

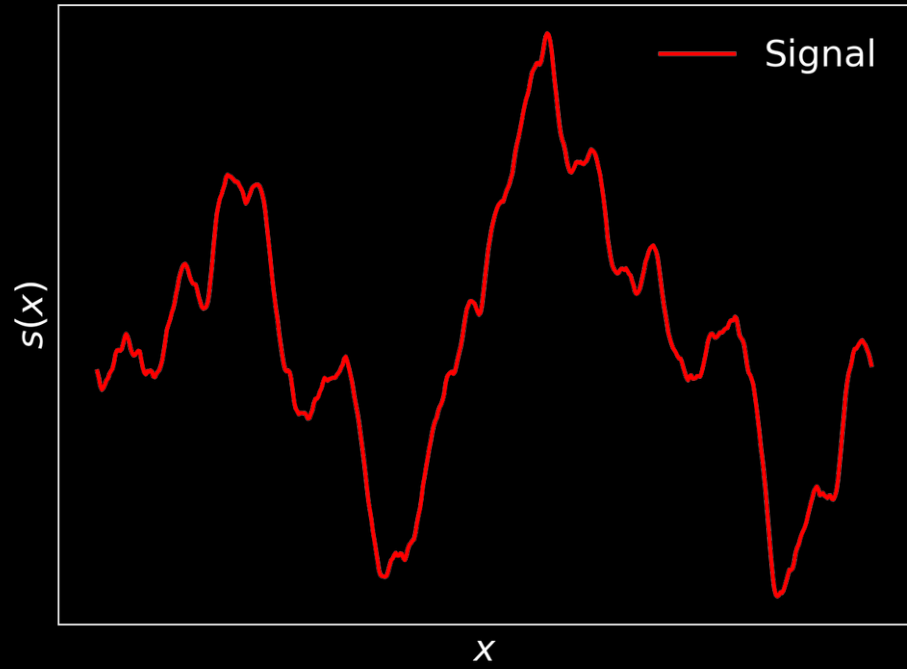
X-UBIK

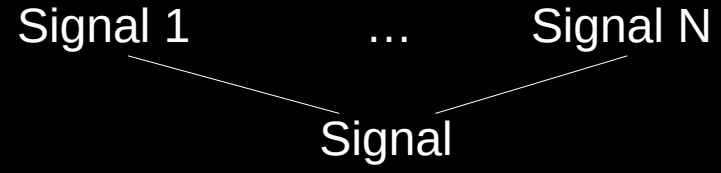
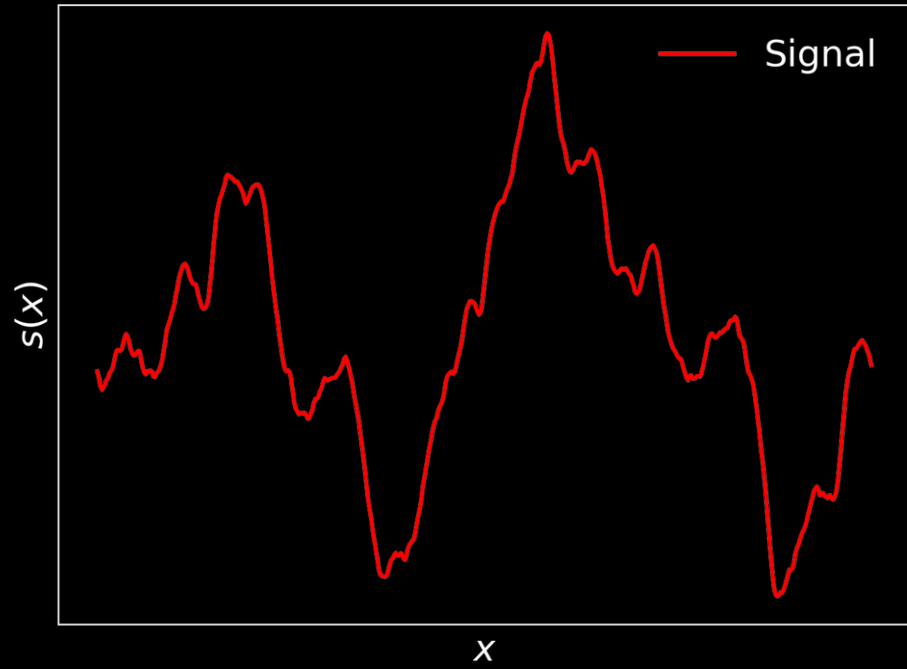
The Universal Bayesian Imaging Kit Applied to X-ray data

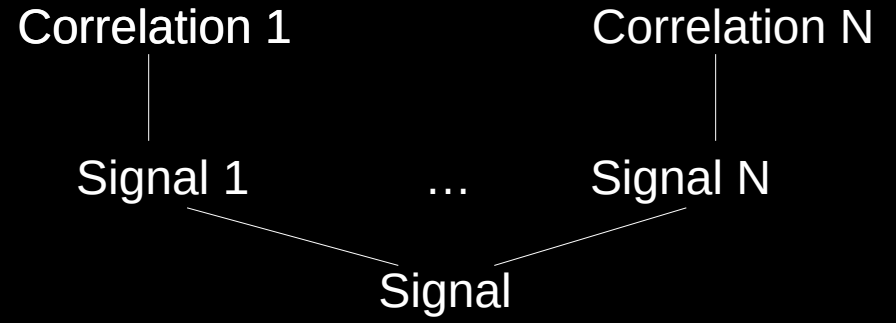
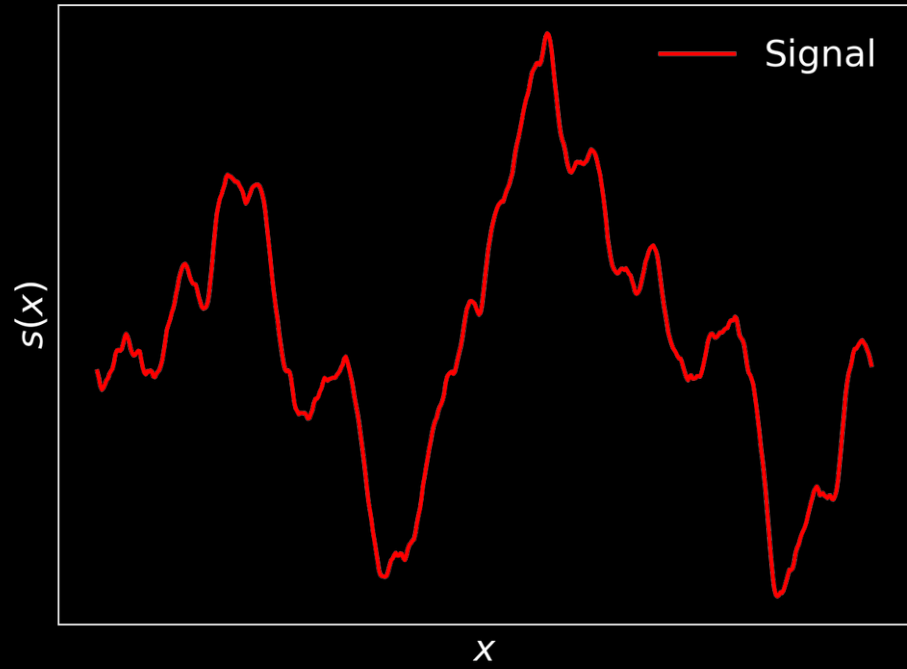
M. Westerkamp, V. Eberle, M. Guardiani, P. Frank, T. Enßlin

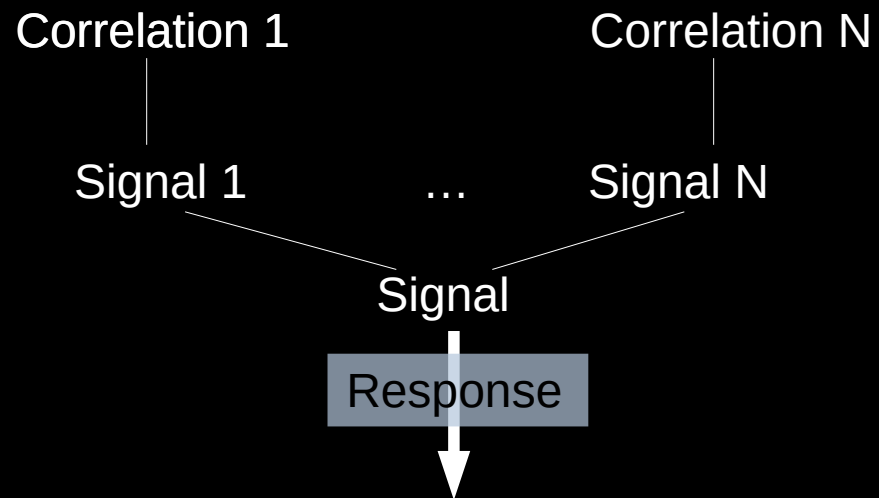
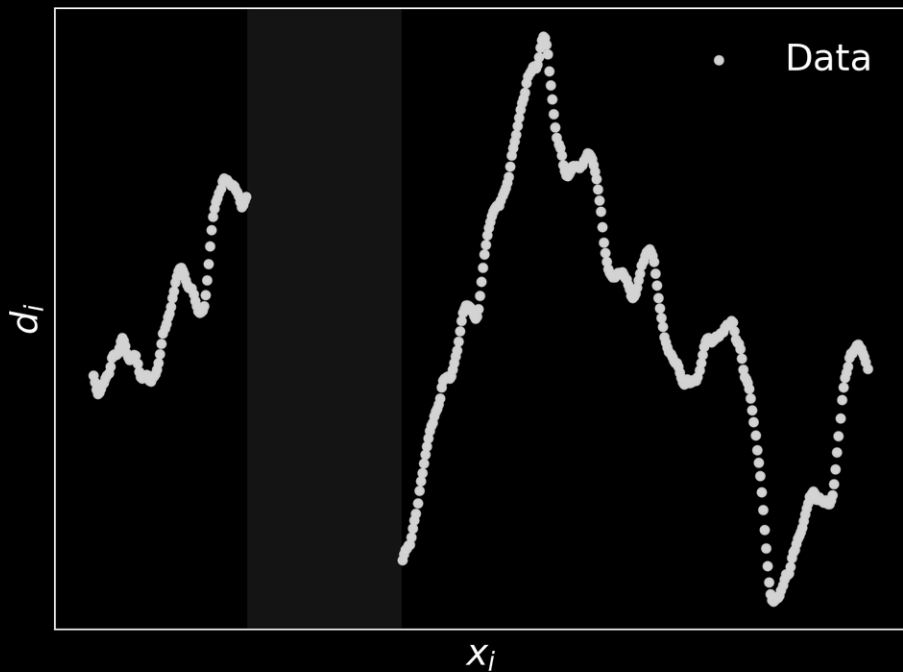
Sep. 19, 2024 Garching
First Results from the SRG/eROSITA All-Sky Survey
From Stars to Cosmology

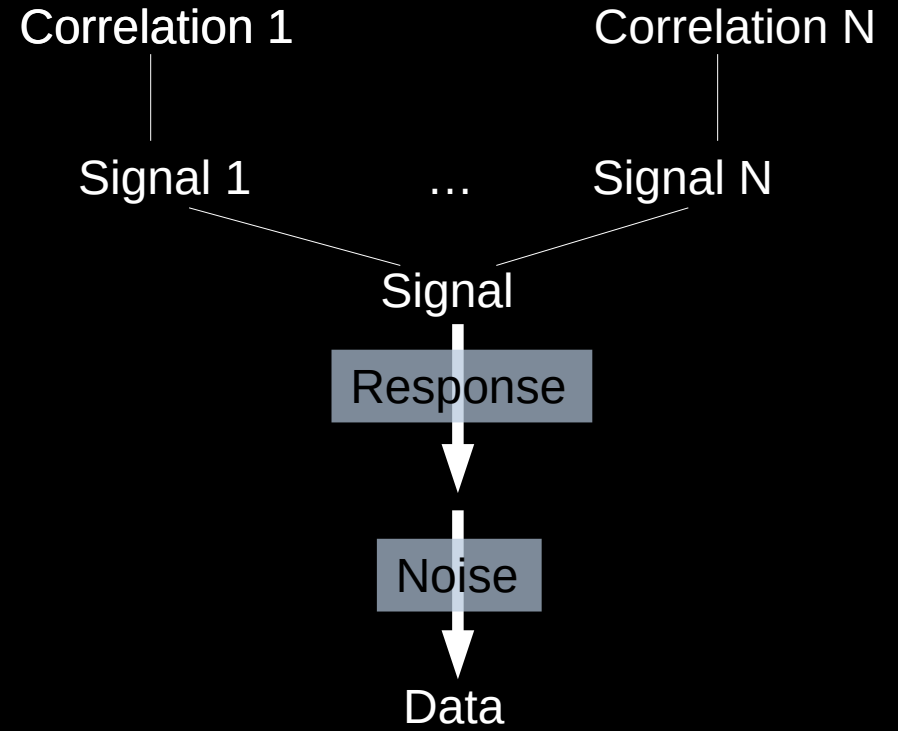
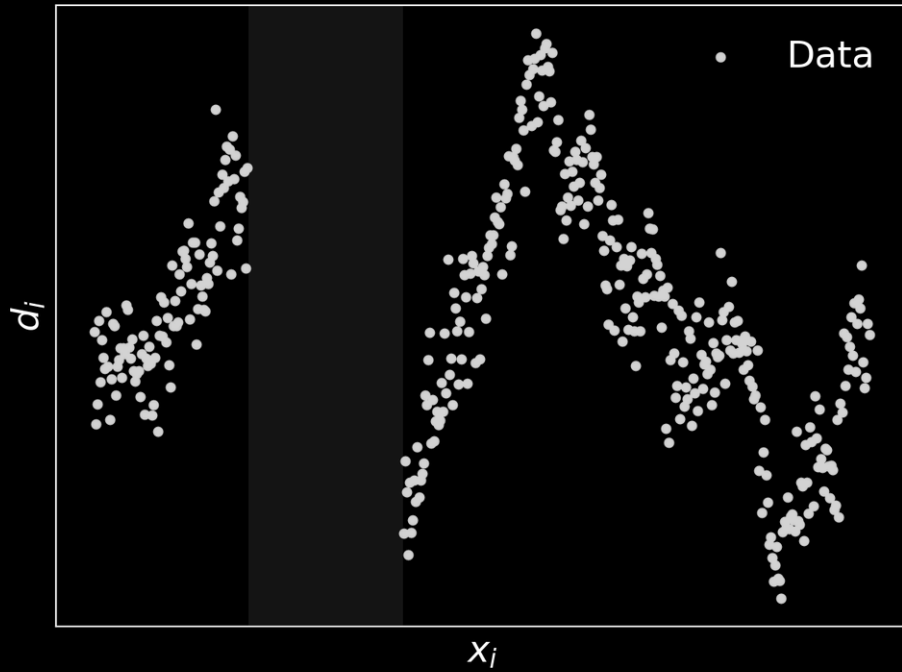


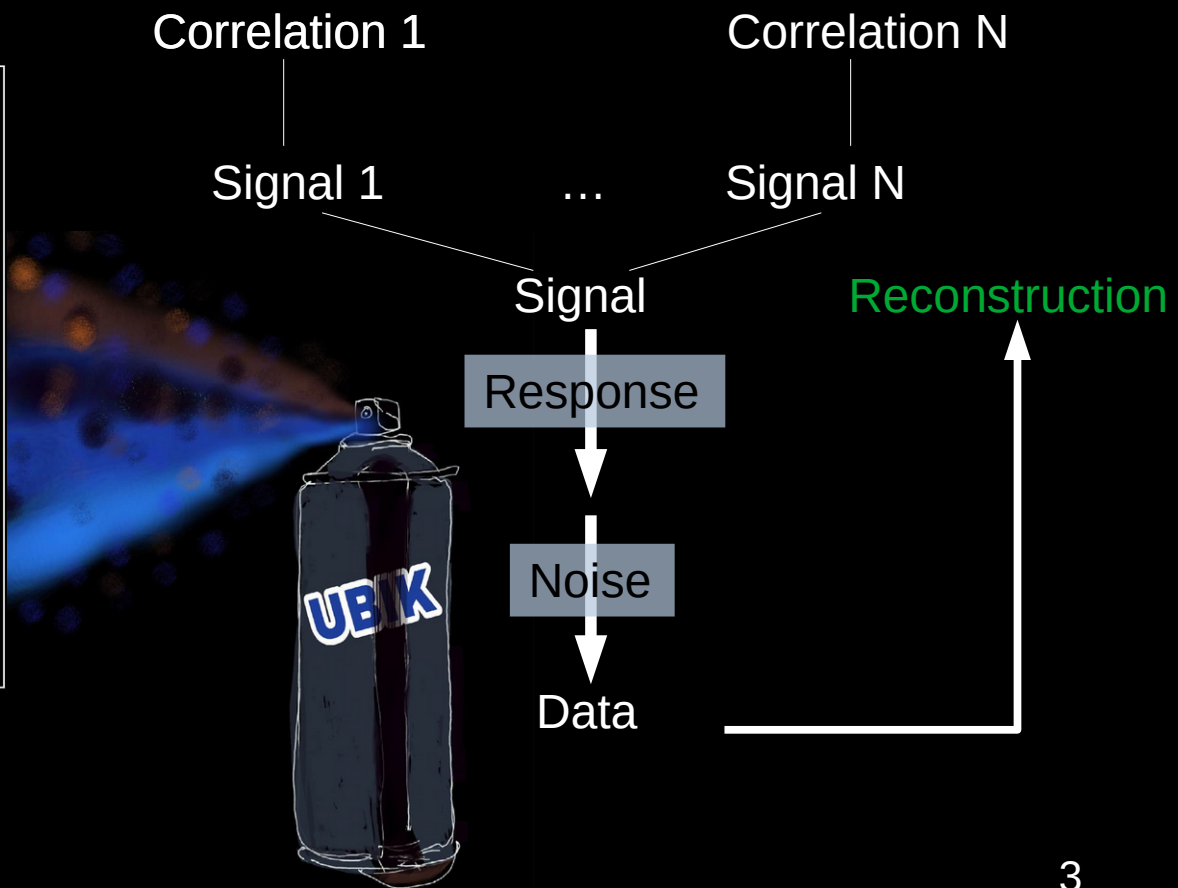
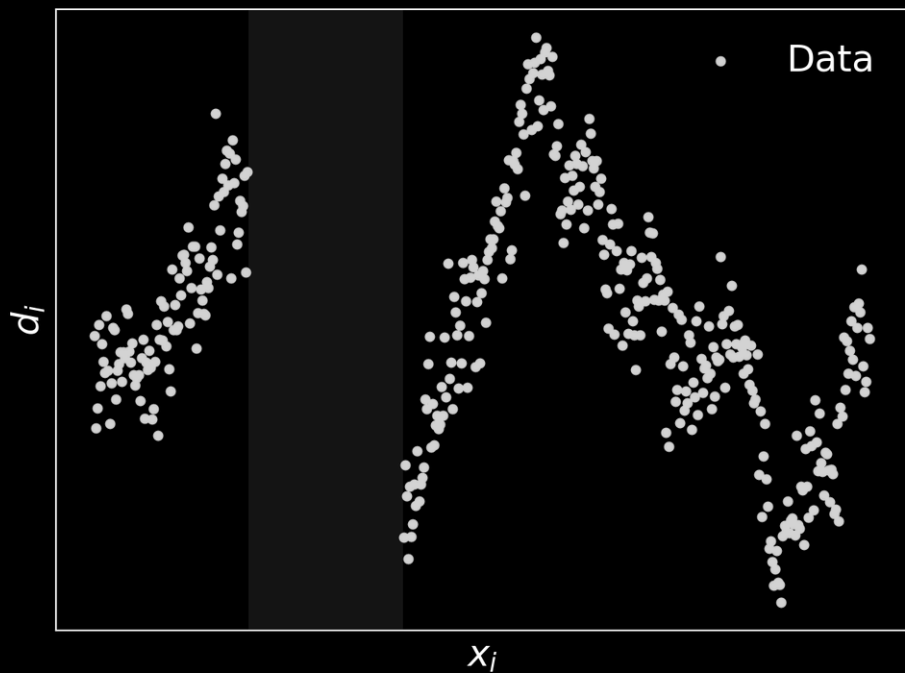


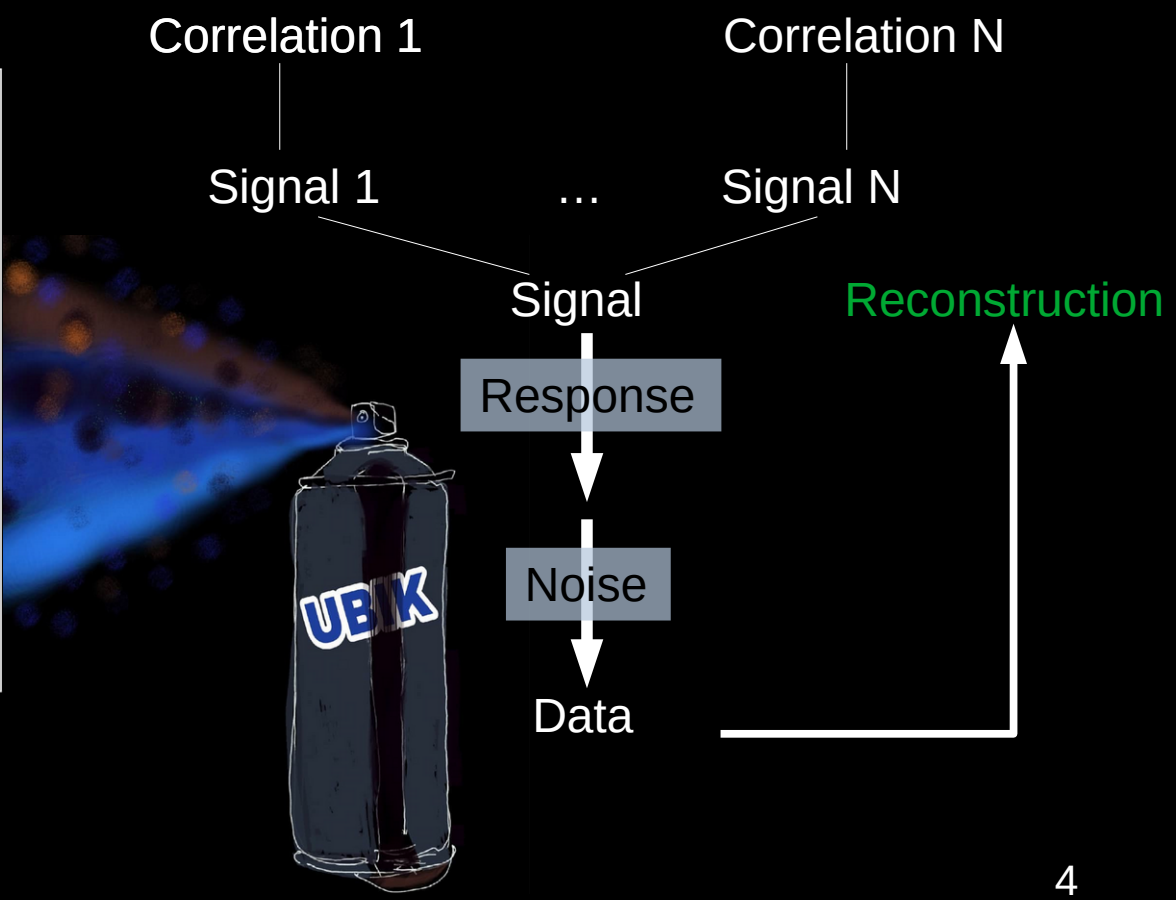
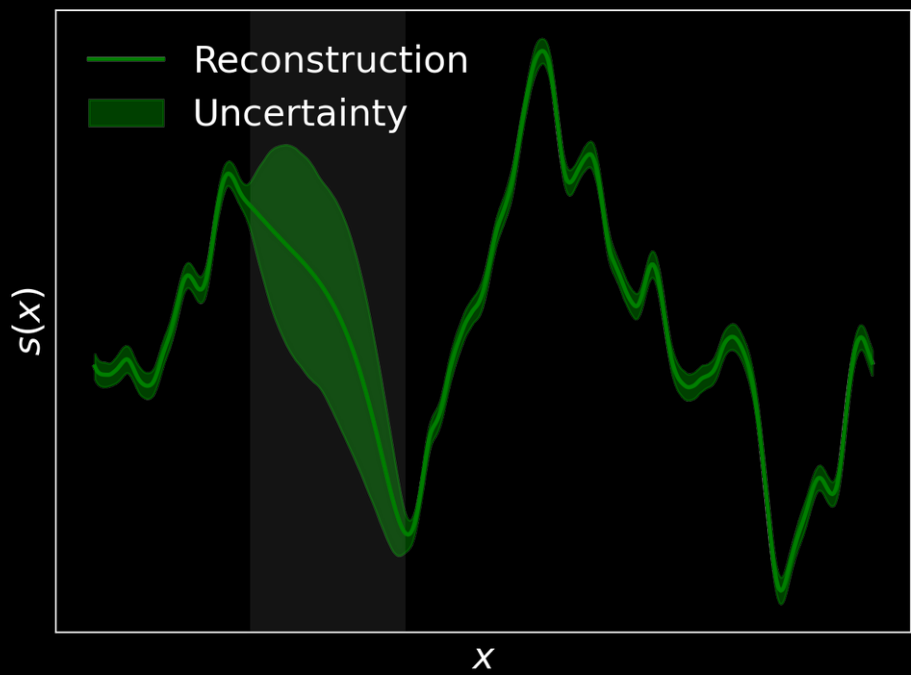












Challenges

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- Non-trivial correlation structures
- Mixture of several components

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- Response representation & application
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Information Field Theory

$$P(s|d) = \frac{P(d|s)P(s)}{P(d)}$$

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Variational inference:

$$\min D_{KL}(Q(s|d)|P(s|d))$$

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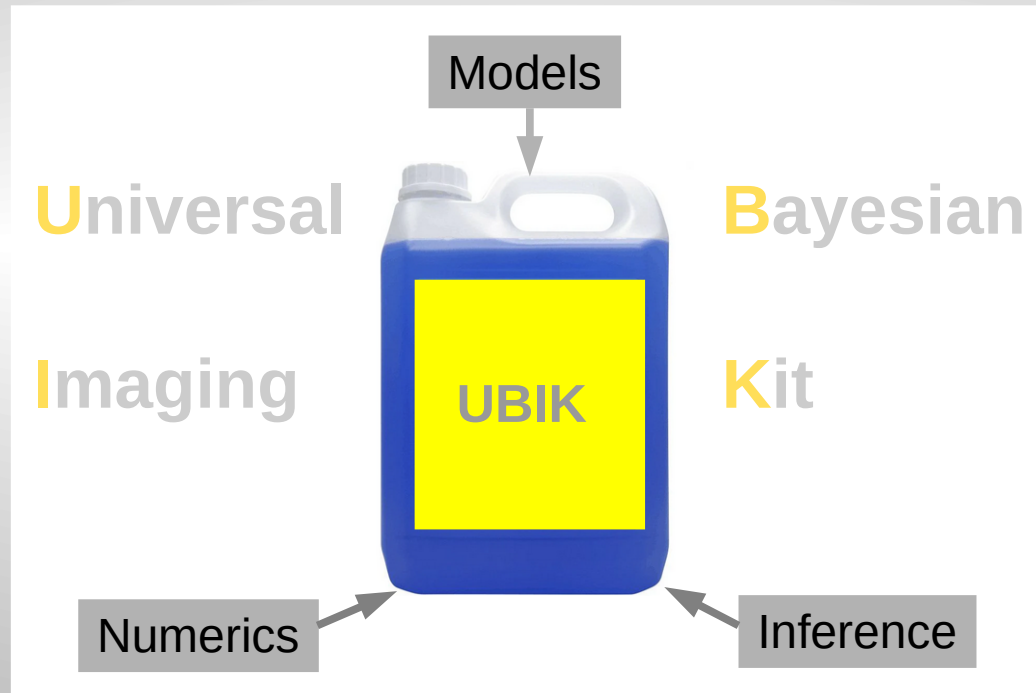
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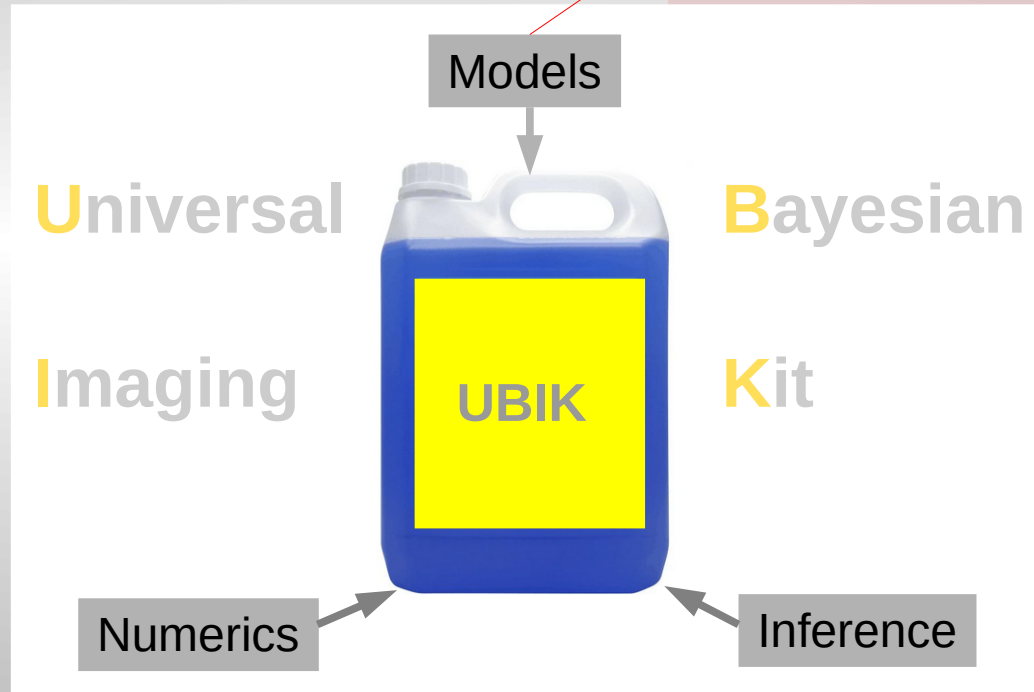
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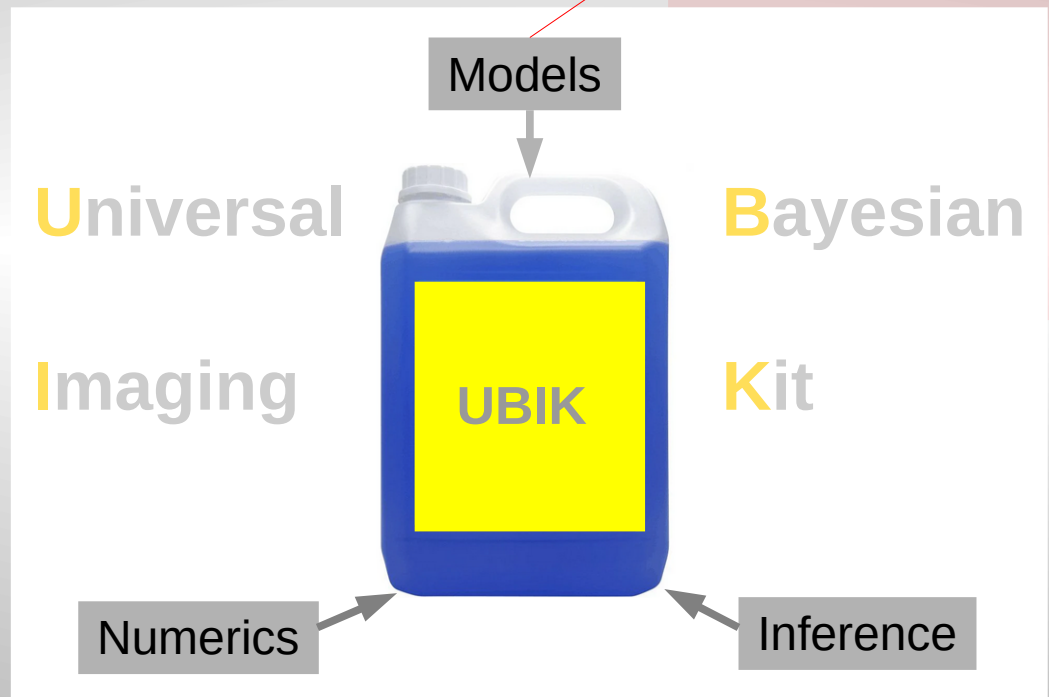
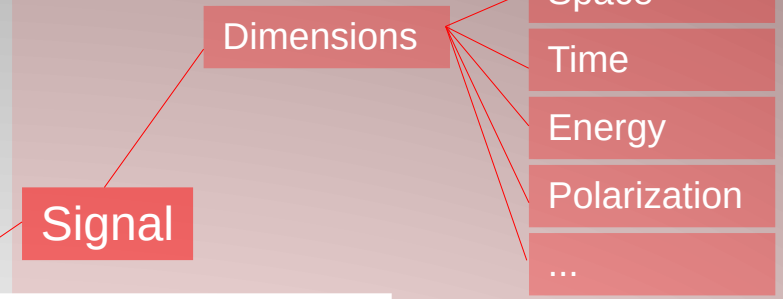
$$m = \langle s \rangle_Q$$
$$\sigma^2 = \langle (s - m)^2 \rangle_Q$$

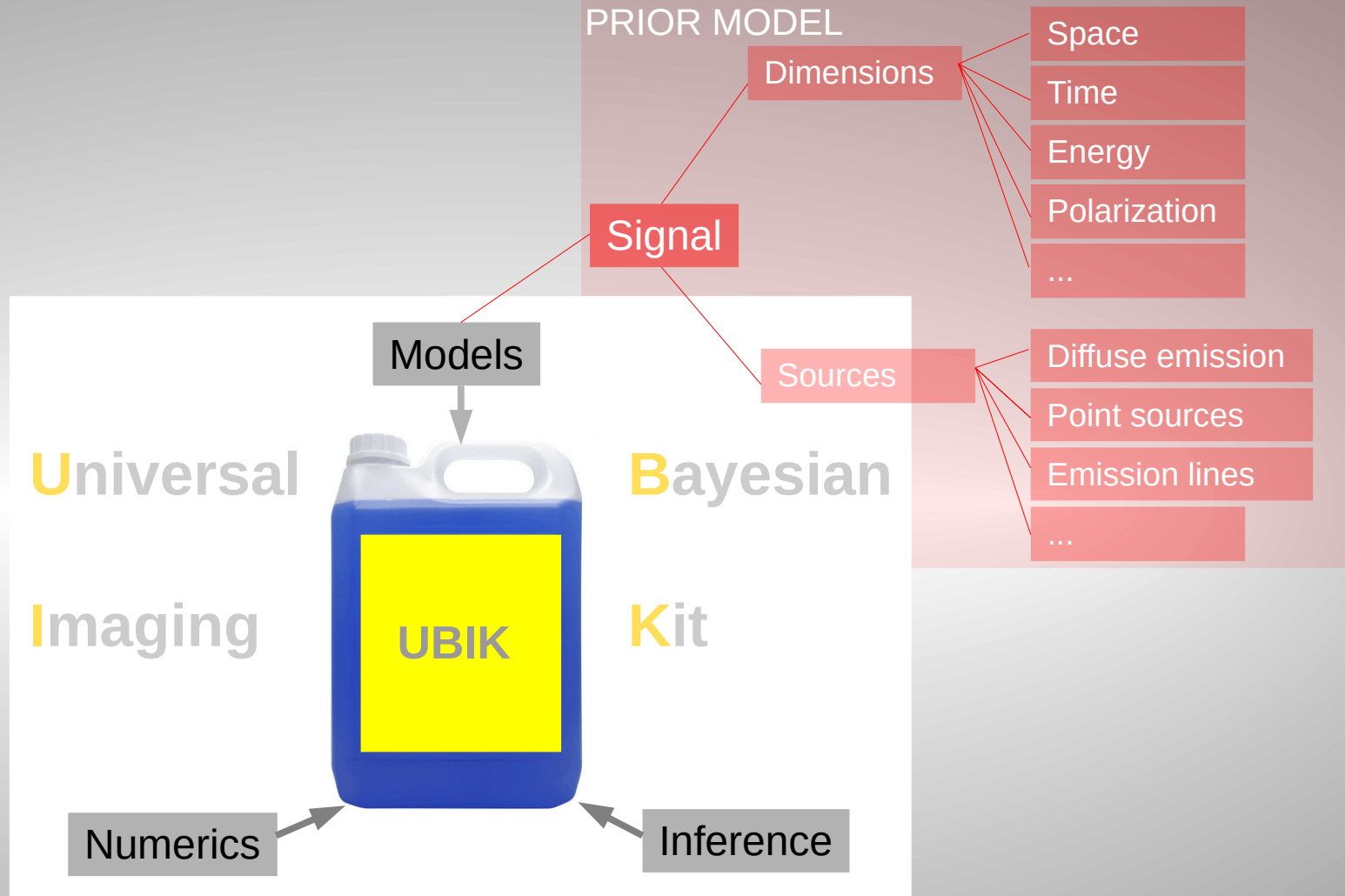


Signal



PRIOR MODEL





LIKELIHOOD MODEL

PRIOR MODEL

- Gauss
- Poisson
- ...

Noise

Instrument

Models

Signal

Dimensions

Sources

- Space
- Time
- Energy
- Polarization
- ...

- Diffuse emission
- Point sources
- Emission lines
- ...

Universal

Bayesian

Imaging

Kit



Numerics

Inference

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MAP

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geoVI

MGVI

HMC

...

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NIFTY



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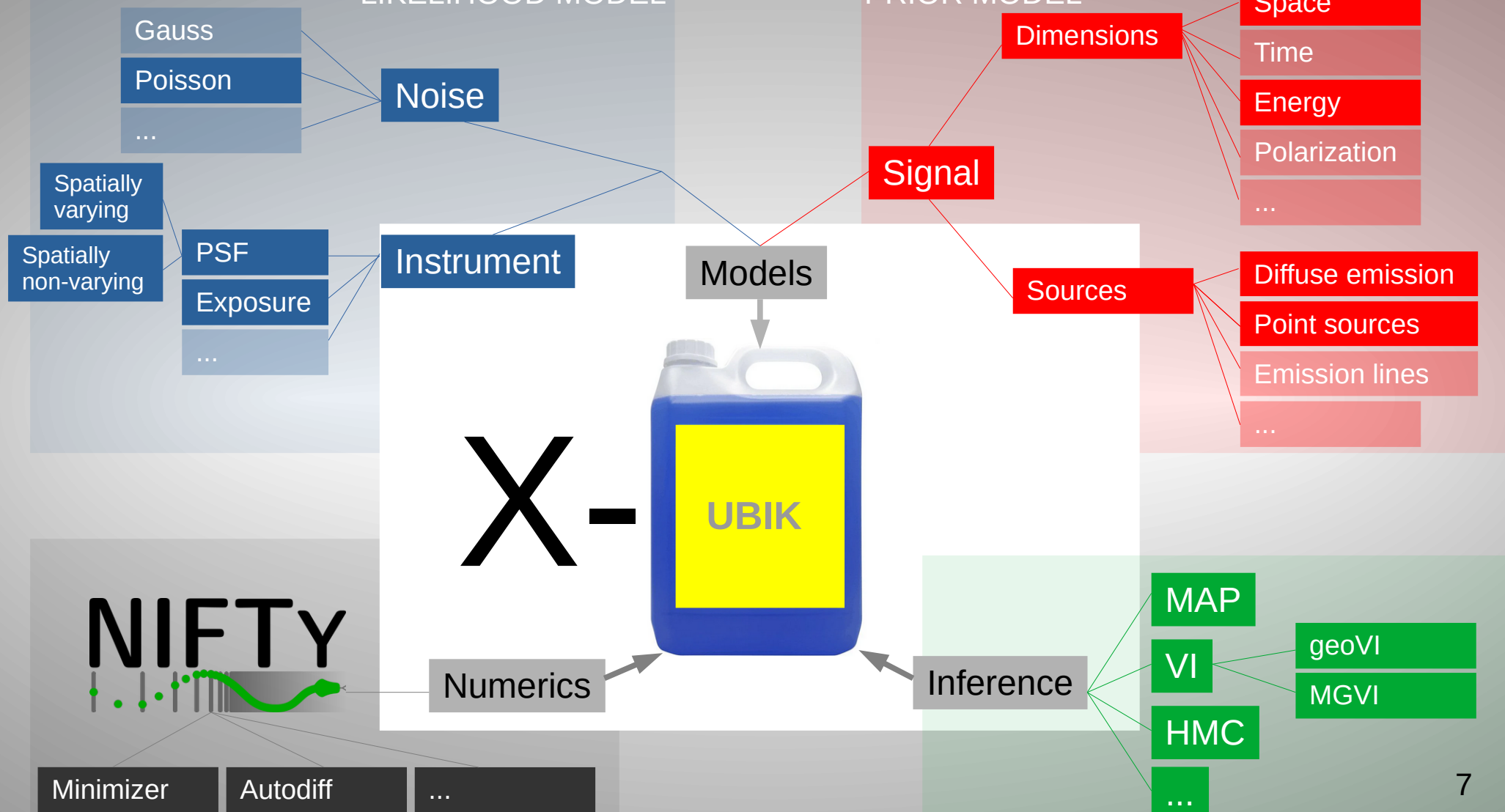
Minimizer

Autodiff

...

LIKELIHOOD MODEL

PRIOR MODEL



Likelihood Model

Poissonian log-likelihood for observation k:

$$-\ln P(d_k | \lambda_k) = -\sum_{i=1}^N [\lambda_k^i - d_k^i \ln \lambda_k^i + \ln(d_k^i!)]$$

Likelihood Model

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count data d_k

Likelihood Model

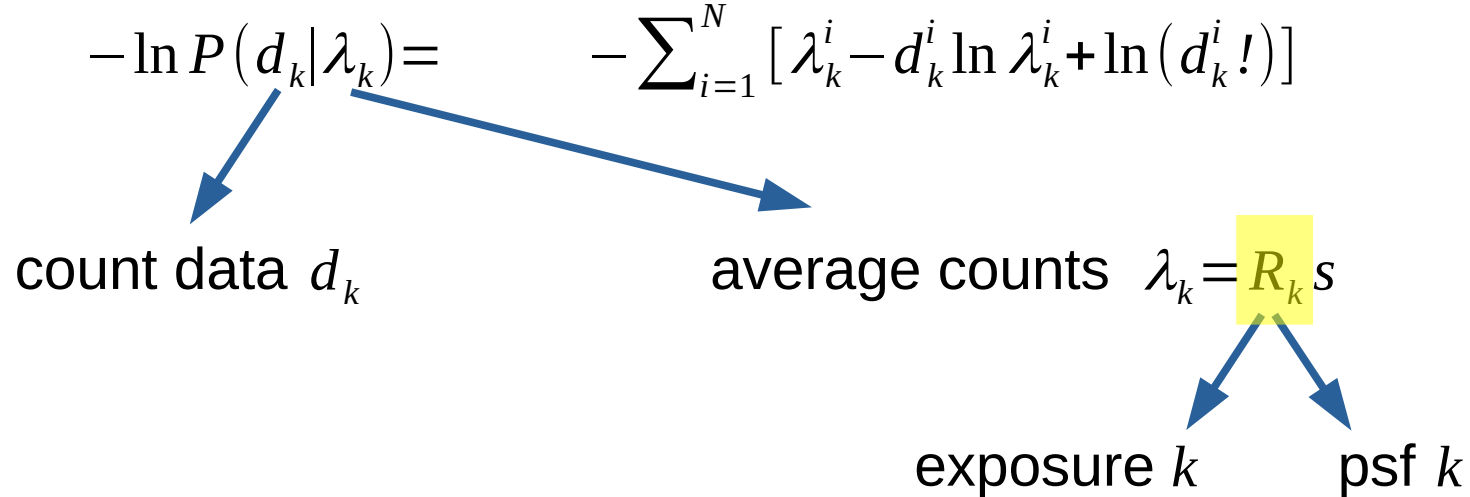
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Likelihood Model

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Likelihood Model

Poissonian log-likelihood:

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exposure k

psf k

Prior Generative Models

$$s = s(\xi), \quad P(\xi) = \mathcal{N}(\xi, 1)$$

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diffuse emission

point sources

$$s(\xi) = \quad + \quad + \quad \dots$$

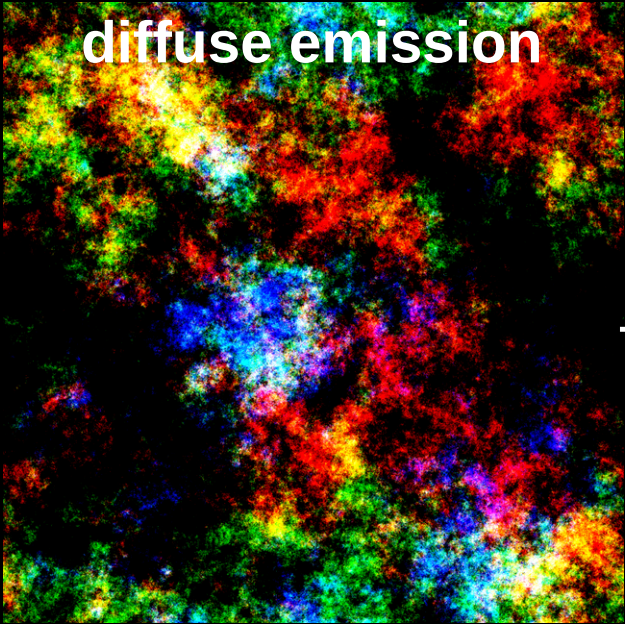
Prior Generative Models

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diffuse emission

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$$s(\xi) = \text{diffuse emission} + \text{point sources} + \dots$$



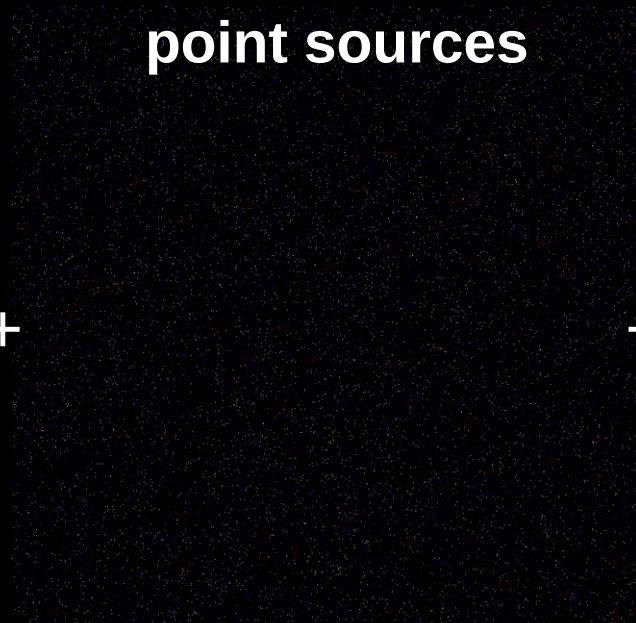
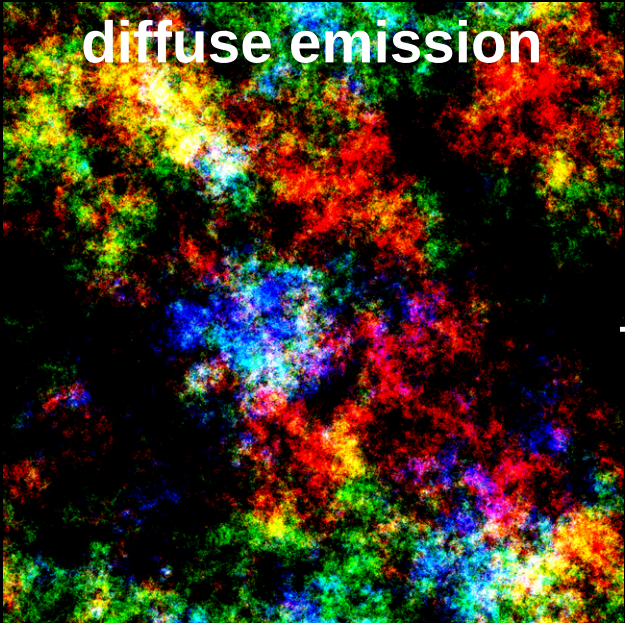
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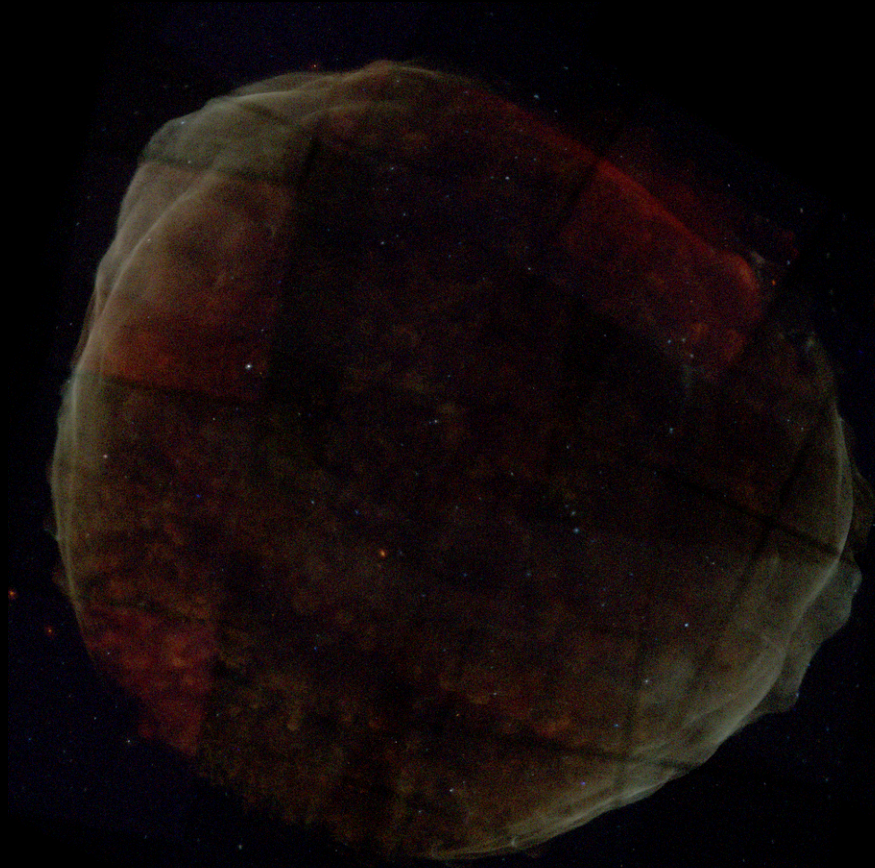
diffuse emission

point sources

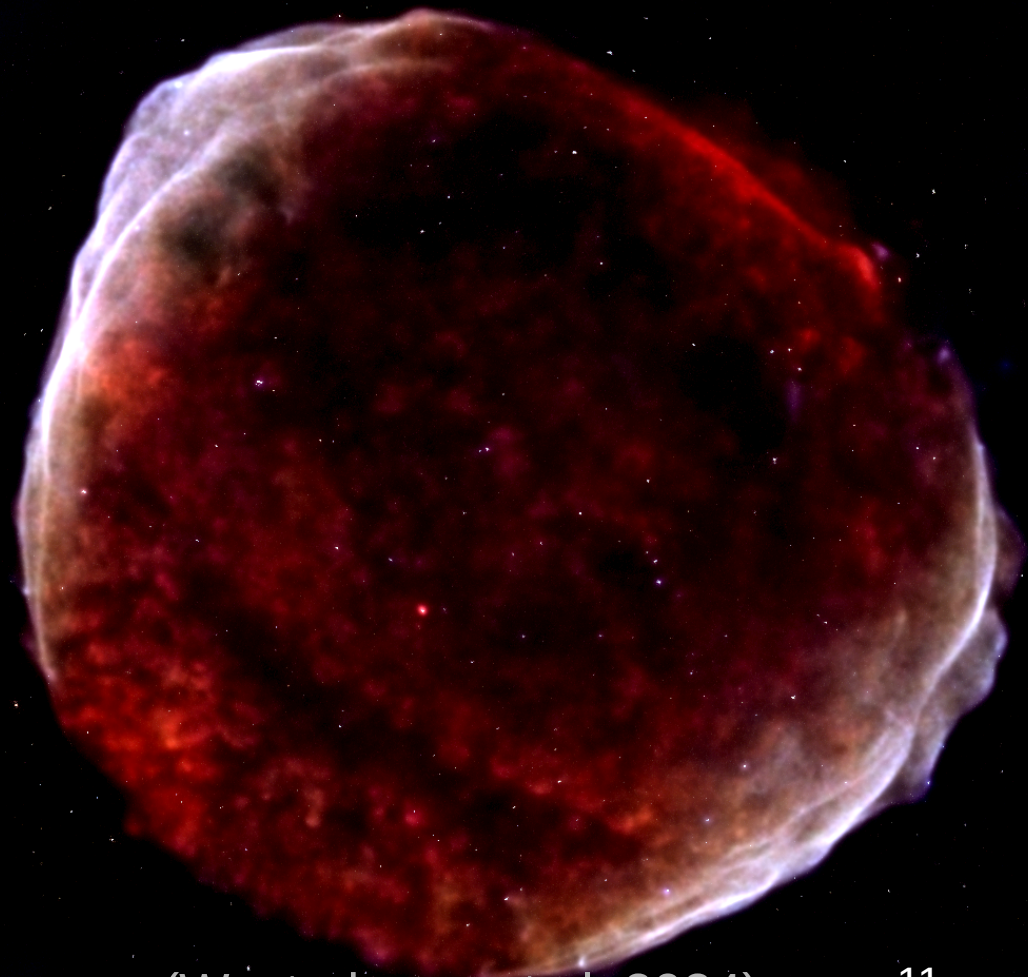
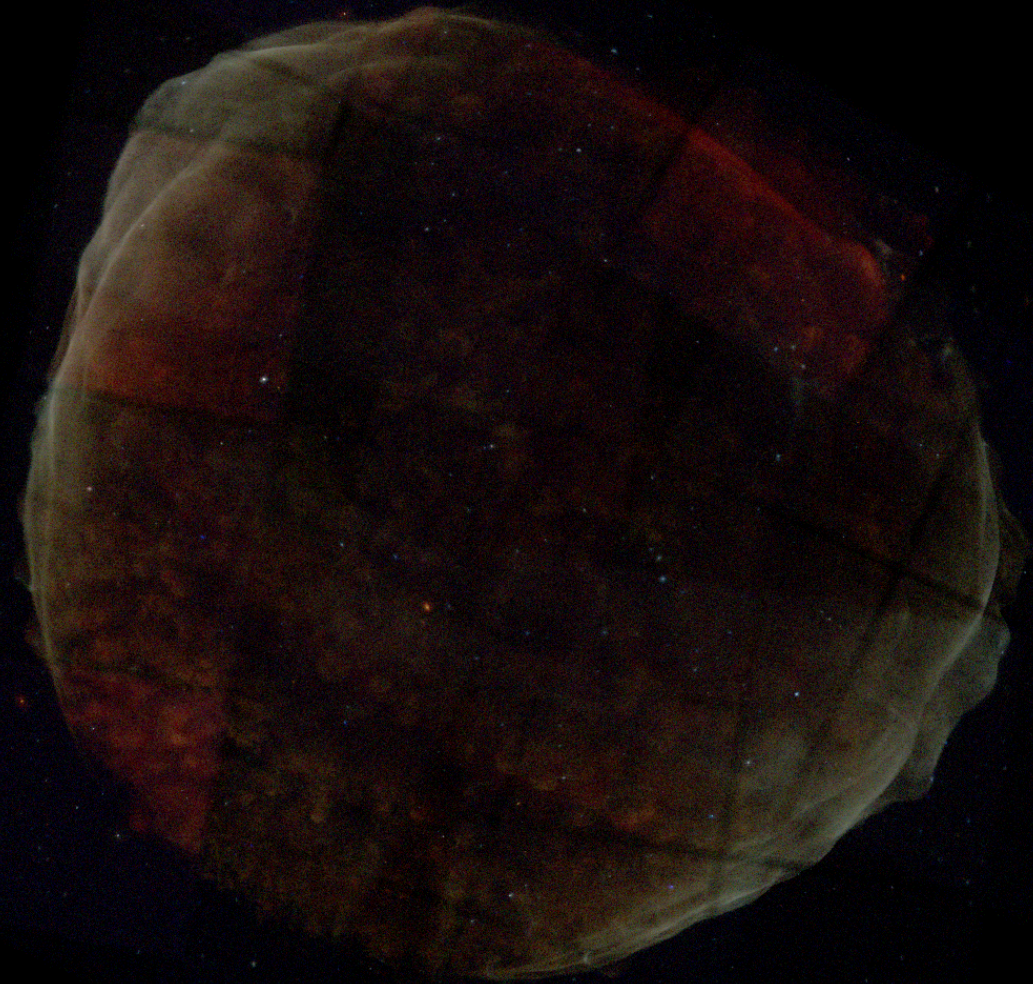
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Chandra - SN1006

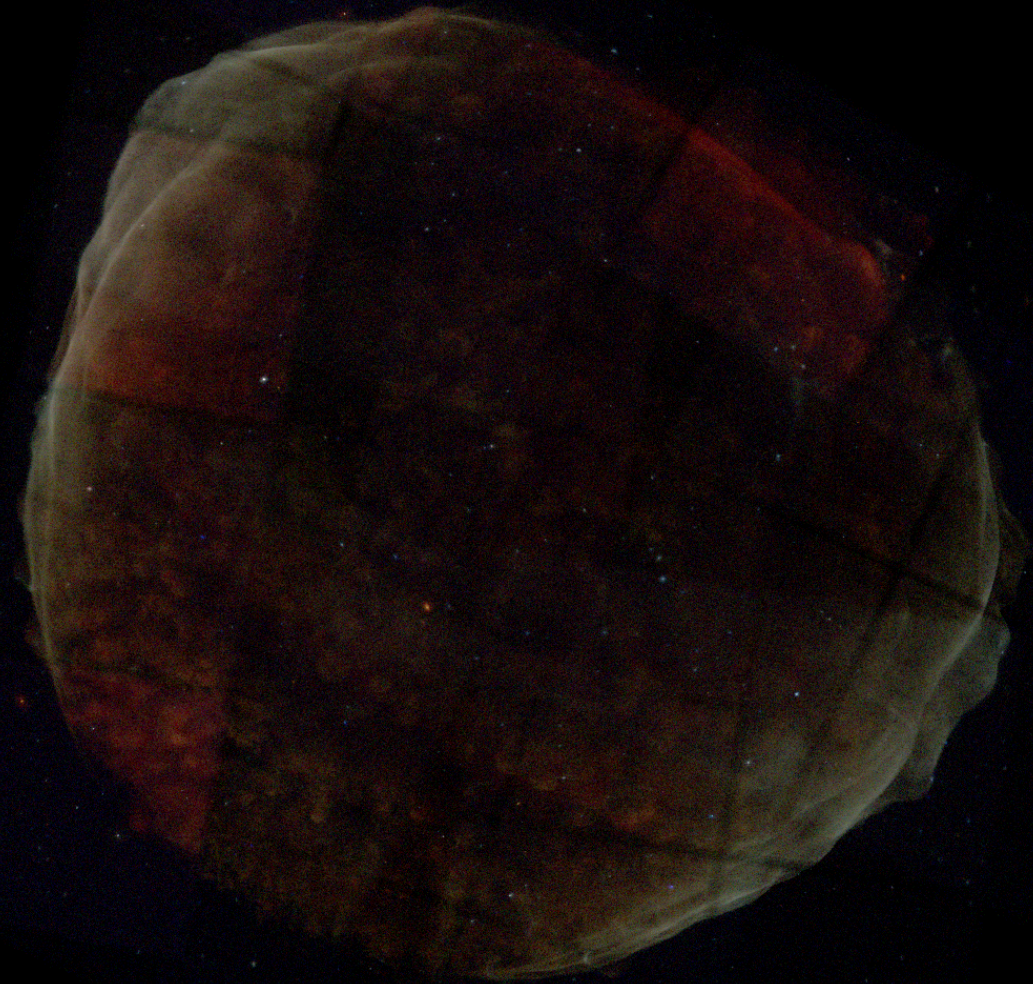


Reconstructed Sky



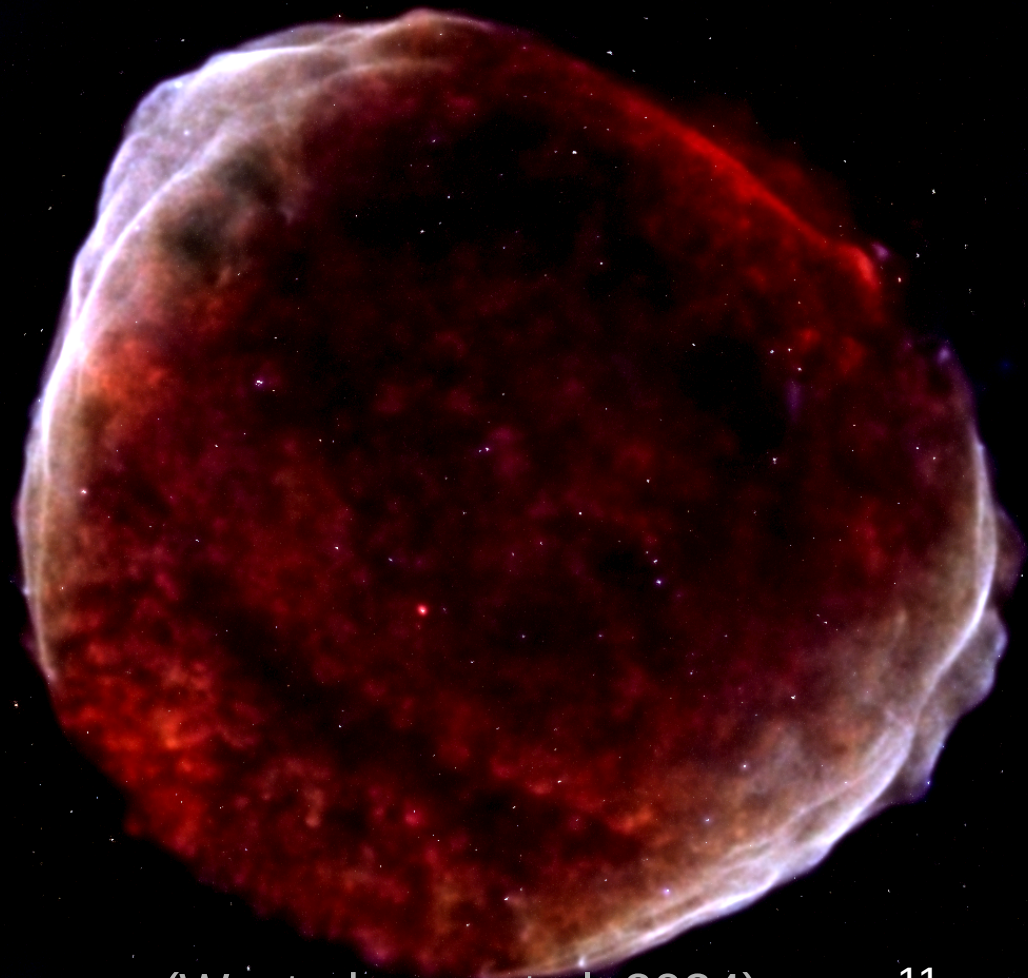
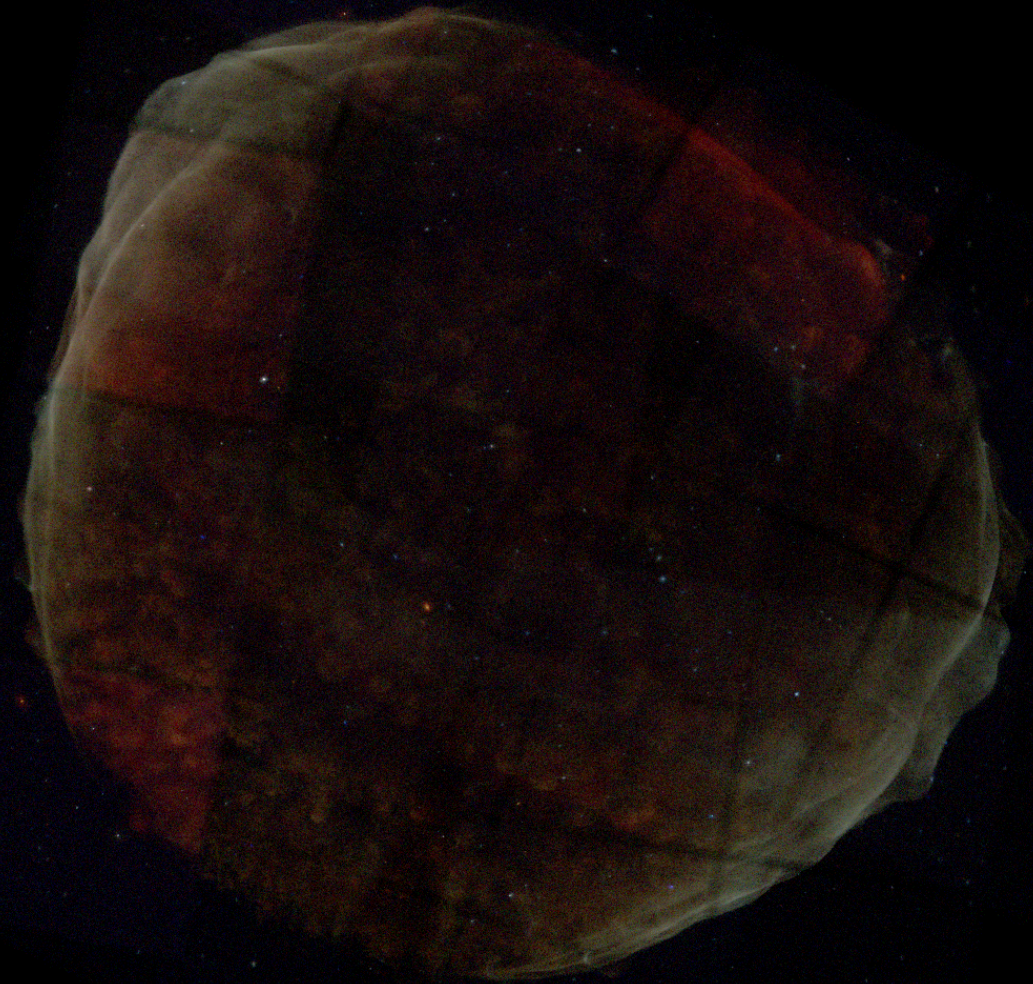
(Westerkamp et al. 2024)

Reconstructed Point Sources



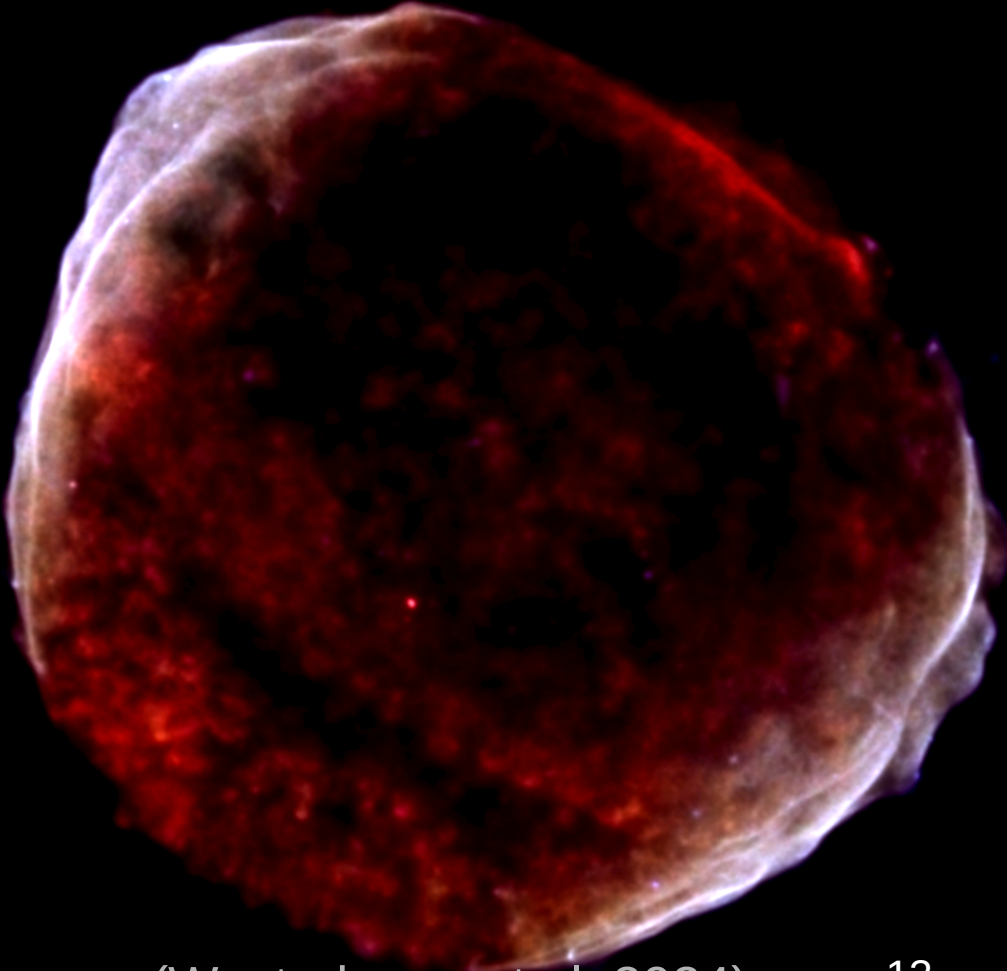
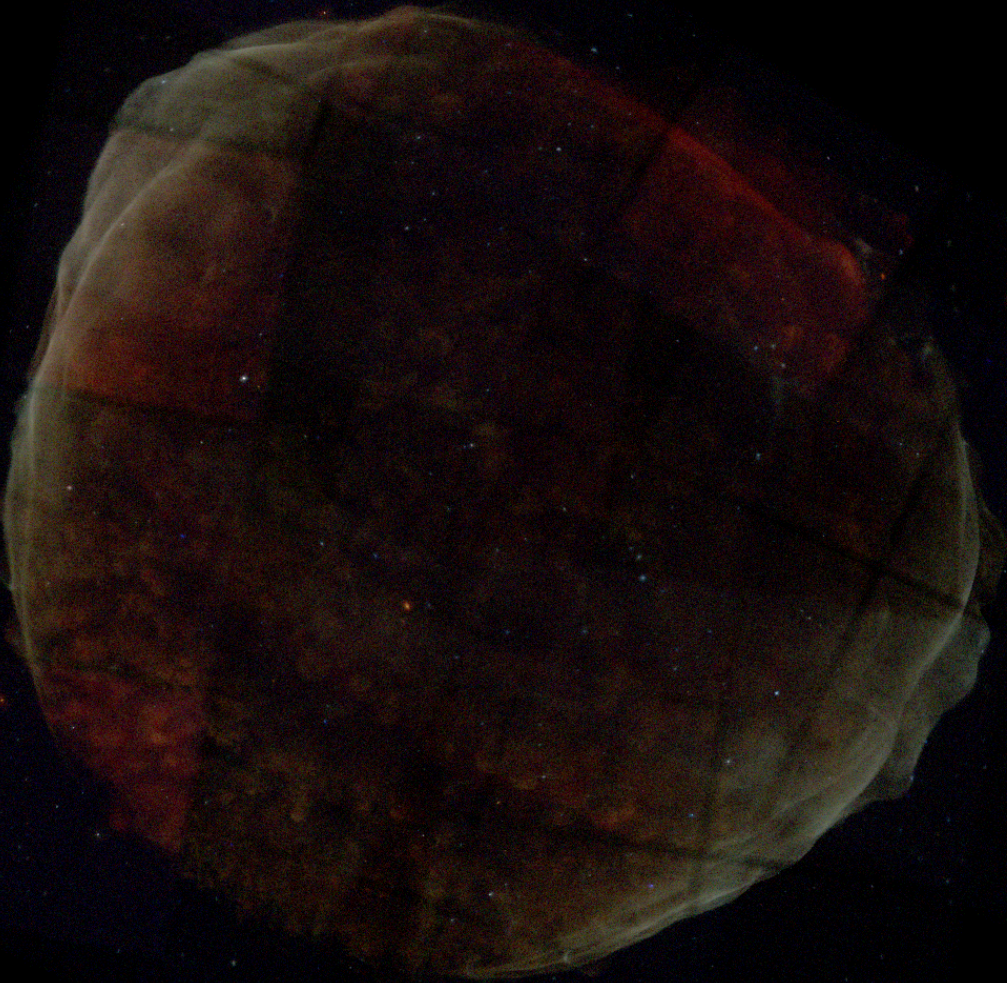
(Westerkamp et al. 2024)

Reconstructed Sky



(Westerkamp et al. 2024)

Reconstructed Diffuse Emission



(Westerkamp et al. 2024)

... and more

eROSITA:



Eberle, Guardiani, Westerkamp
et. al in prep.

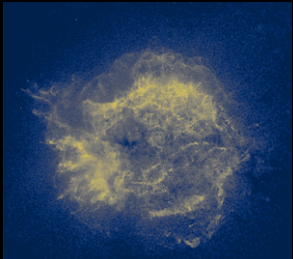
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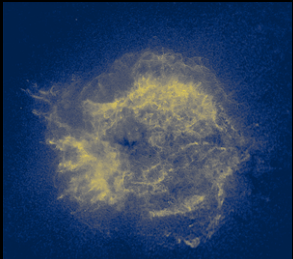
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XMM-Newton:



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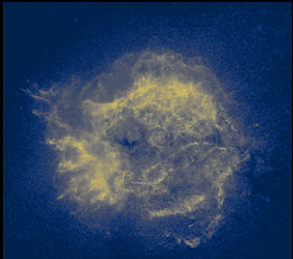
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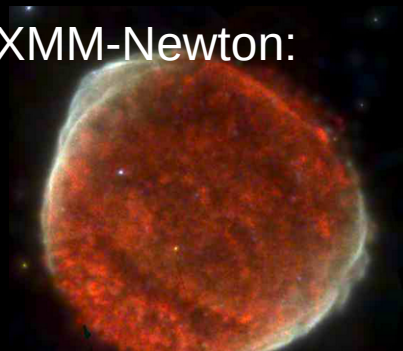
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Rüstig et. al 2023

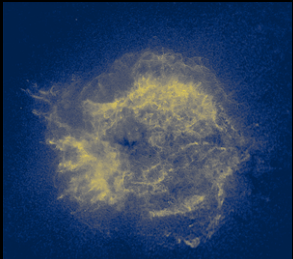
... and more

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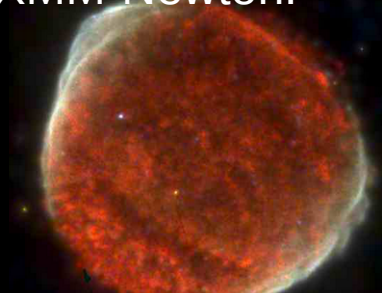
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Eberle et. al in prep.

XMM-Newton:



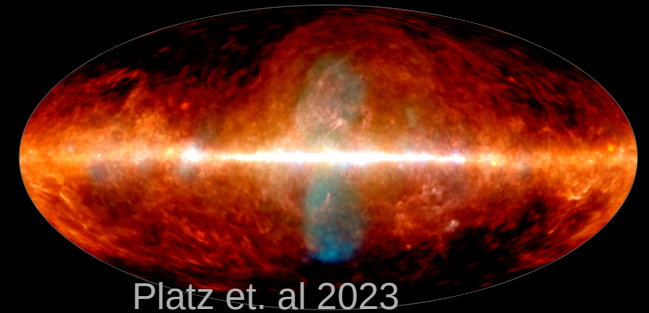
Credit: ESA/ XMM-Newton

JWST:



Rüstig et. al 2023

Fermi:



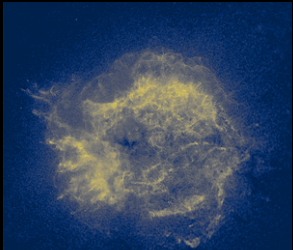
Platz et. al 2023

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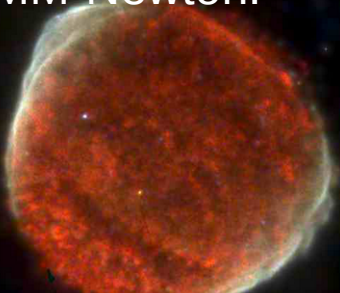
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J-UBIK

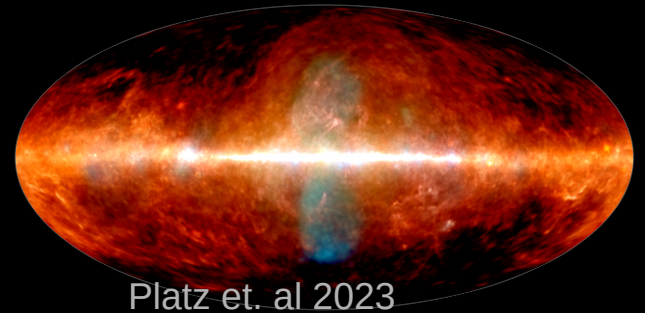
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