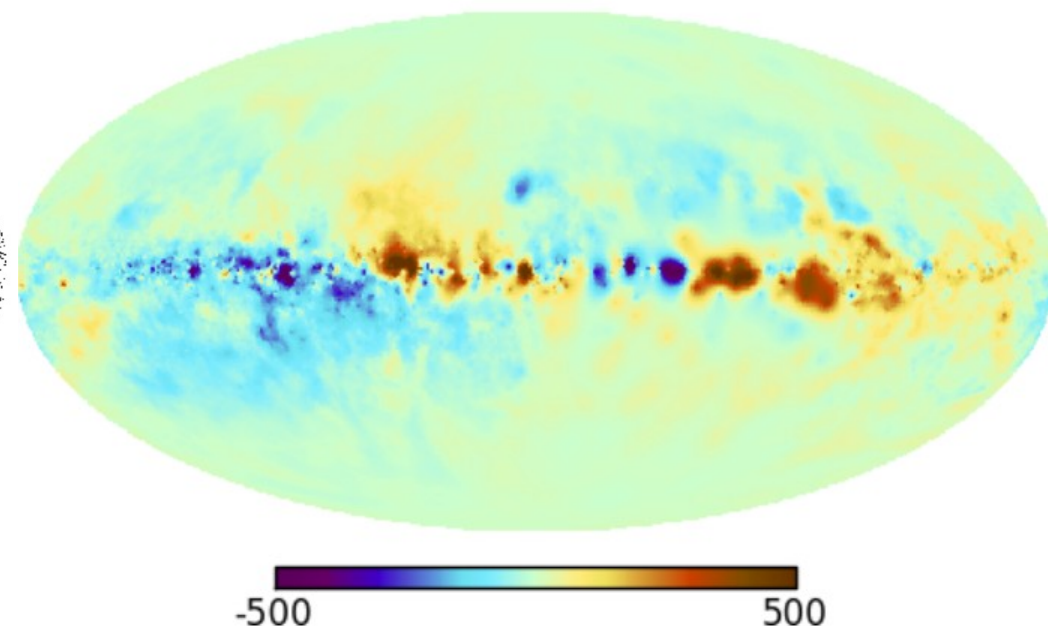
-2.23

0.89

Locations of 41,330 measurements



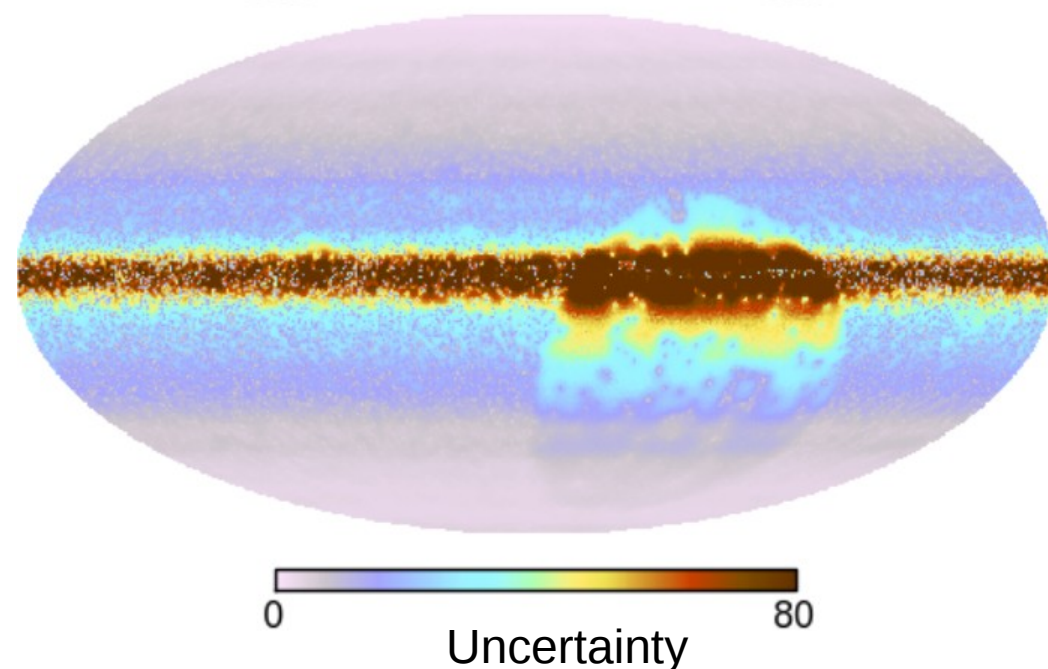
Inferred Galactic Faraday depth



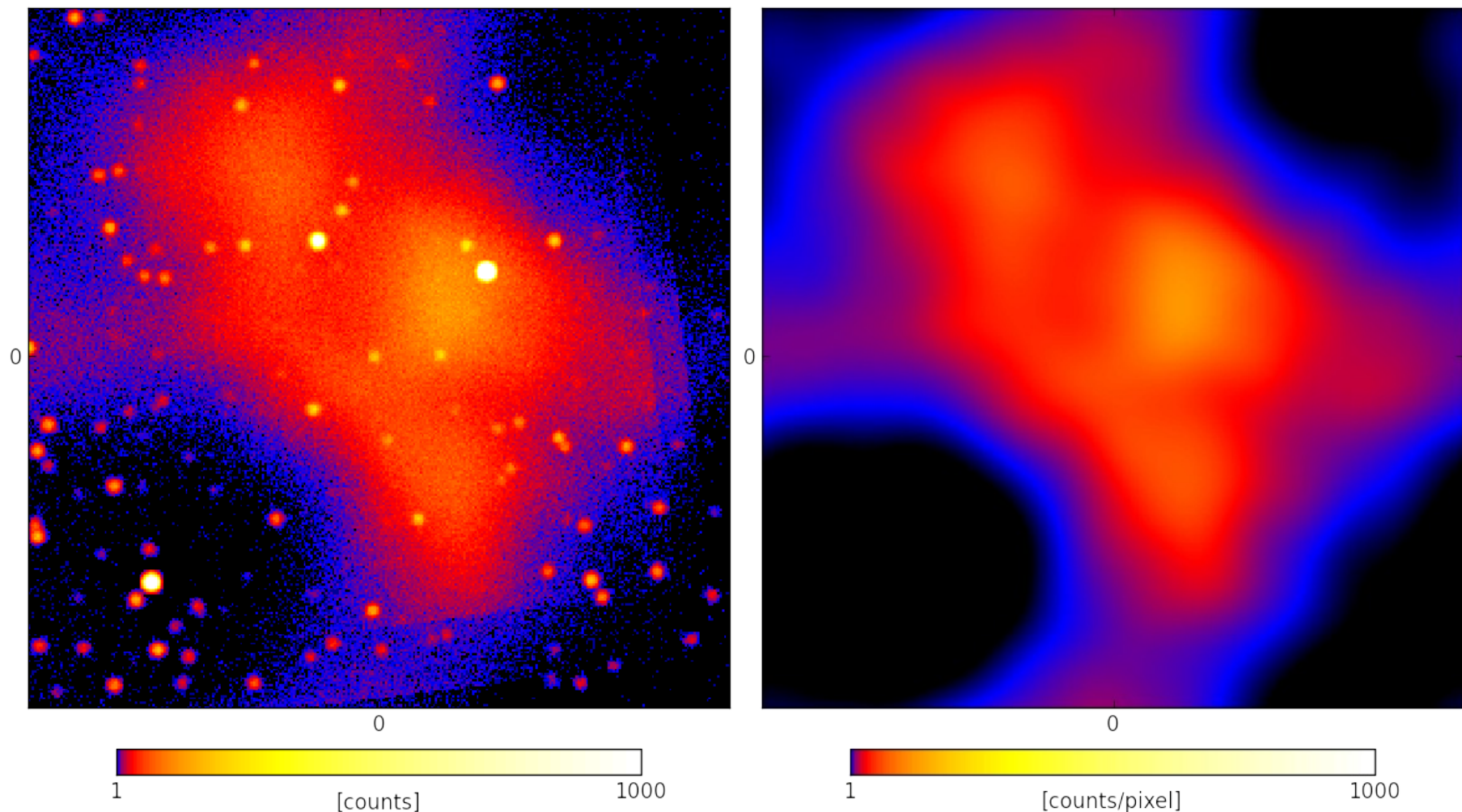
Studying magnetic fields in the Milky Way through inference of “Faraday depth.”

Oppermann, et al. (2012)
<http://arxiv.org/abs/1111.6186>

Oppermann, et al. (2014)
<http://arxiv.org/abs/1404.3701>

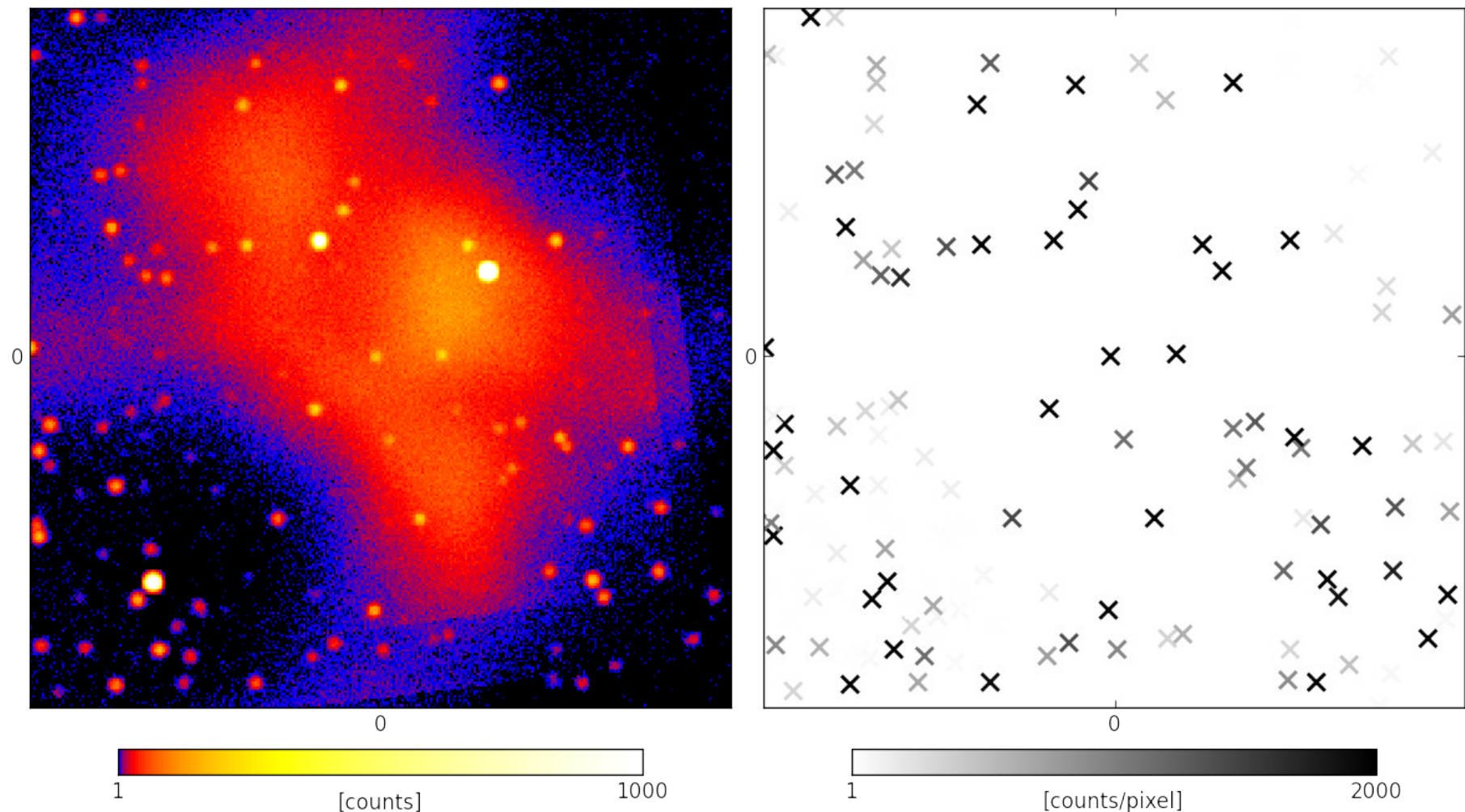


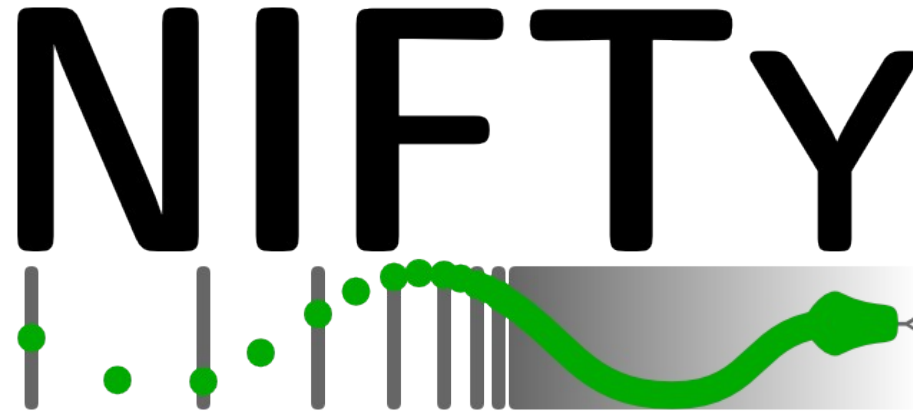
Denoising, Deconvolving, and Decomposing Photon Observations – **D³PO** (Selig, et al., in prep.)



Denoising, Deconvolving, and Decomposing Photon Observations – **D³PO**

(Selig, et al., in prep.)





NIFTy project homepage: <http://www.mpa-garching.mpg.de/ift/nifty/>

and on GitHub: <https://github.com/mselig/nifty>

bellmr@gmail.com
@mryanbell

Thank you!