



The diffuse Milky Way

—

Sharpening the picture with new inference techniques

Niels Oppermann

with

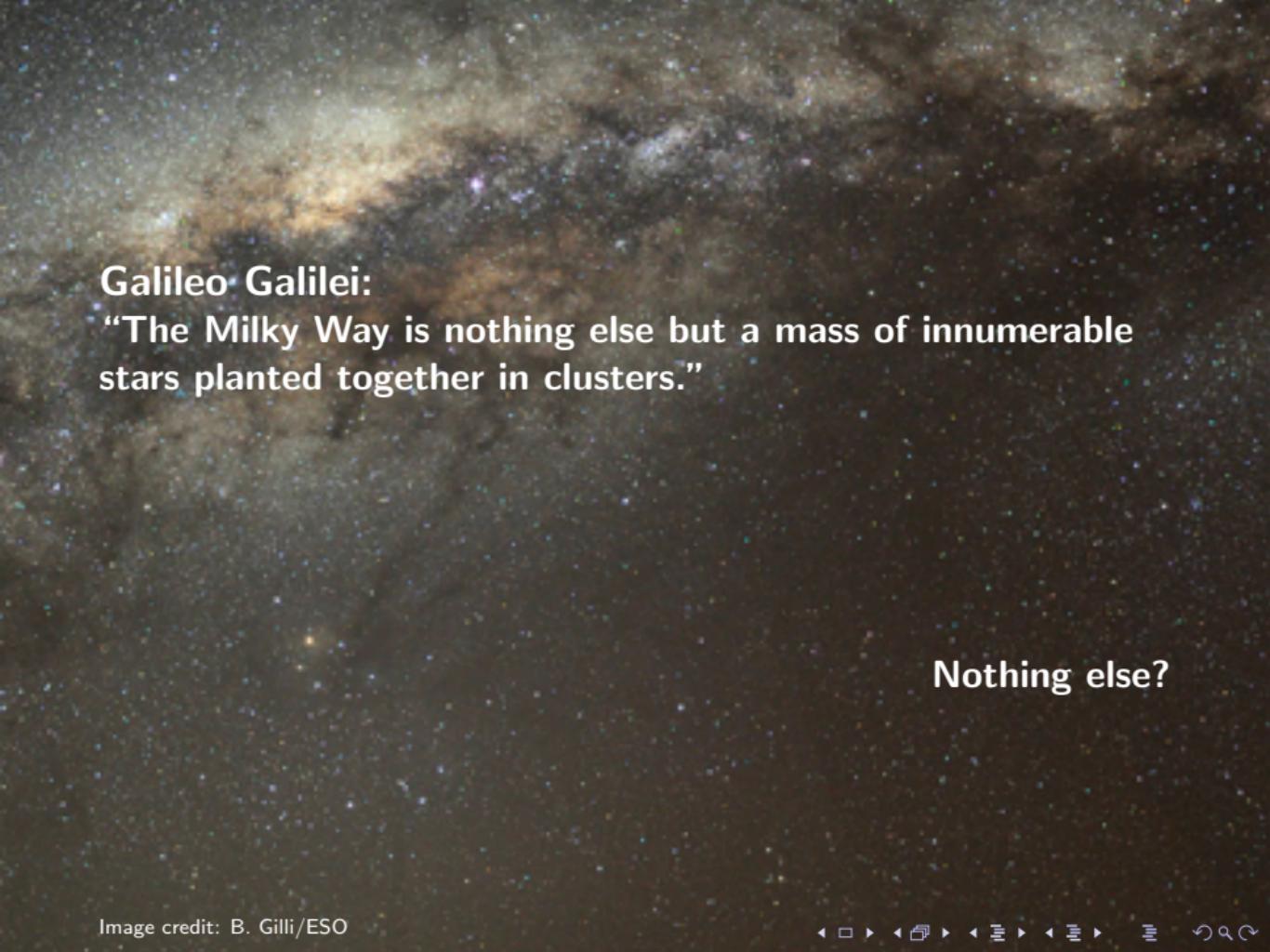
T.A. Enßlin, M.R. Bell, M. Greiner, H. Junklewitz, M. Selig

MPA seminar, Garching, 2013-07-01

The background of the slide is a deep space photograph showing the dense cluster of stars within the Milky Way galaxy. The stars are of various colors, mostly white and blue, with some yellow and orange ones, set against a dark, textured background.

Galileo Galilei:

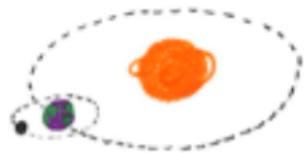
“The Milky Way is nothing else but a mass of innumerable stars planted together in clusters.”

The background of the slide is a dark, textured image of the Milky Way galaxy, showing a dense concentration of stars and interstellar dust.

Galileo Galilei:

“The Milky Way is nothing else but a mass of innumerable stars planted together in clusters.”

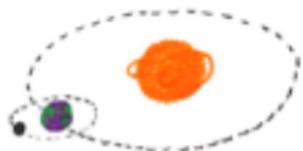
Nothing else?



plasma
 e^-
 e^+
 e^-
 e^-

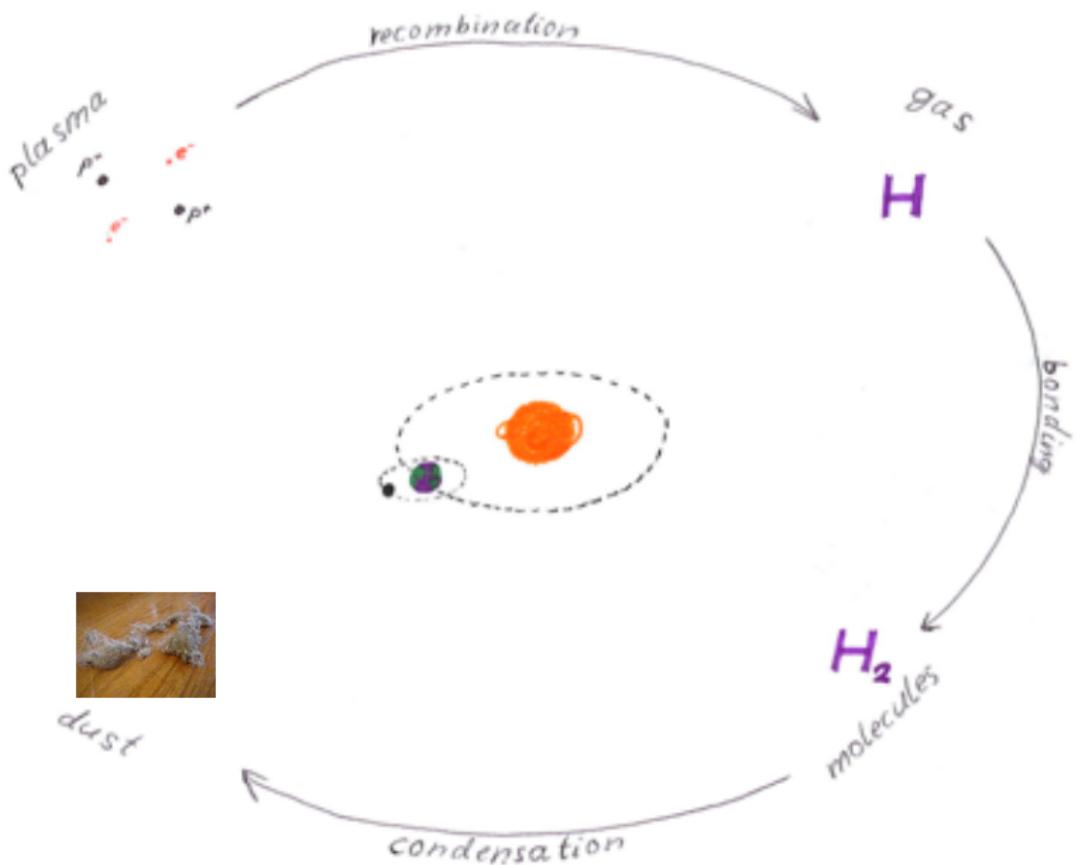
gas

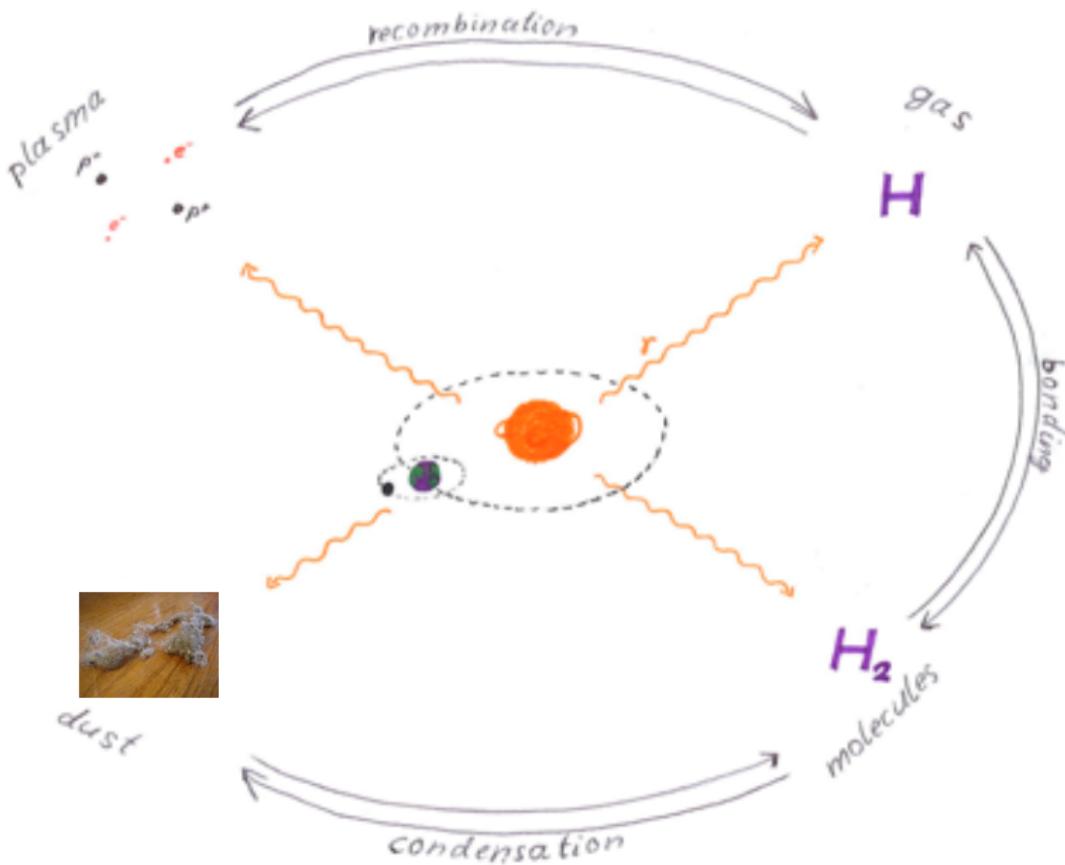
H

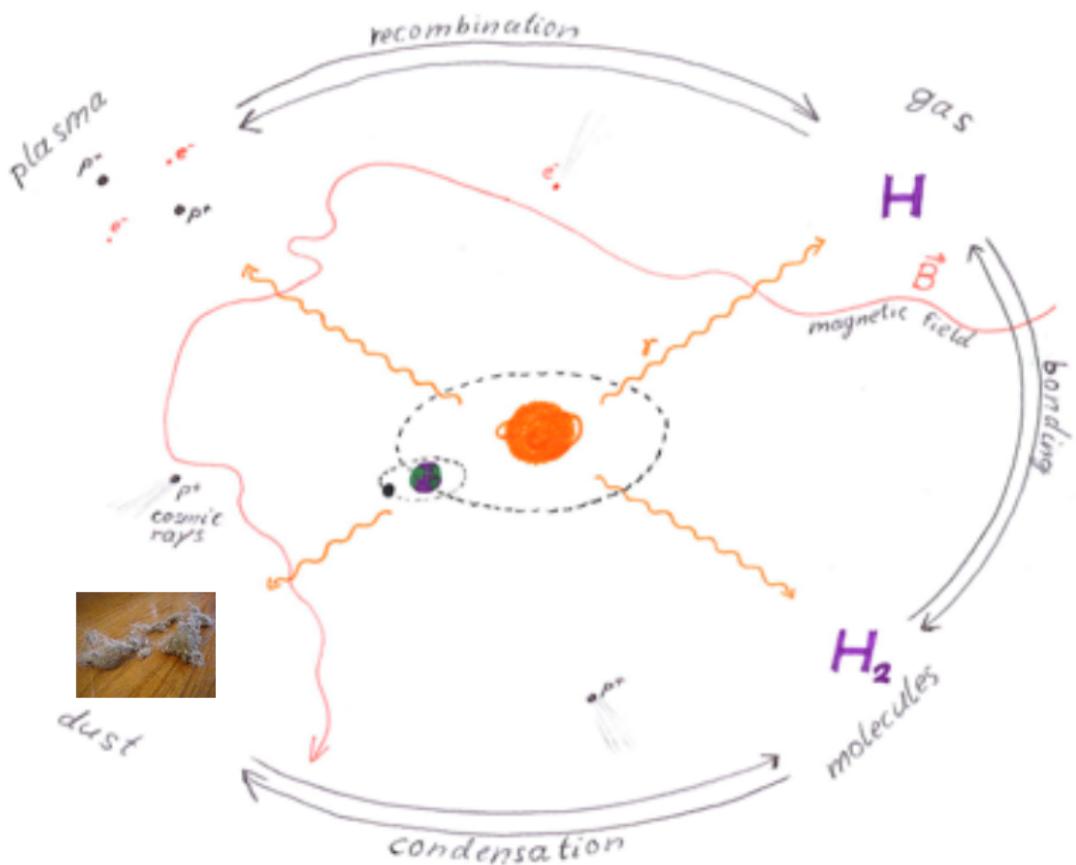


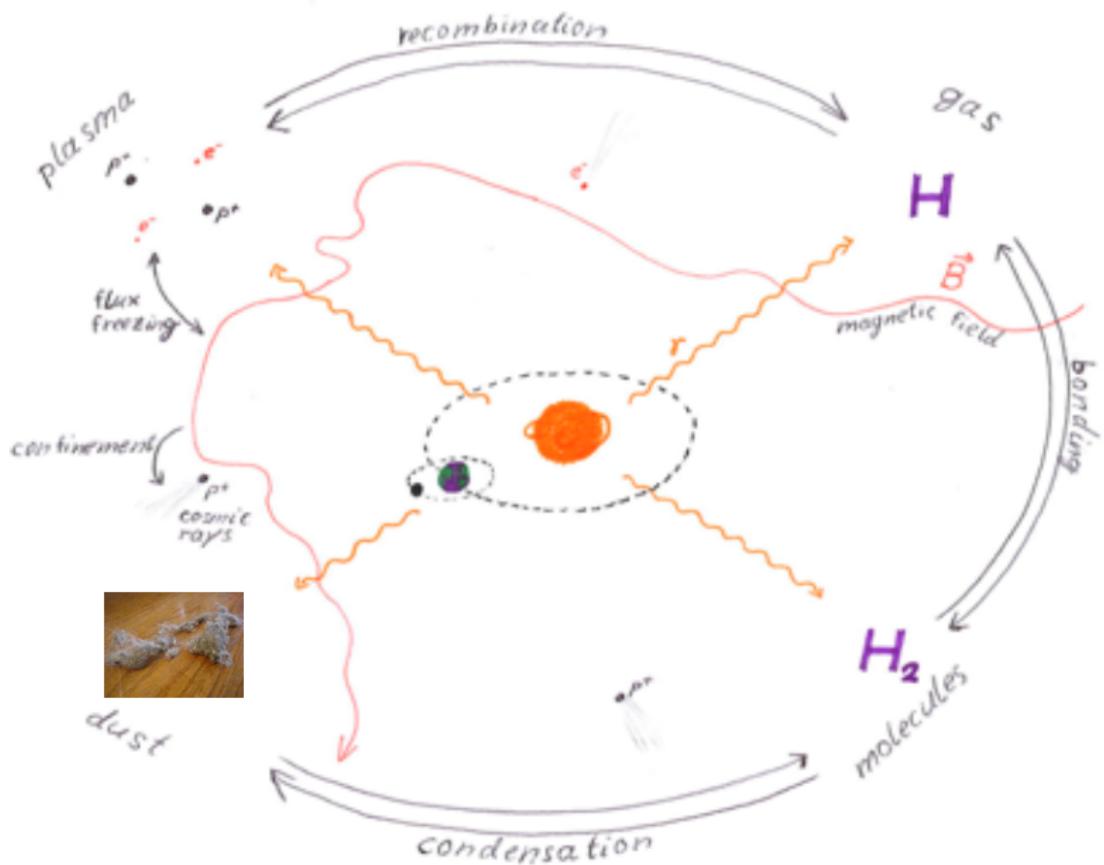
dust

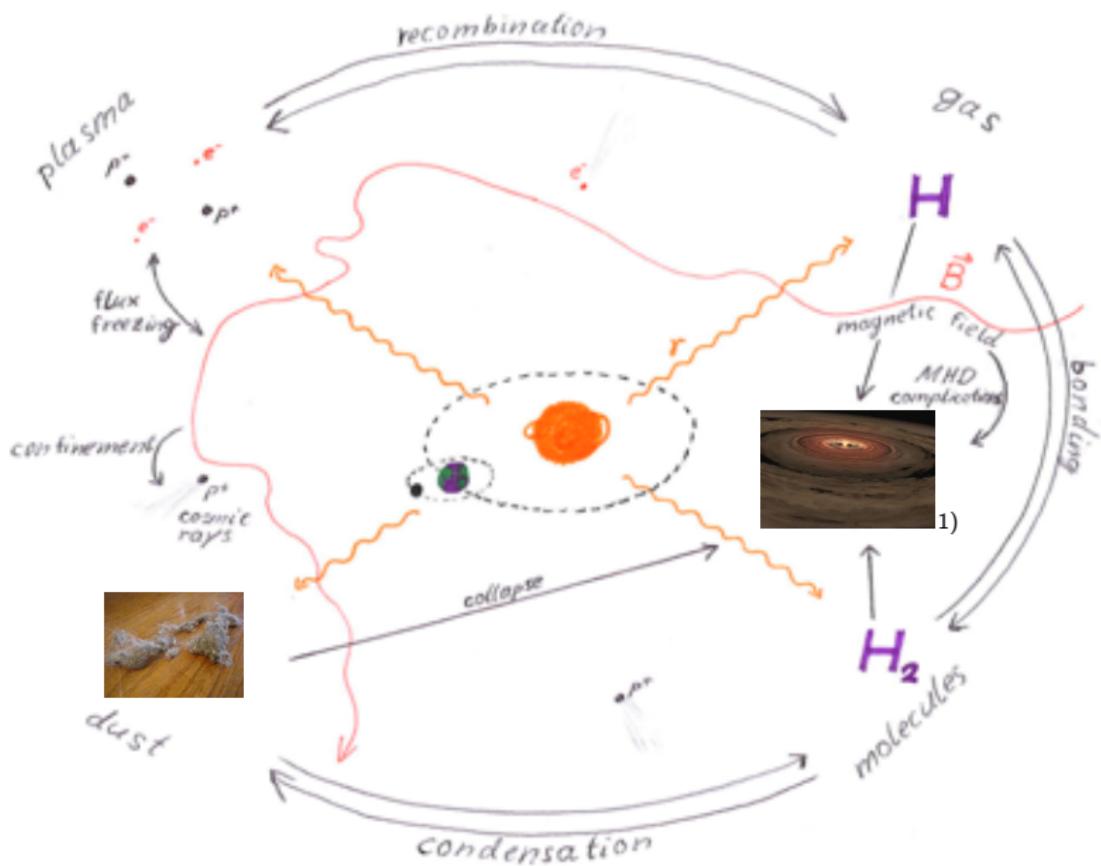
H_2
molecules

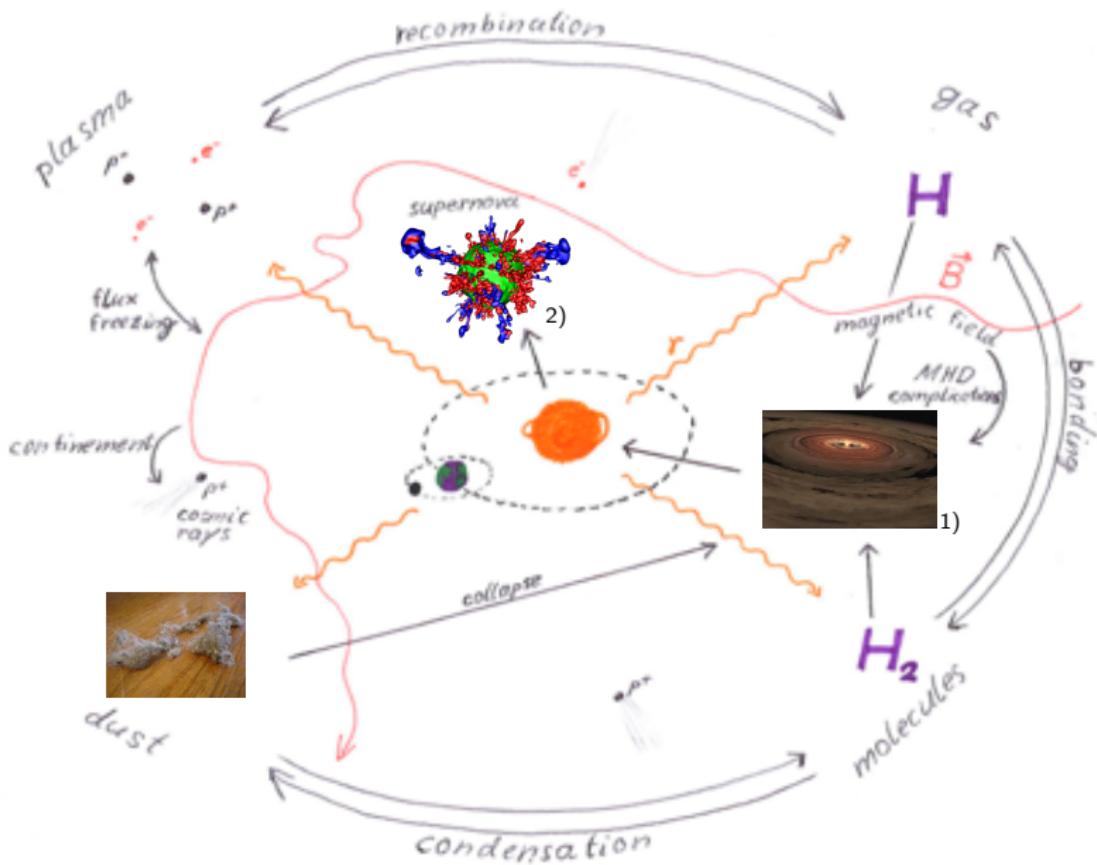


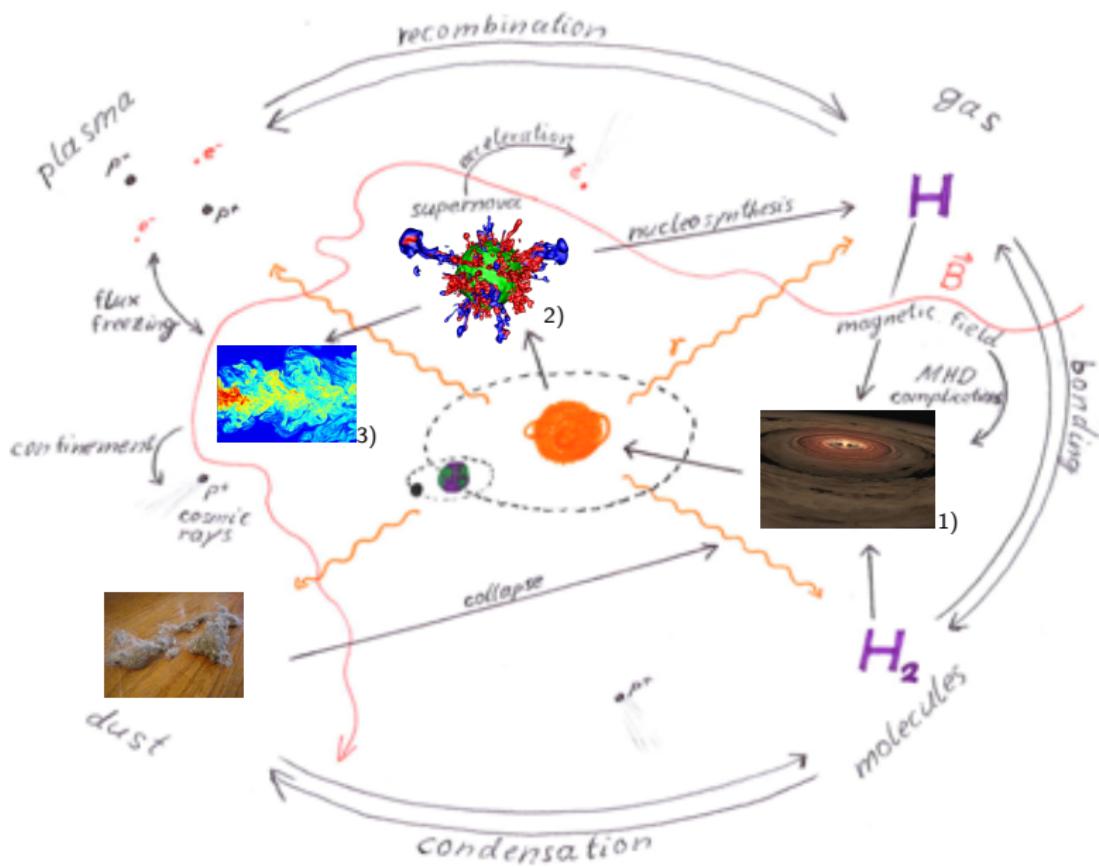












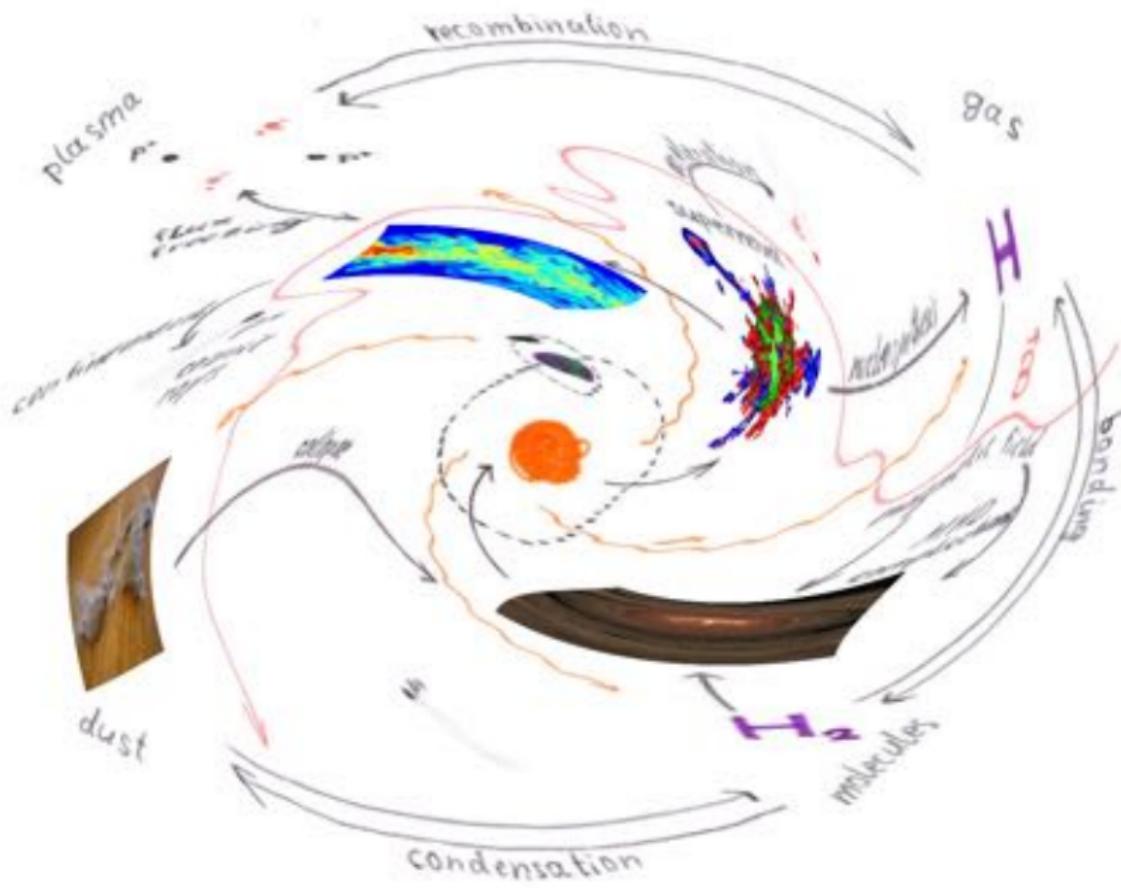
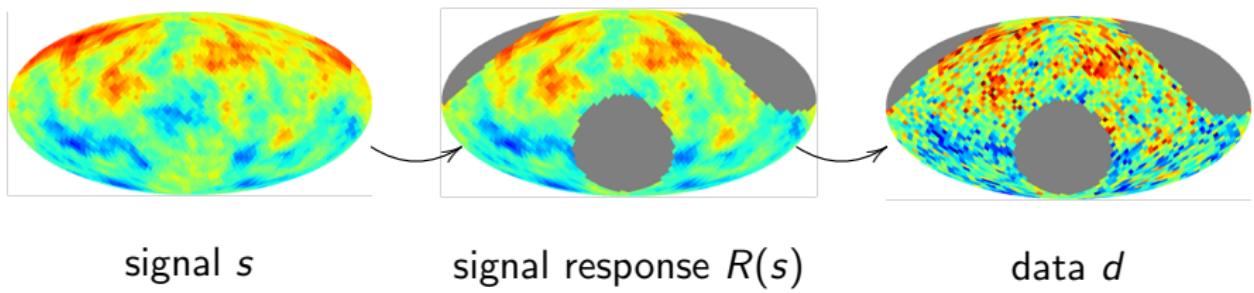
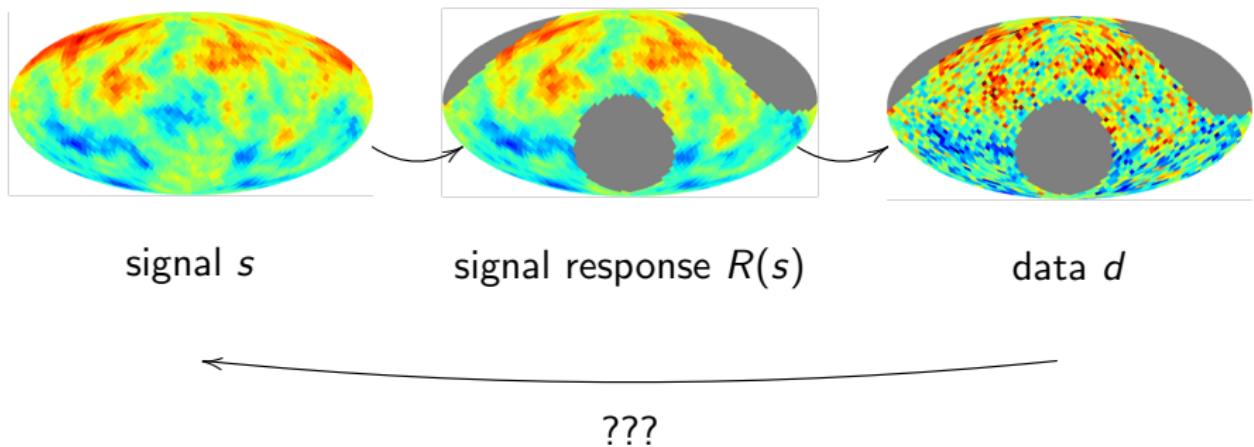


Image credits: 1) D. Darling; 2) N.J. Hammer/MPA; 3) C. Fukushima/TUDelft

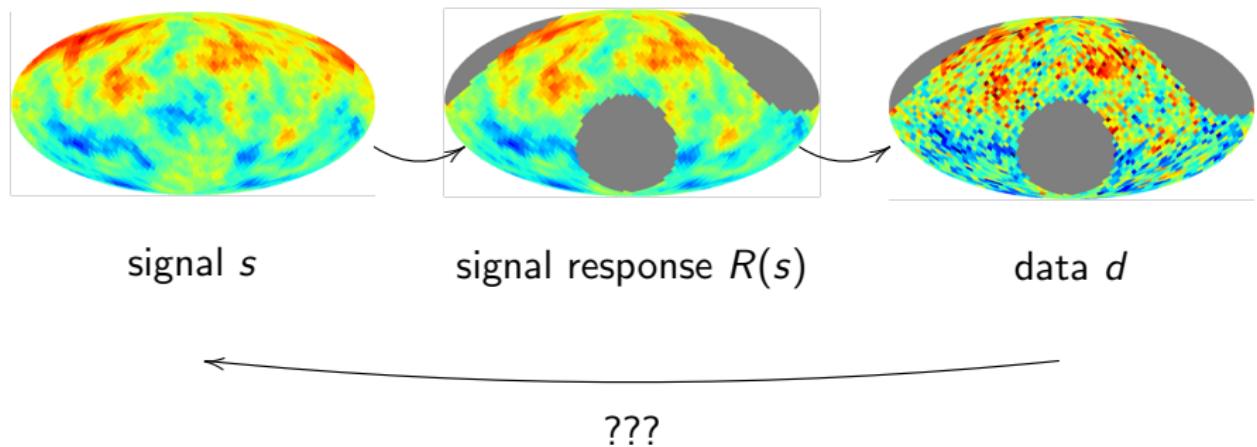
Signal inference



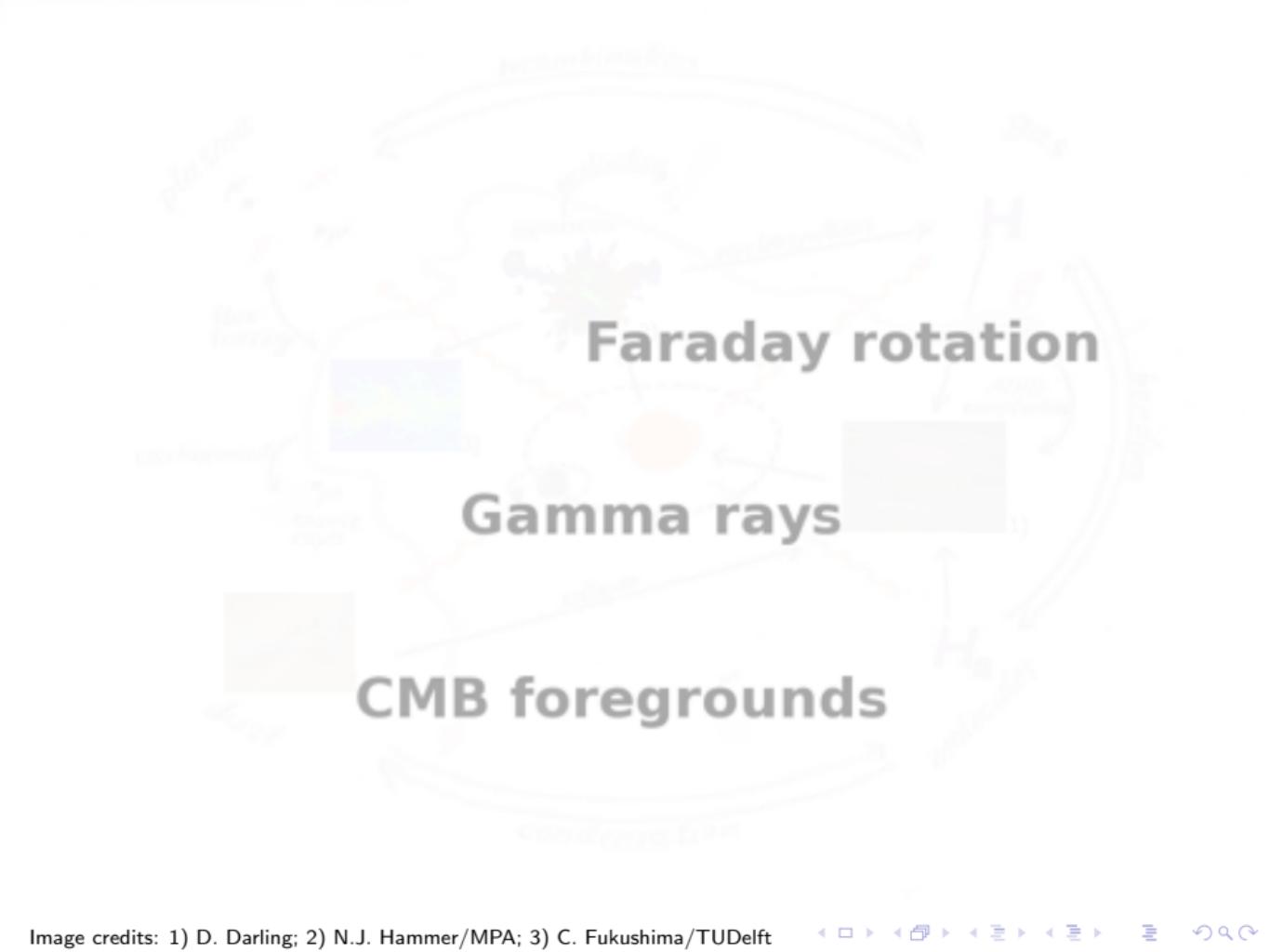
Signal inference



Signal inference



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

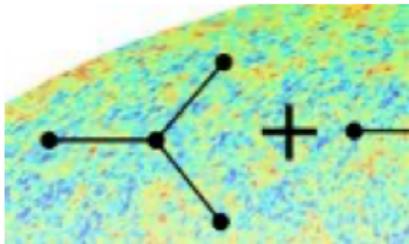


Faraday rotation

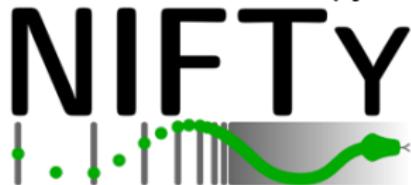
Gamma rays

CMB foregrounds

Information Field Theory



Numerical IFT for python



<http://www.mpa-garching.mpg.de/ift/>

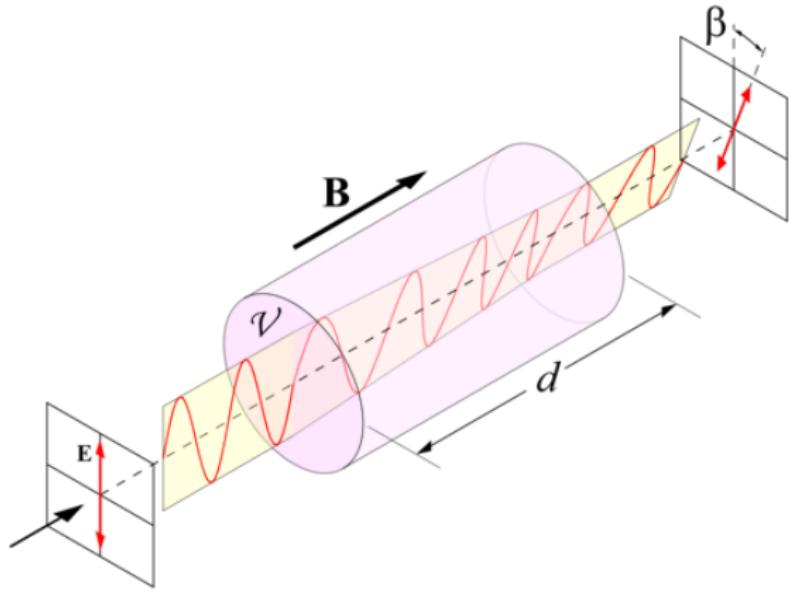
<http://www.mpa-garching.mpg.de/ift/nifty/>



Faraday rotation

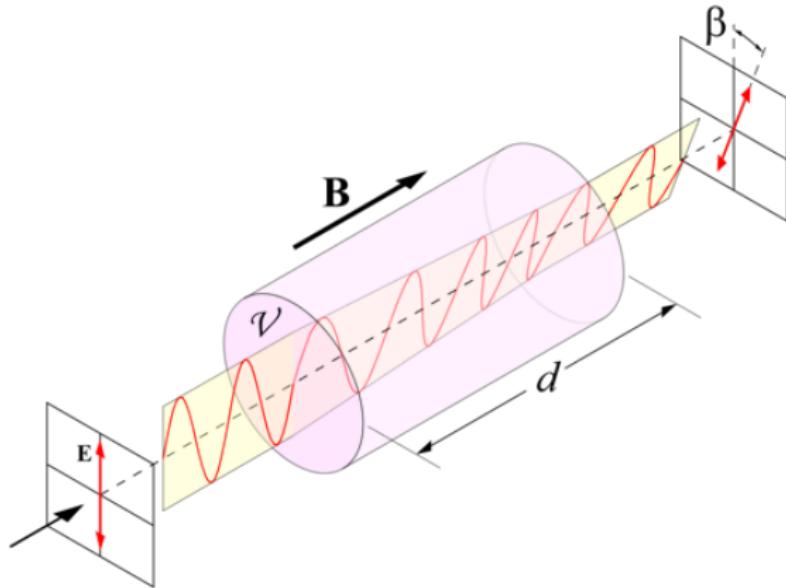
Gamma rays

CMB foregrounds



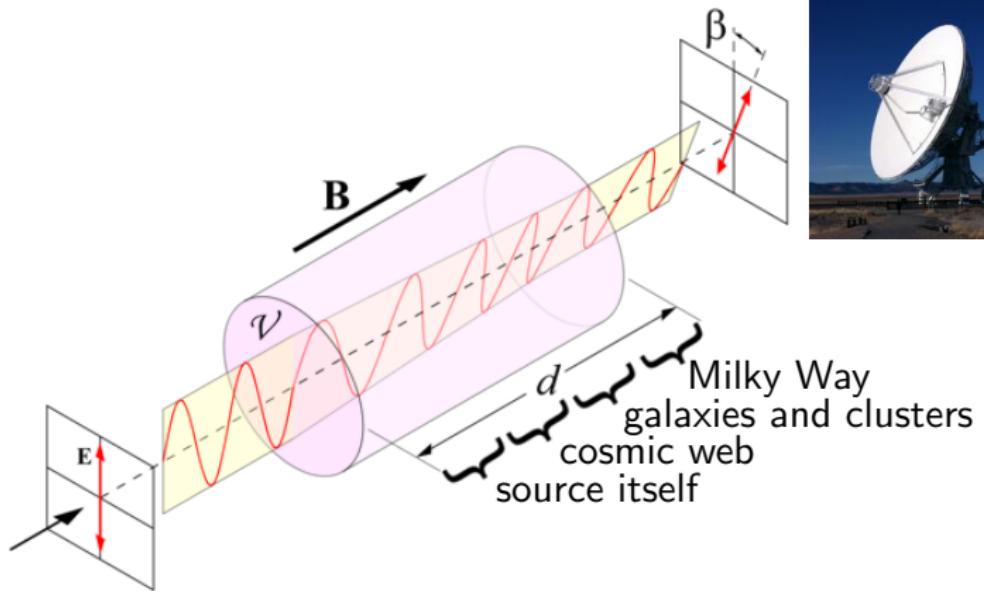
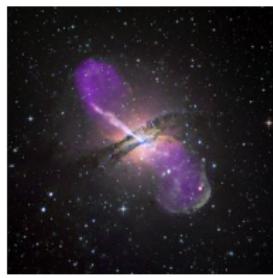
$$d\beta \propto \lambda^2 n_e(\vec{x}) B_r(\vec{x}) dr$$

$$\Rightarrow \beta \propto \lambda^2 \int_{r_{\text{source}}}^0 n_e(\vec{x}) B_r(\vec{x}) dr$$

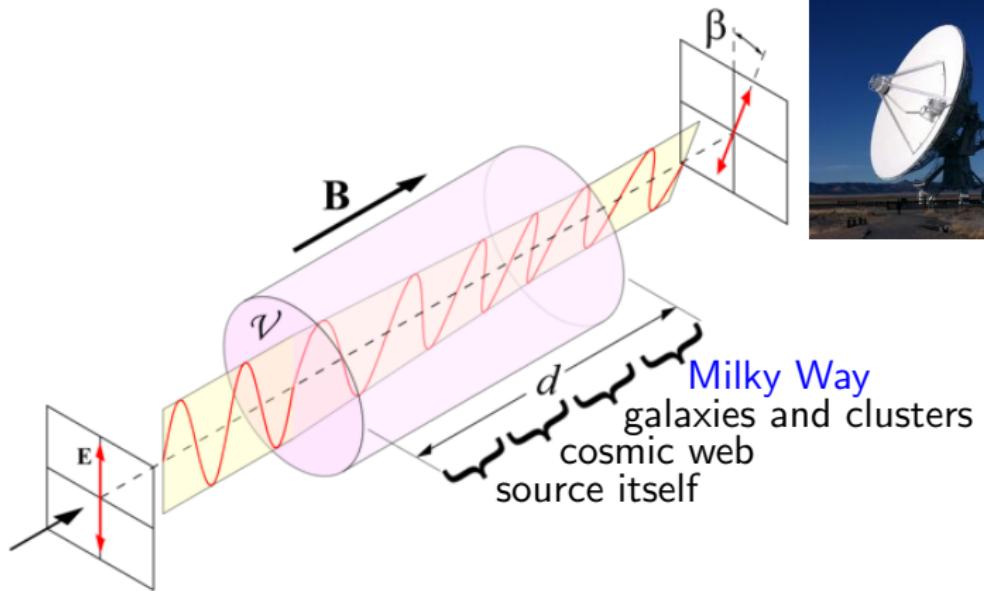
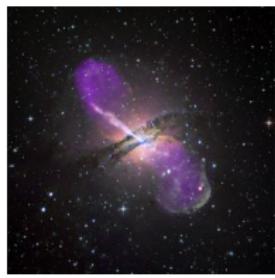


Faraday depth: $\phi \propto \int_{r_{\text{source}}}^0 n_e(\vec{x}) B_r(\vec{x}) dr$

$$\beta = \phi \lambda^2$$

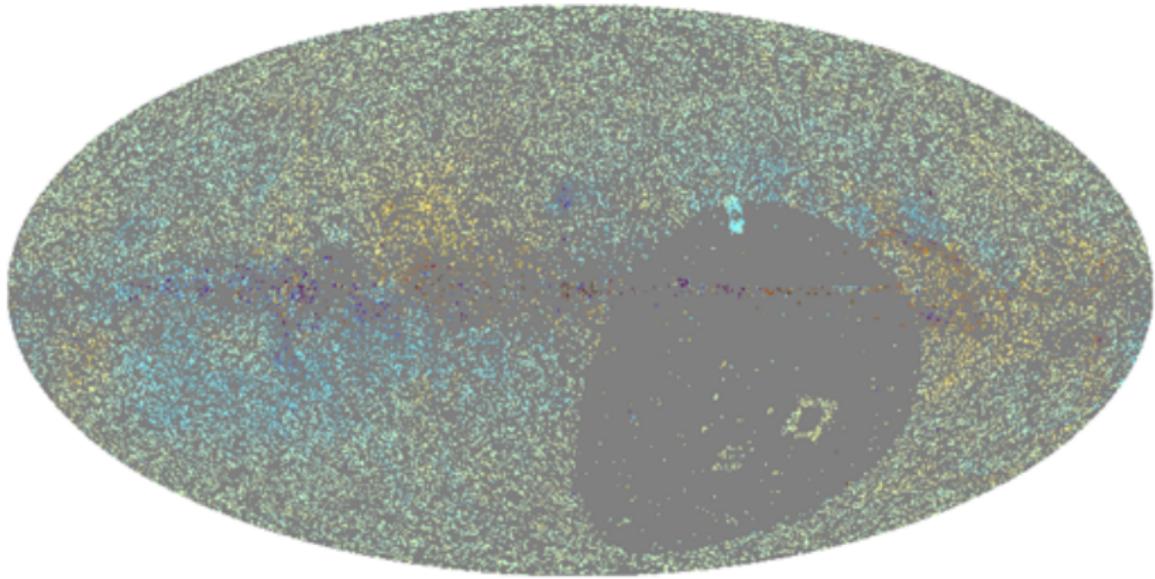


$$\text{Faraday depth: } \phi \propto \int_{r_{\text{source}}}^0 n_e(\vec{x}) B_r(\vec{x}) dr$$
$$\beta = \phi \lambda^2$$

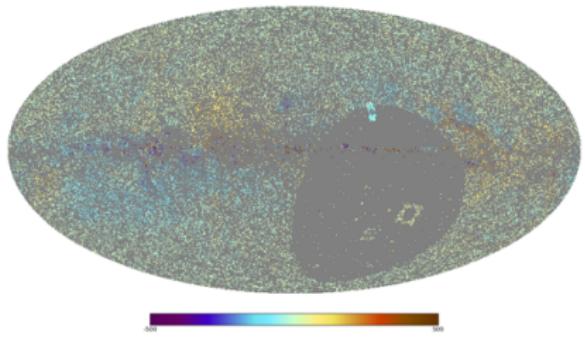


Galactic Faraday depth:

$$\phi \propto \int_{r_{\text{MilkyWay}}}^0 n_e(\vec{x}) B_r(\vec{x}) dr$$

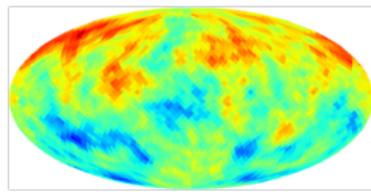


41 330 data points

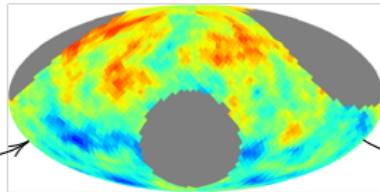


Challenges

- ▶ Regions without data
- ▶ Uncertain error bars:
 - ▶ complicated observations
 - ▶ $n\pi$ -ambiguity
 - ▶ extragalactic contributions unknown

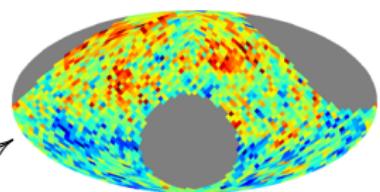


signal s



$R \cdot$

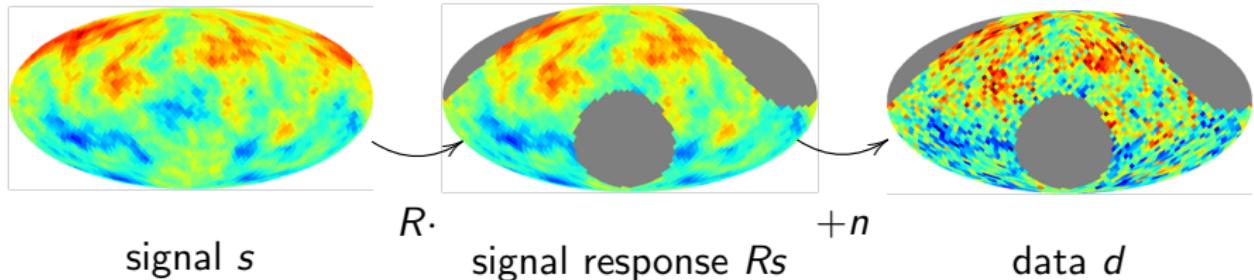
signal response Rs



$+n$

data d

$$d = Rs + n$$

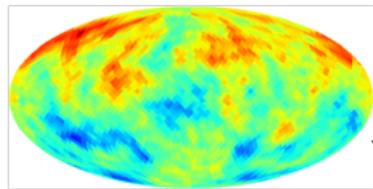


$$\mathcal{P}(s) = \mathcal{G}(s, S)$$

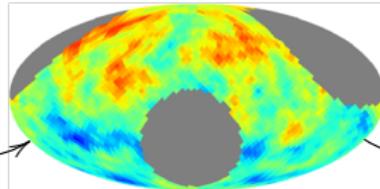
$$\mathcal{P}(n) = \mathcal{G}(n, N)$$

$$d = Rs + n$$

$$\mathcal{G}(s, S) = \frac{1}{|2\pi S|^{1/2}} \exp \left[\frac{1}{2} s^\dagger S^{-1} s \right]$$



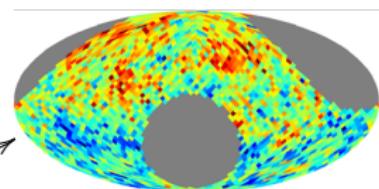
signal s



$R \cdot$

signal response Rs

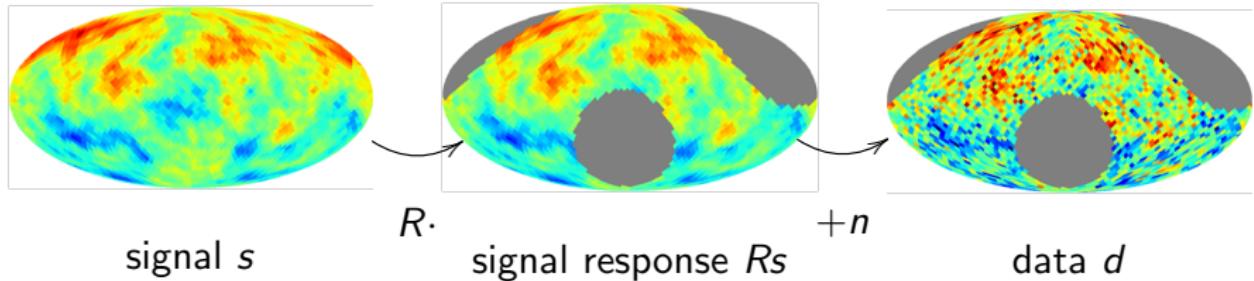
$+ n$



data d

$$d = Rs + n$$

$$m = \int \mathcal{D}s \ s \ \mathcal{P}(s|d)$$



Wiener Filter

$$d = R s + n$$

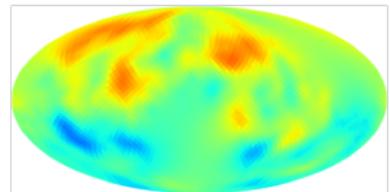
$$m = \int \mathcal{D} s \ s \ \mathcal{P}(s|d)$$

$m = Dj$, where

$$j = R^\dagger N^{-1} d$$

$$D = (S^{-1} + R^\dagger N^{-1} R)^{-1}$$

$$\downarrow DR^\dagger N^{-1}.$$



$$S(\hat{n}, \hat{n}') = \int \mathcal{D}s \ s(\hat{n})s(\hat{n}')\mathcal{P}(s)$$

$$\Rightarrow S_{(\ell m), (\ell' m')} = \int \mathcal{D}s \ s_{\ell m} s_{\ell' m'}^* \mathcal{P}(s)$$

$$\begin{aligned} S(\hat{n}, \hat{n}') &= \int \mathcal{D}s \ s(\hat{n})s(\hat{n}')\mathcal{P}(s) \\ &= S(\hat{n} \cdot \hat{n}') \\ \Rightarrow S_{(\ell m), (\ell' m')} &= \int \mathcal{D}s \ s_{\ell m} s_{\ell' m'}^* \mathcal{P}(s) \\ &= \delta_{\ell\ell'} \delta_{mm'} C_\ell \end{aligned}$$

↪ angular power spectrum

$$\begin{aligned}
S(\hat{n}, \hat{n}') &= \int \mathcal{D}s \ s(\hat{n})s(\hat{n}')\mathcal{P}(s) \\
&= S(\hat{n} \cdot \hat{n}') \\
\Rightarrow S_{(\ell m), (\ell' m')} &= \int \mathcal{D}s \ s_{\ell m} s_{\ell' m'}^* \mathcal{P}(s) \\
&= \delta_{\ell\ell'} \delta_{mm'} \textcolor{blue}{C}_\ell \\
&\hookrightarrow \text{angular power spectrum}
\end{aligned}$$

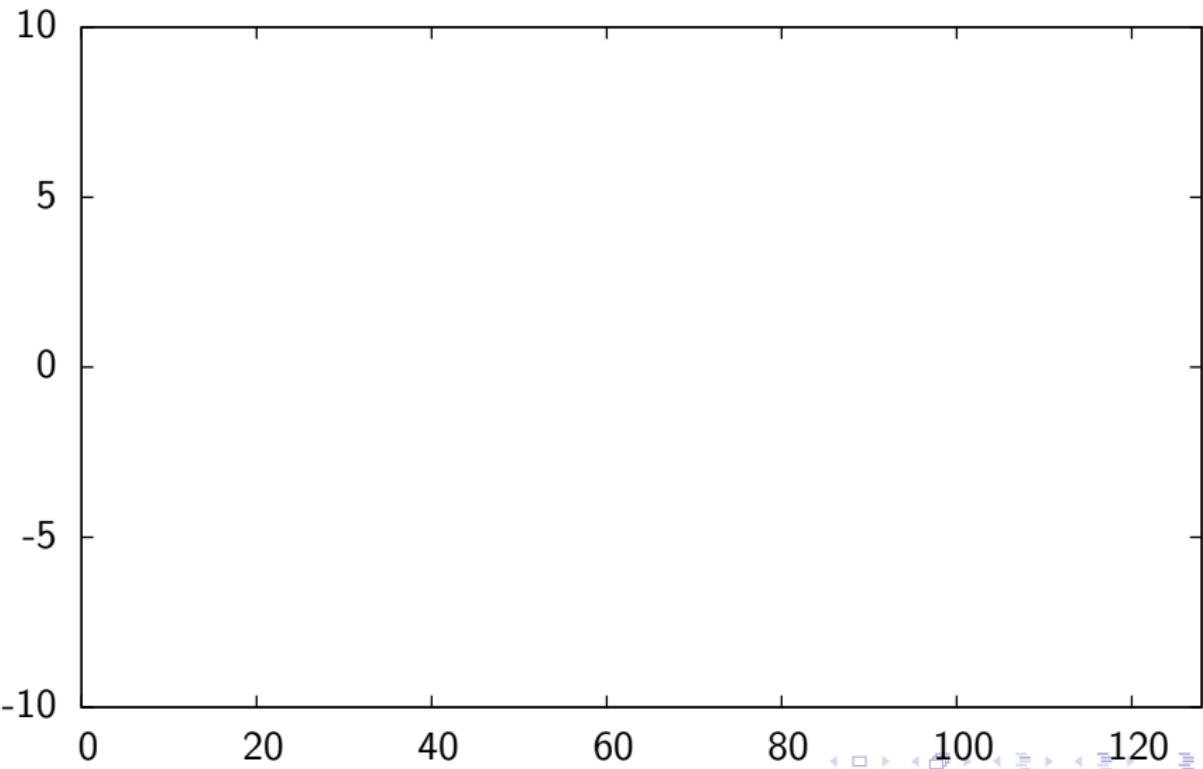
$$N_{ij} = \delta_{ij} \sigma_i^2 \textcolor{blue}{\eta_i}$$

\hookrightarrow error bar correction factors

(uncorrelated noise)

1D example

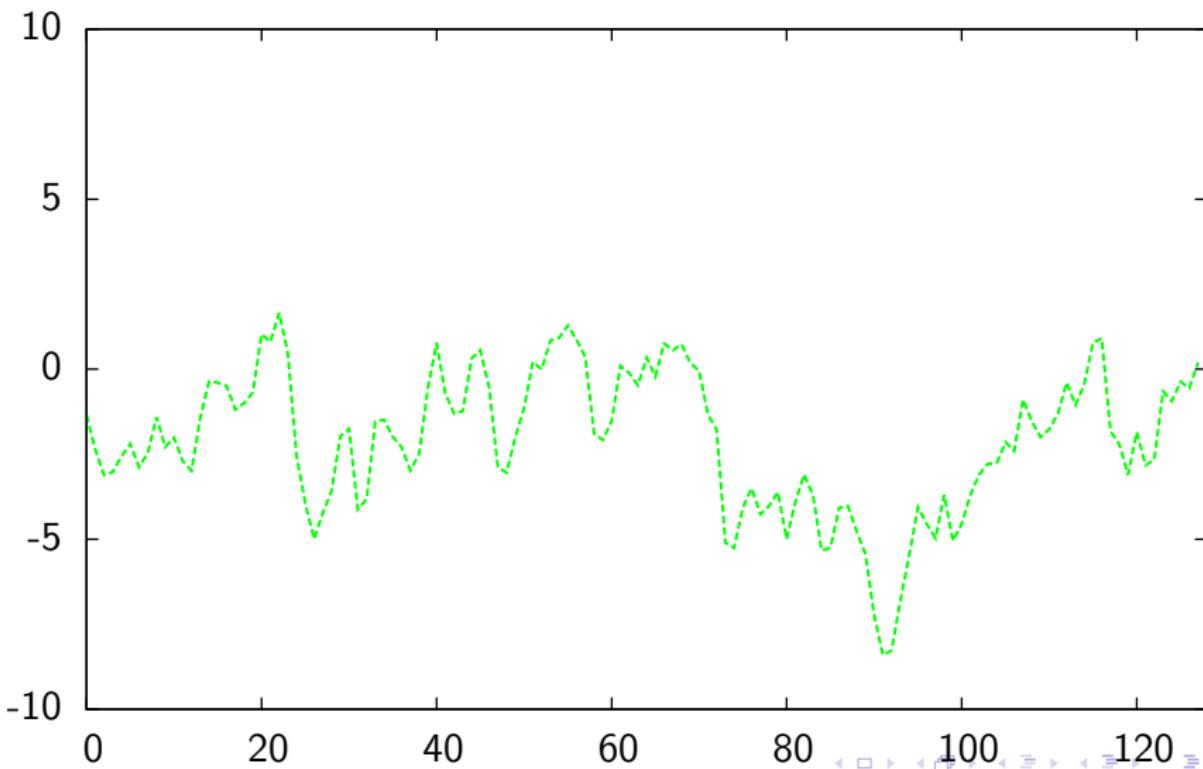
Assumptions:



1D example

Assumptions:

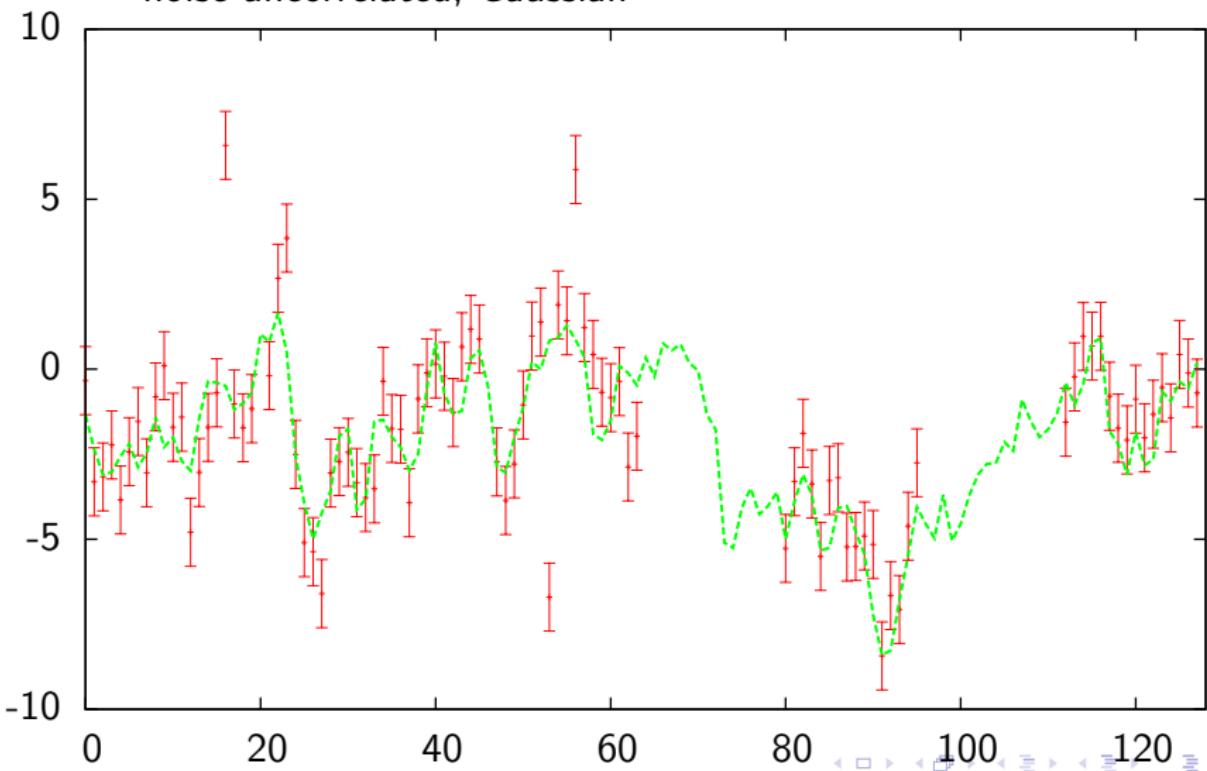
- ▶ signal field statistically homogeneous Gaussian random field
- ▶



1D example

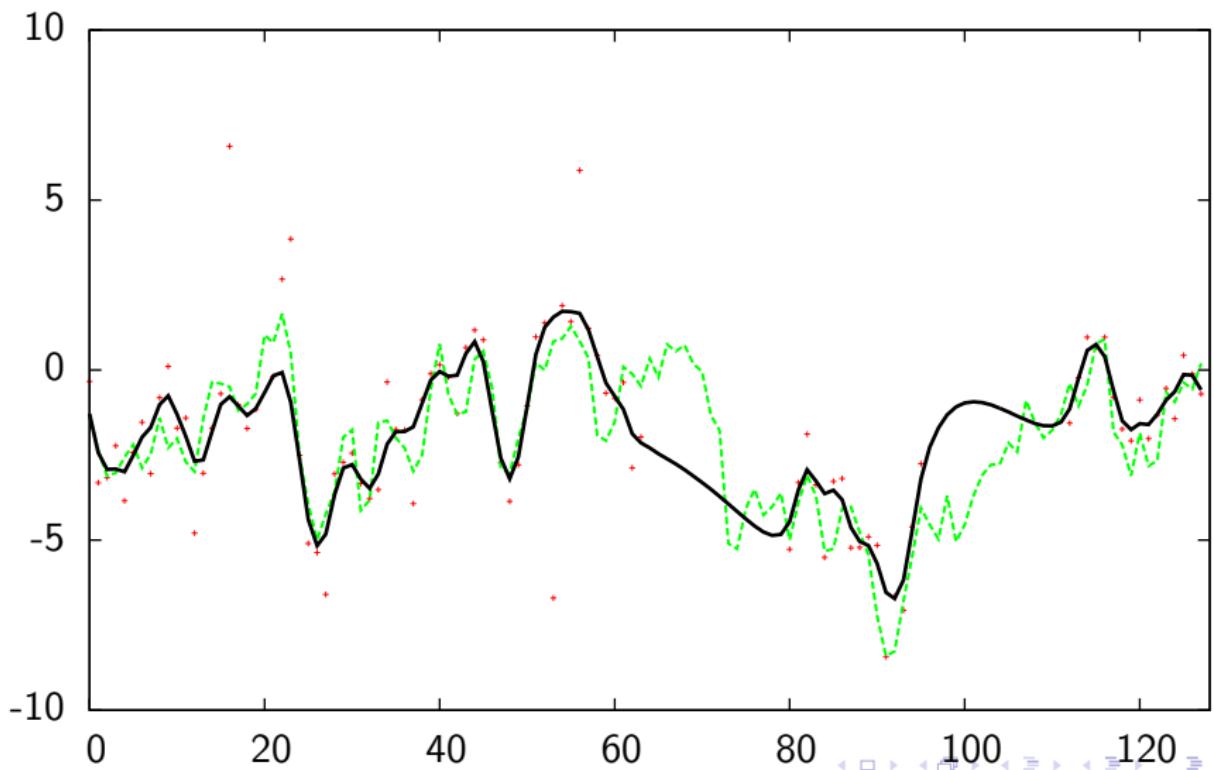
Assumptions:

- ▶ signal field statistically homogeneous Gaussian random field
- ▶ noise uncorrelated, Gaussian



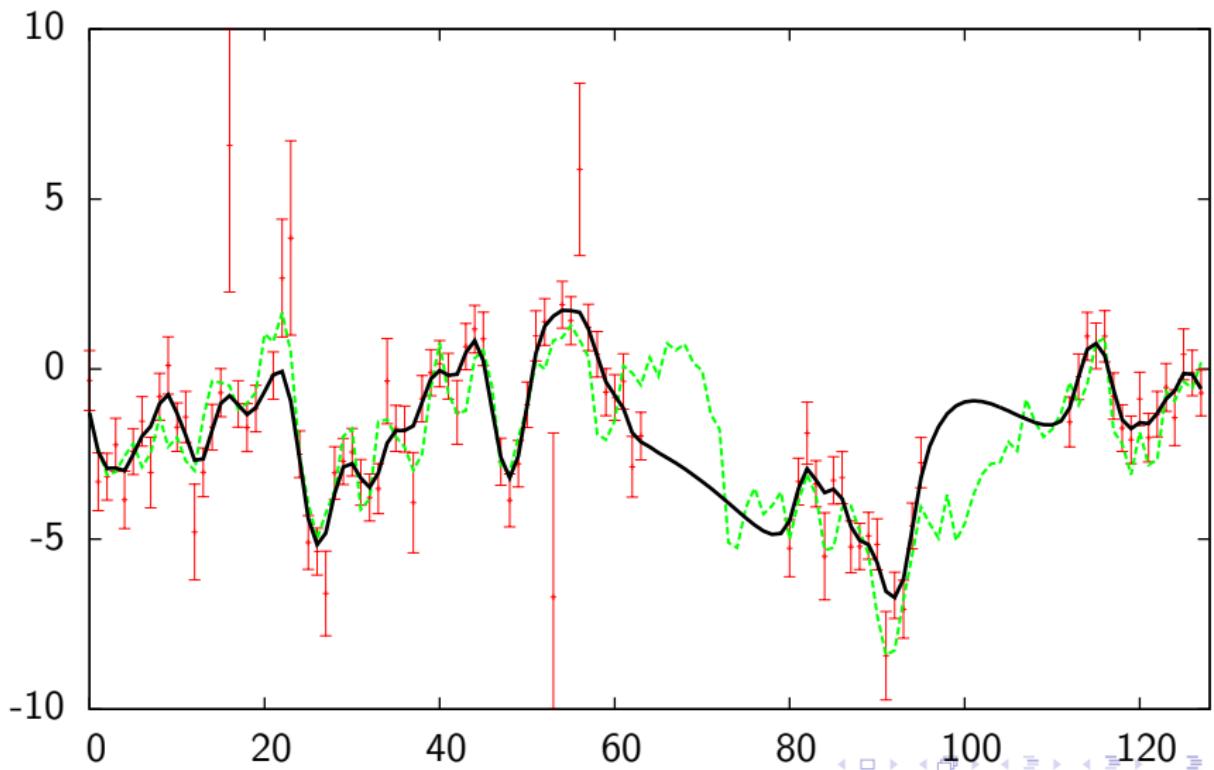
1D example

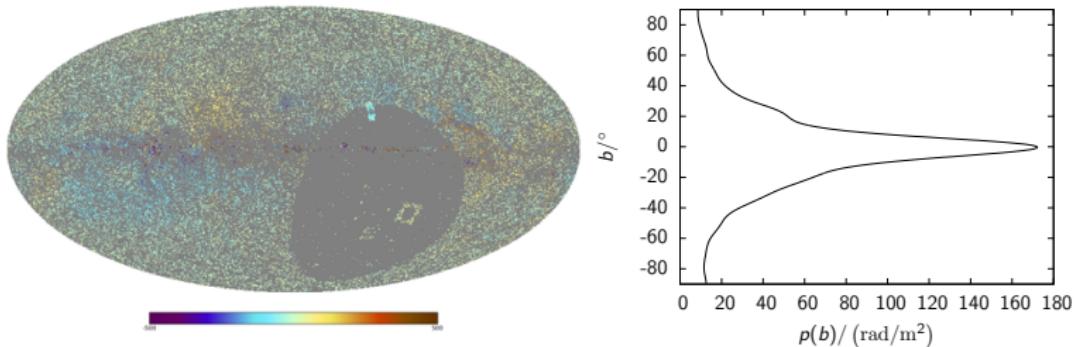
- ▶ Reconstruct (iteratively):
signal, power spectrum, noise variance



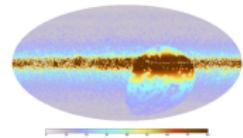
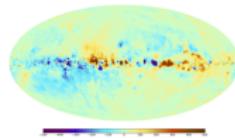
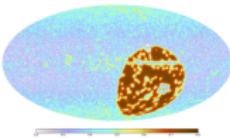
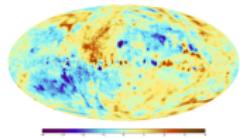
1D example

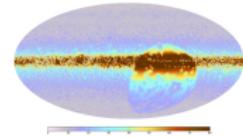
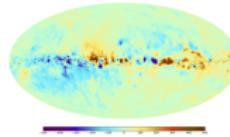
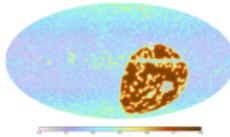
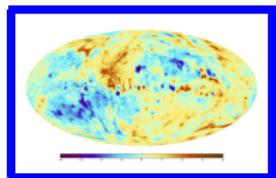
- ▶ Reconstruct (iteratively):
signal, power spectrum, noise variance



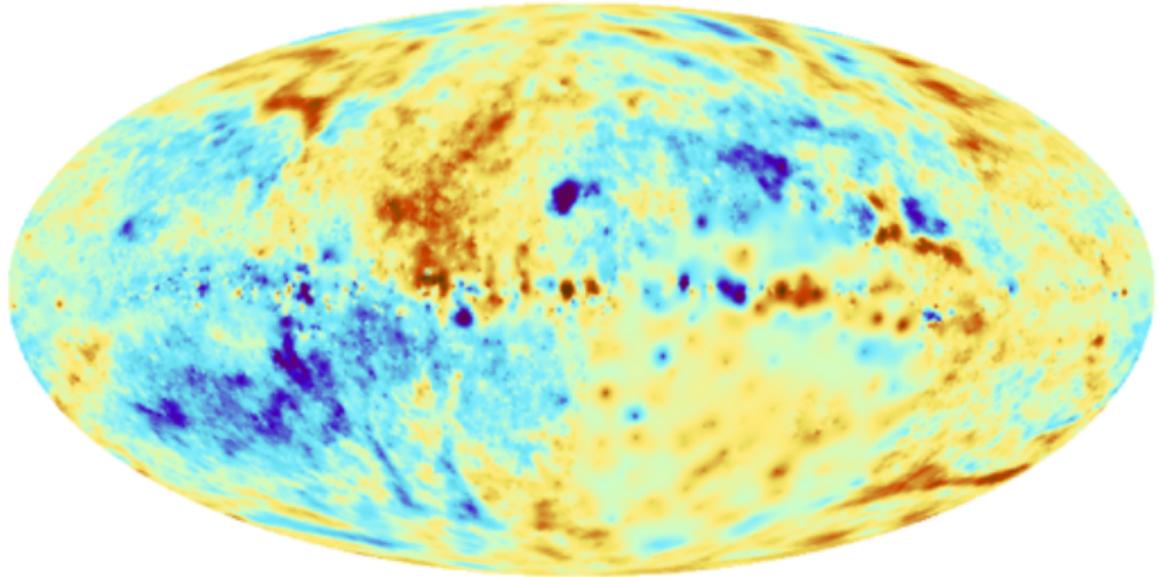


- ▶ Approximate $s(b, l) := \frac{\phi(b, l)}{p(b)}$ as a statistically isotropic Gaussian field
- ▶ R : multiplication with $p(b)$ and projection on directions of sources
- ▶ $N_{ij} = \delta_{ij}\eta_i\sigma_i^2$



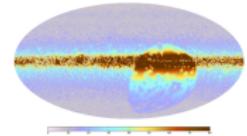
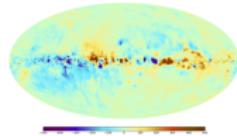
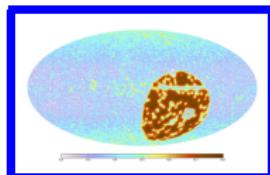
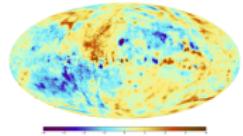


posterior mean of the signal

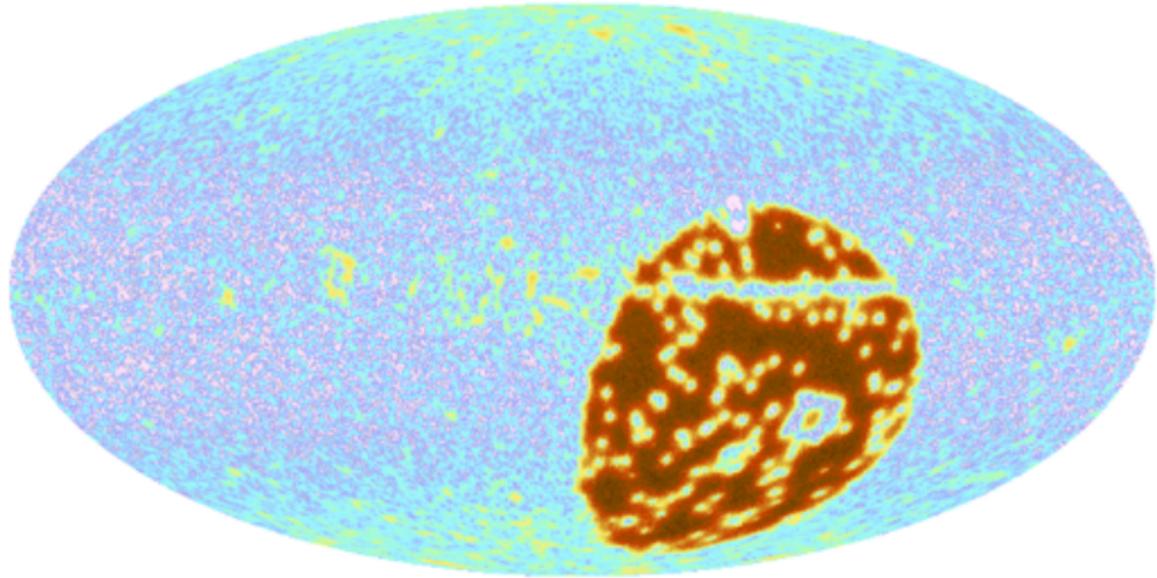


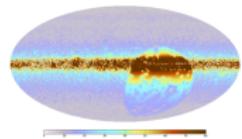
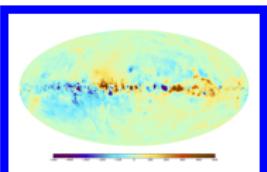
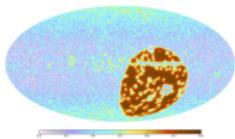
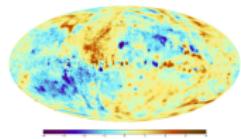
A horizontal color bar representing a numerical scale from -4 to 4. The colors transition from dark purple at -4 to black at 0, then through cyan, yellow, and orange towards red at 4.

m

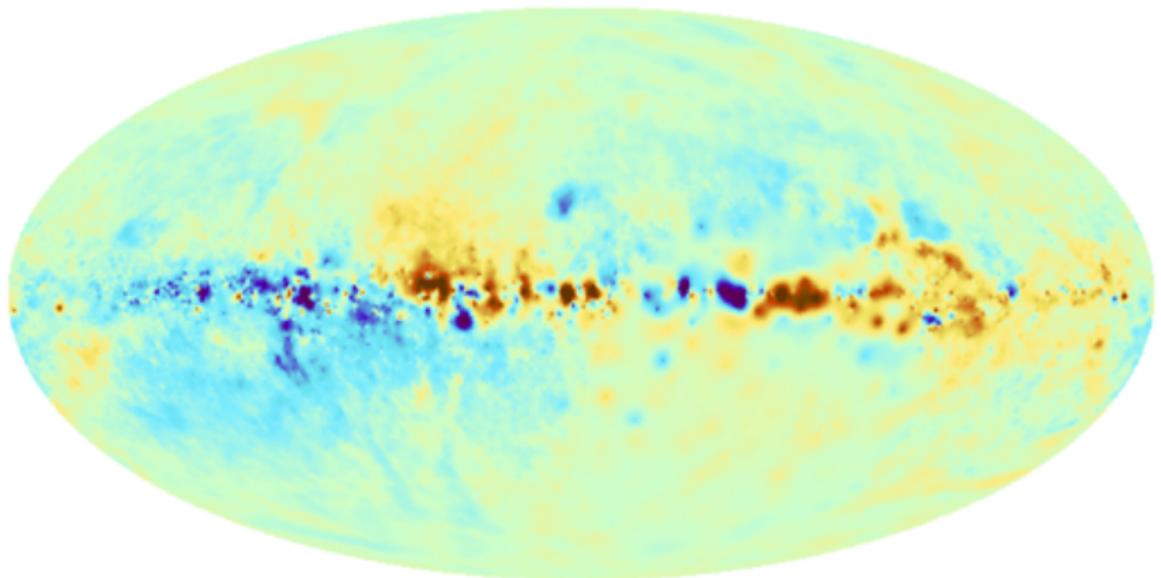


uncertainty of the signal map



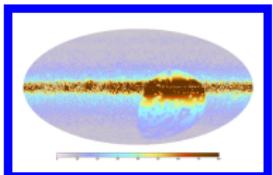
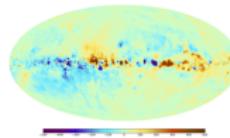
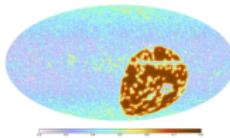
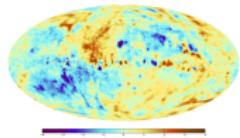


posterior mean of the Faraday depth

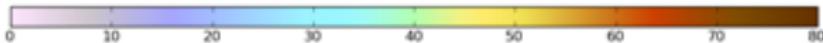
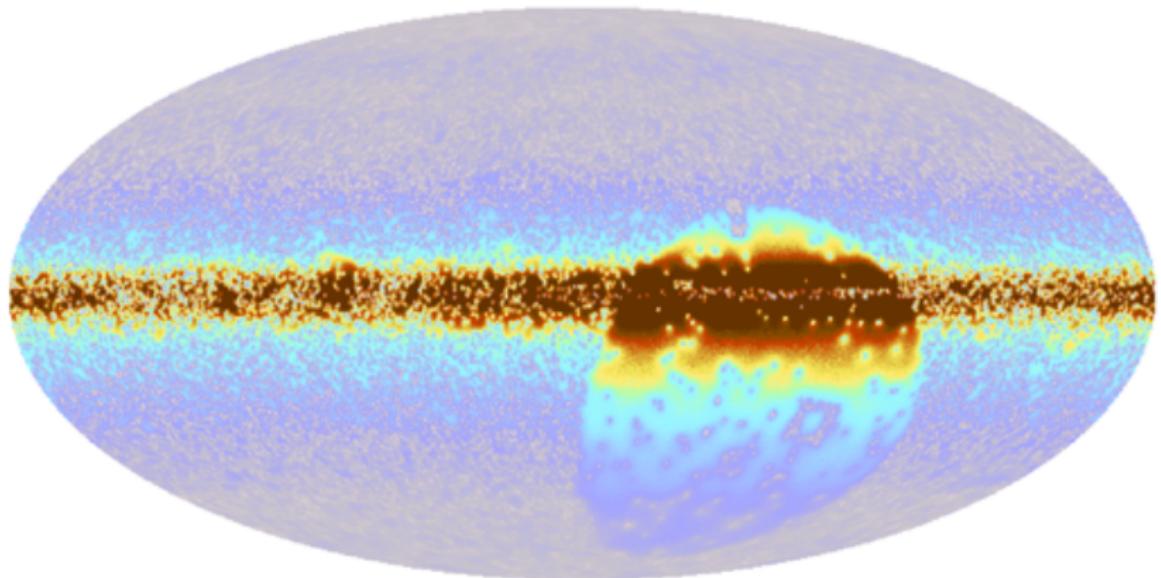


A horizontal color bar representing a gradient from purple (-500) to red (+500). The scale is marked at intervals of 100, ranging from -500 to 500.

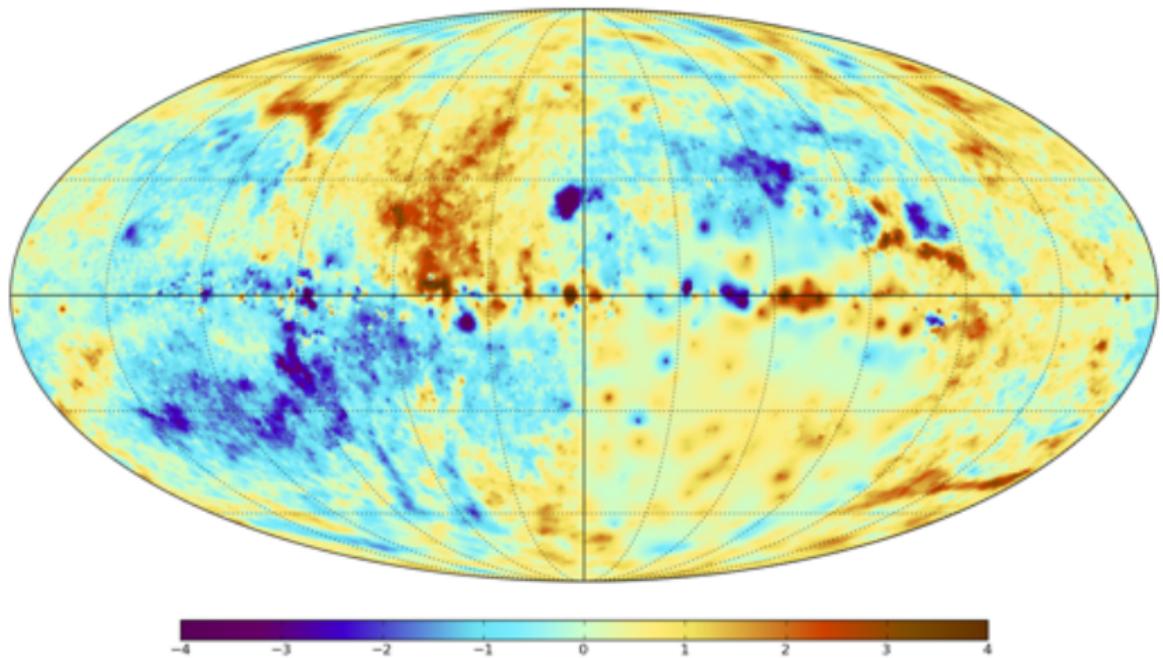
pm

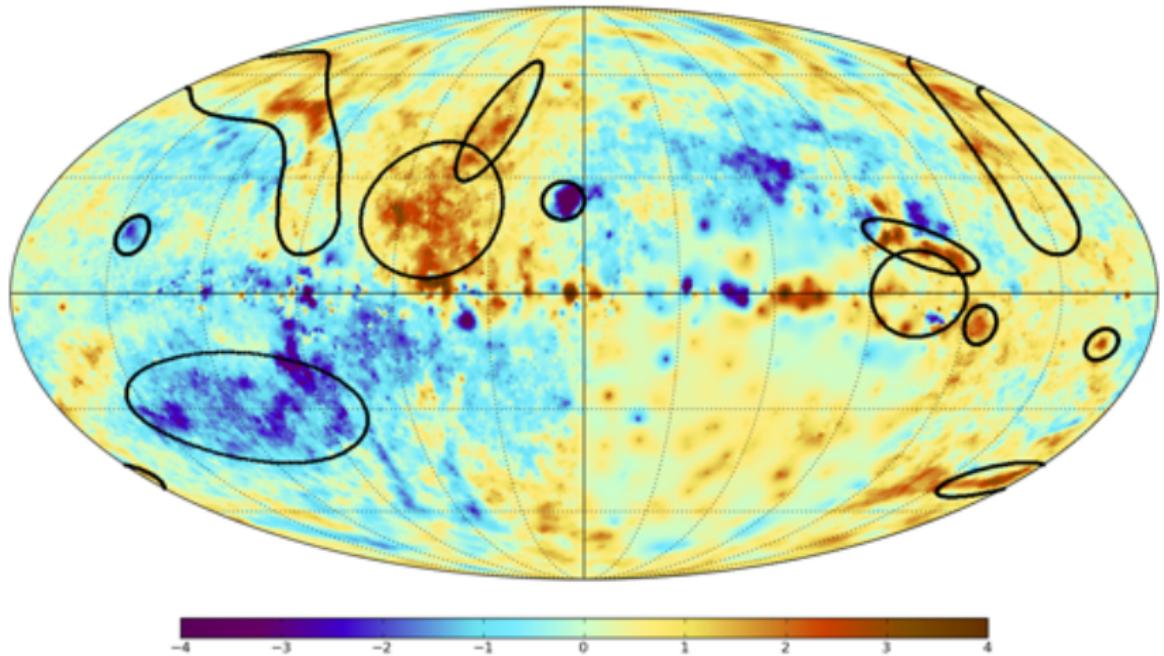


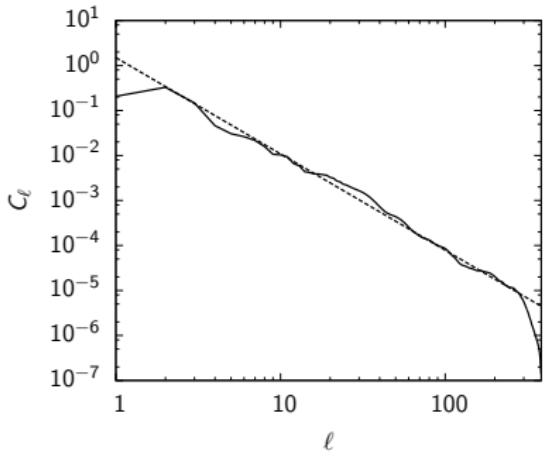
uncertainty of the Faraday depth



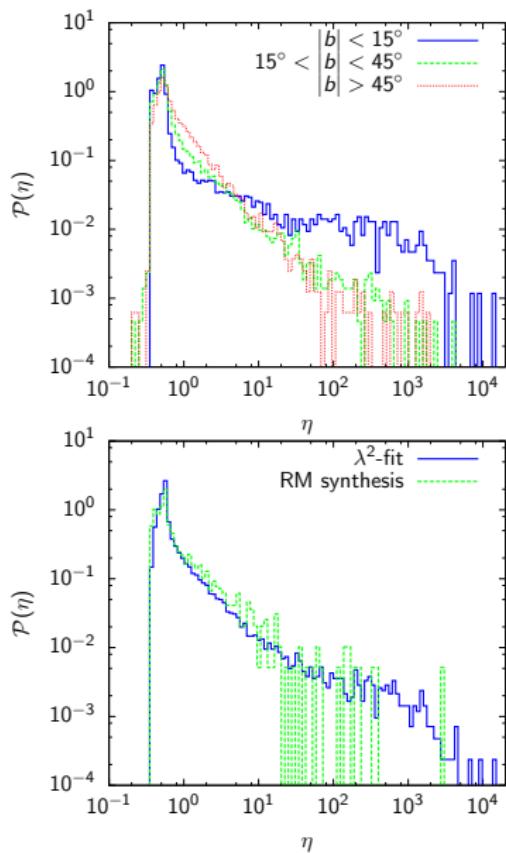
$$p\sqrt{\text{diag}(D)}$$







$$C_\ell \propto \ell^{-2.17}$$



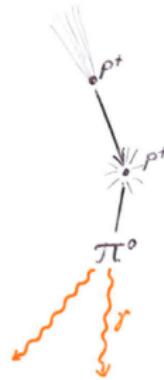
$$N_{ij} = \langle n_i n_j \rangle = \delta_{ij} \eta_i \sigma_i^2$$

Faraday rotation

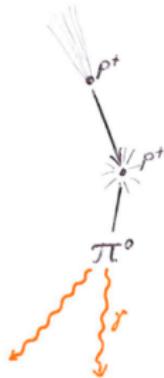
Gamma rays

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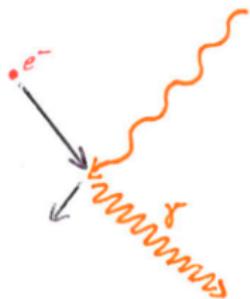
Pion decay



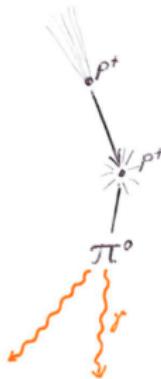
Pion decay



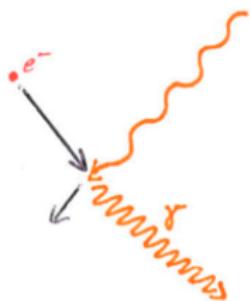
Inverse-Compton



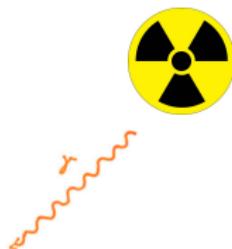
Pion decay



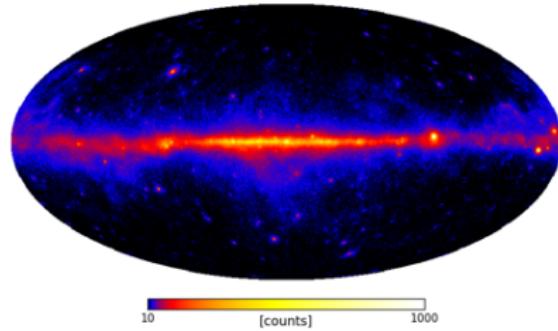
Inverse-Compton



Radioactive decay



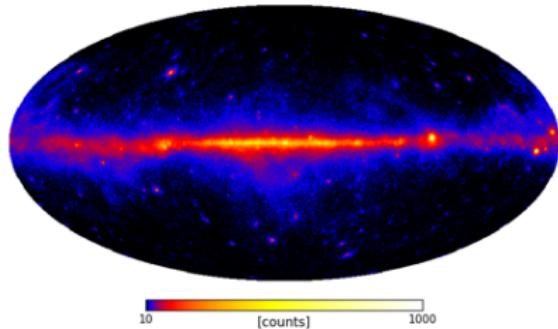
Gamma-ray photon counts:



FERMI data

- ▶ non-Gaussian
- ▶ always positive
- ▶ varying over several orders of magnitude

Gamma-ray photon counts:

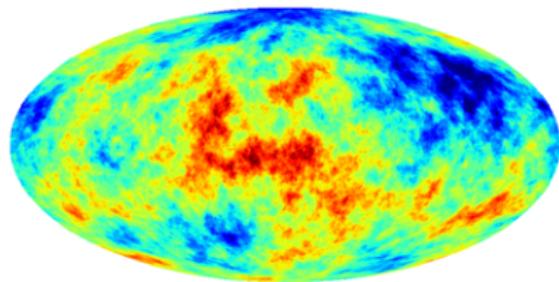


FERMI data

- ▶ non-Gaussian
- ▶ always positive
- ▶ varying over several orders of magnitude

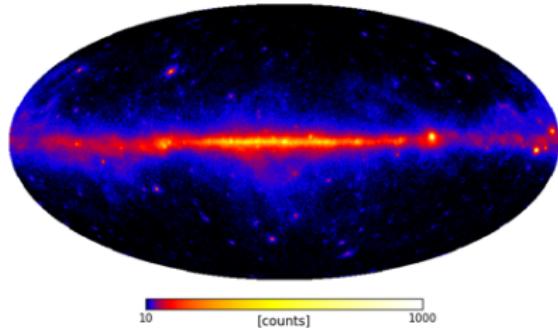
The log-normal model

- ▶ use logarithm of photon flux density as signal
- ▶ model this as Gaussian random field



s

Gamma-ray photon counts:

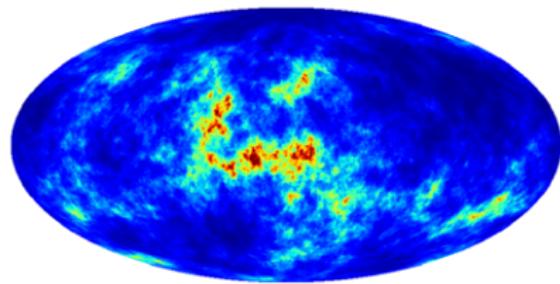


FERMI data

- ▶ non-Gaussian
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- ▶ varying over several orders of magnitude

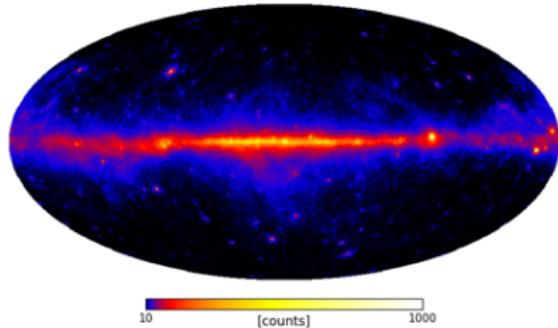
The log-normal model

- ▶ use logarithm of photon flux density as signal
- ▶ model this as Gaussian random field



$$e^s$$

Gamma-ray photon counts:

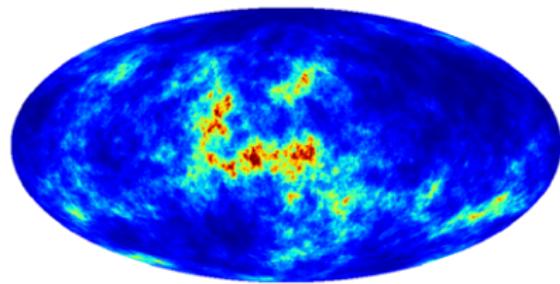


FERMI data

- ▶ non-Gaussian
- ▶ always positive
- ▶ varying over several orders of magnitude

The log-normal model

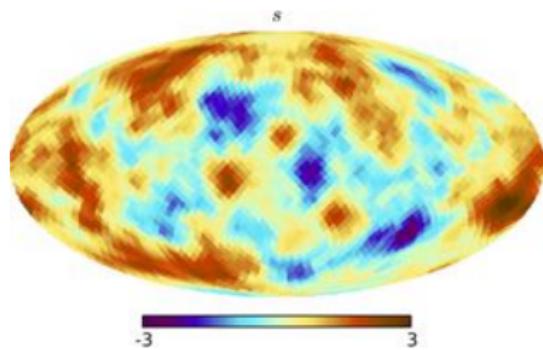
- ▶ use logarithm of photon flux density as signal
- ▶ model this as Gaussian random field



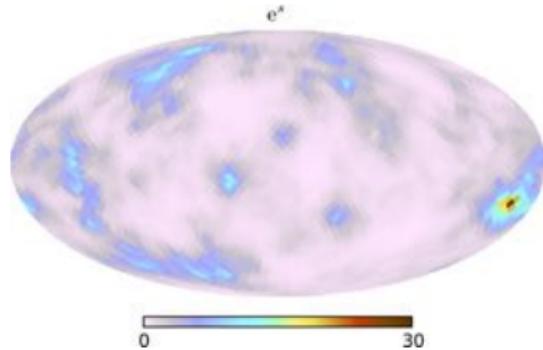
e^s

$$d = Re^s + n$$

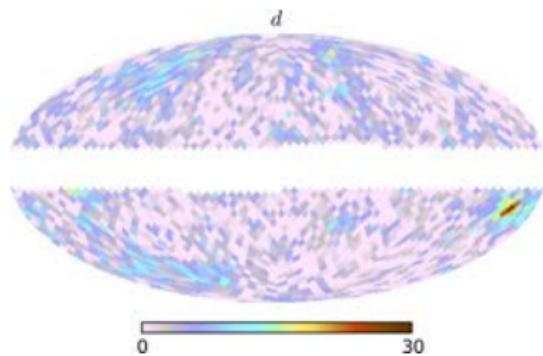
signal



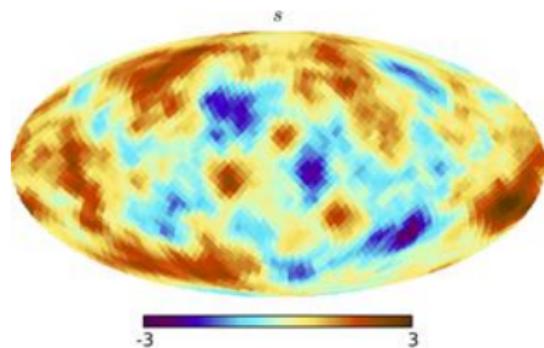
exponentiated signal



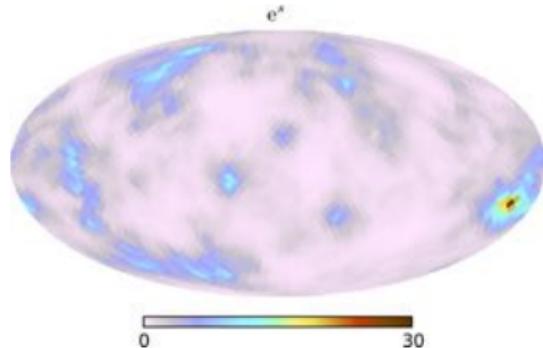
data



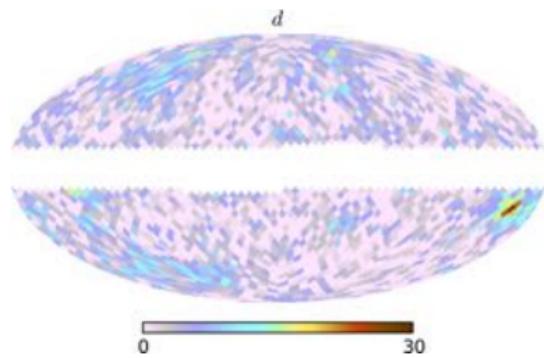
signal



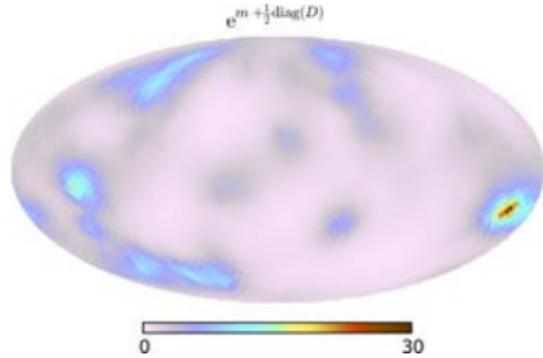
exponentiated signal



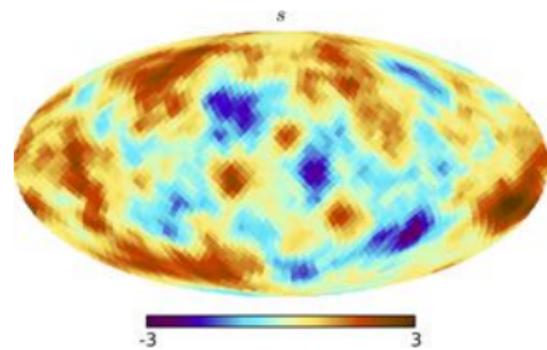
data



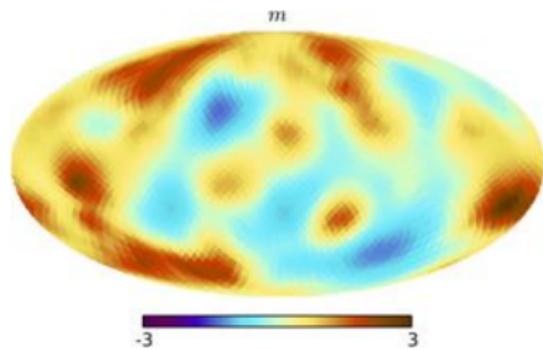
reconstructed exponentiated signal



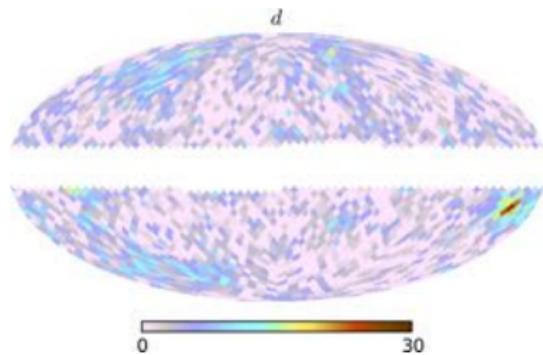
signal



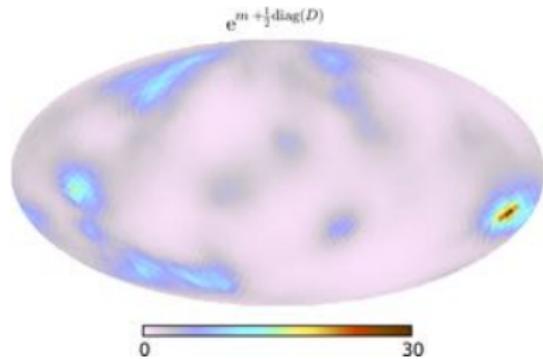
reconstructed signal



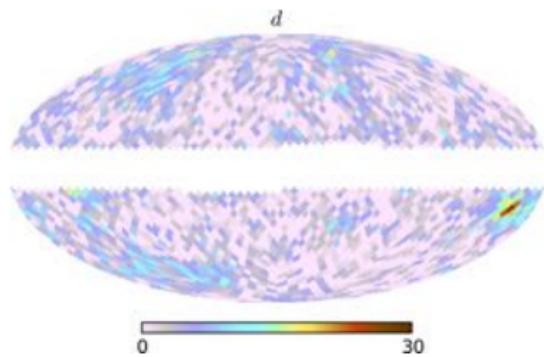
data



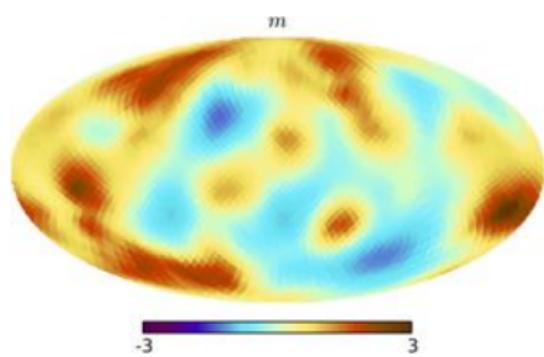
reconstructed exponentiated signal



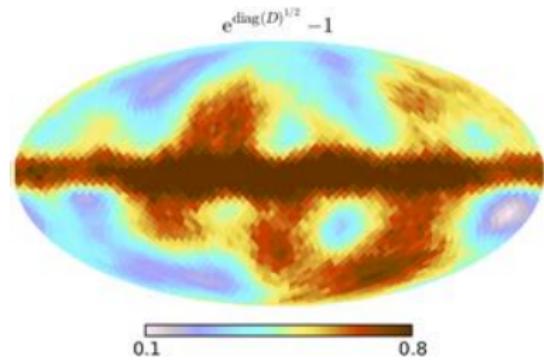
uncertainty



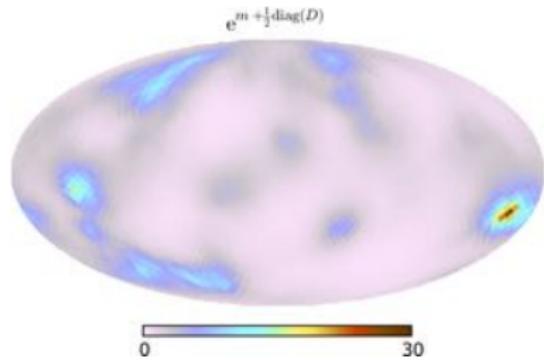
reconstructed signal



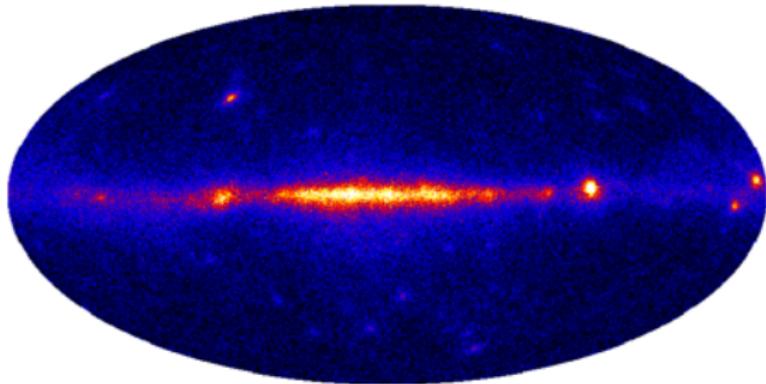
fractional uncertainty



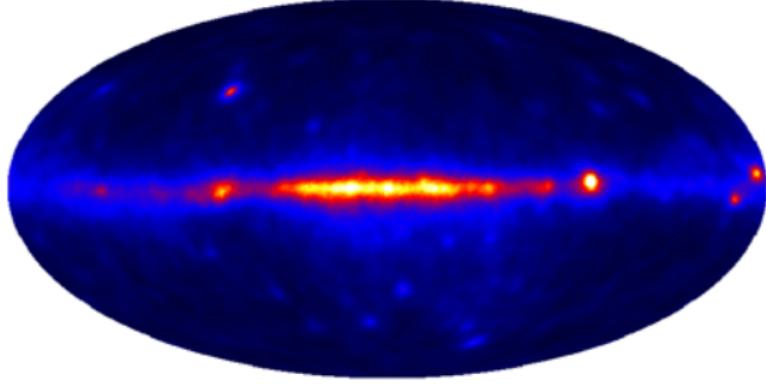
reconstructed exponentiated signal



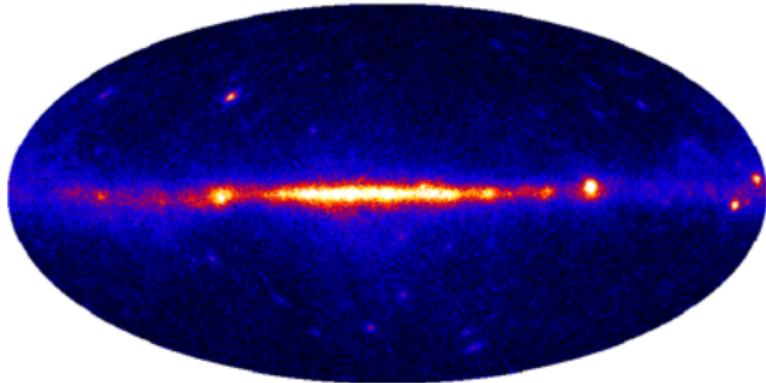
100 - 158 MeV



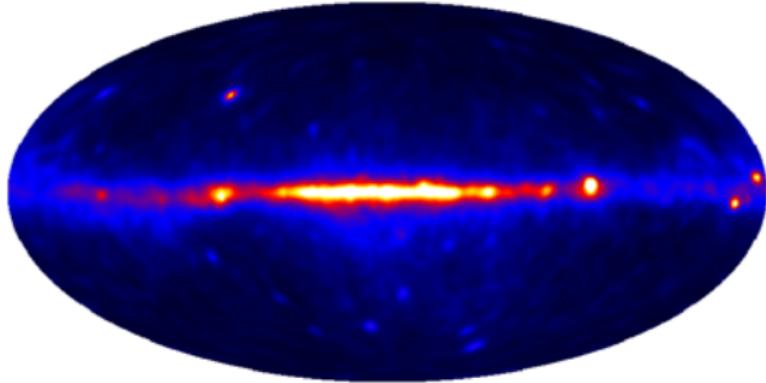
100 - 158 MeV

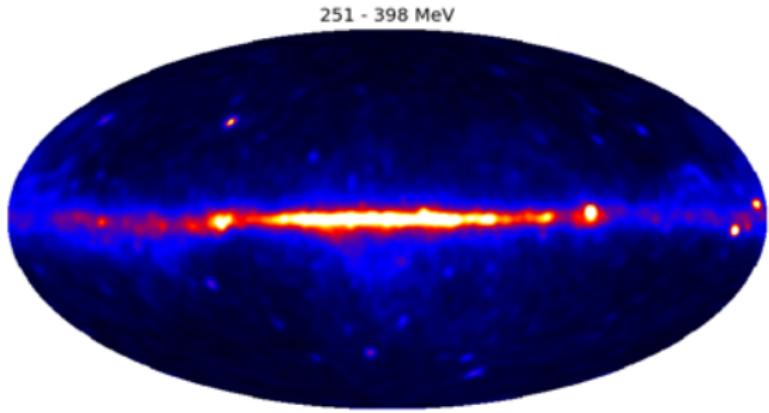
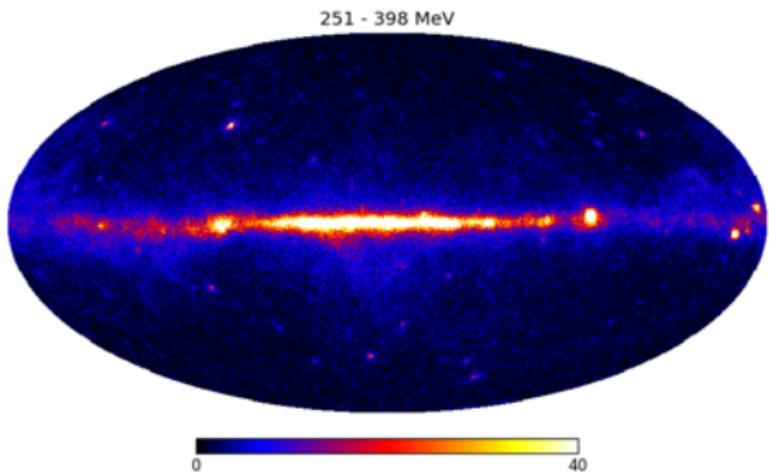


158 - 251 MeV

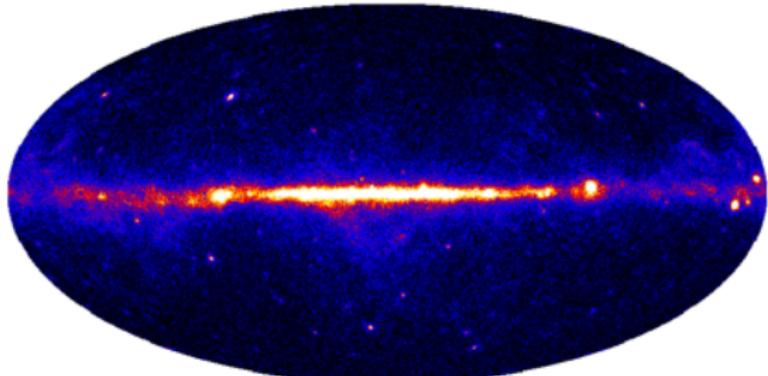


158 - 251 MeV

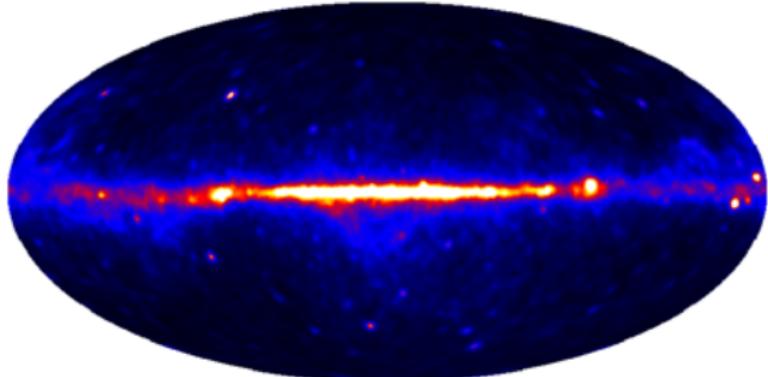




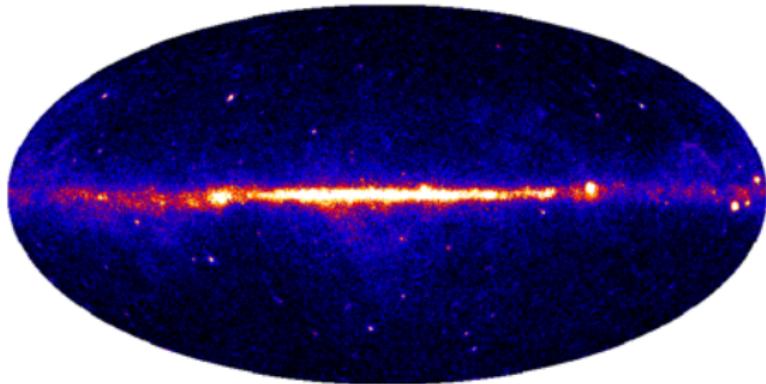
398 - 631 MeV



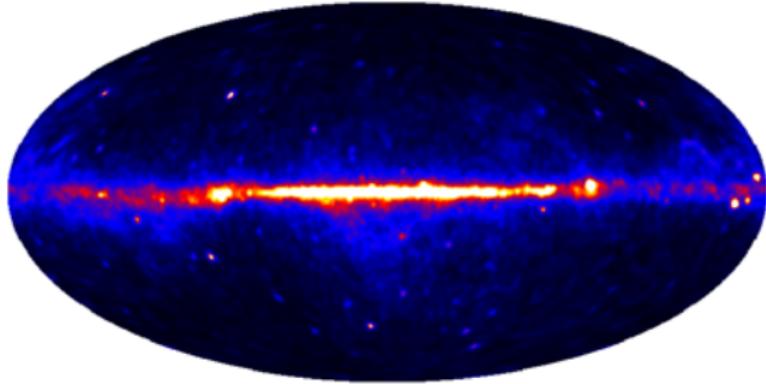
398 - 631 MeV



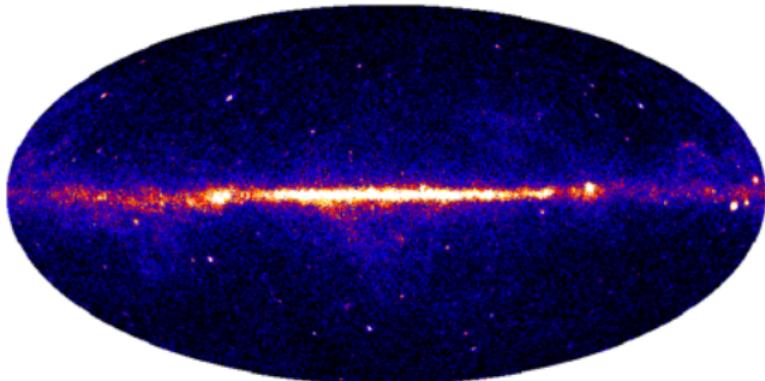
631 - 1000 MeV



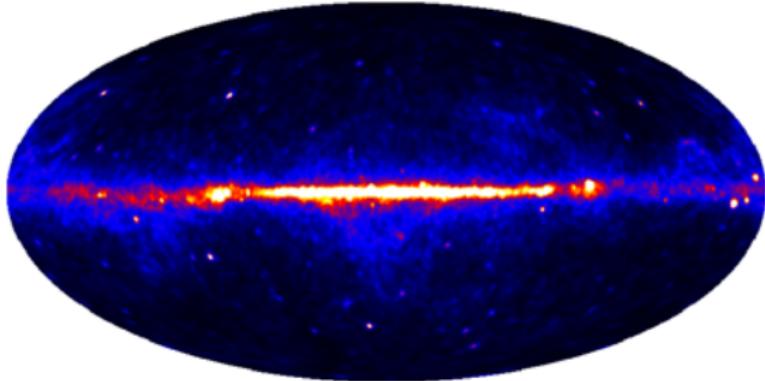
631 - 1000 MeV



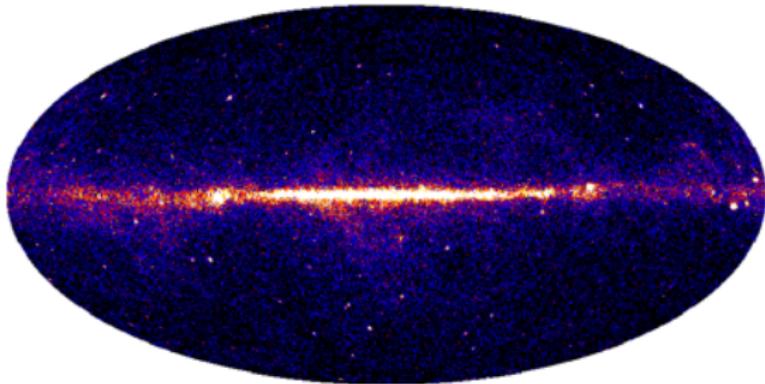
1.0 - 1.6 GeV



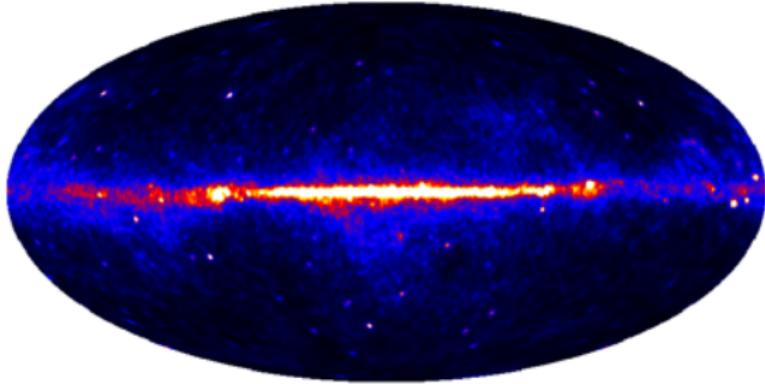
1.0 - 1.6 GeV



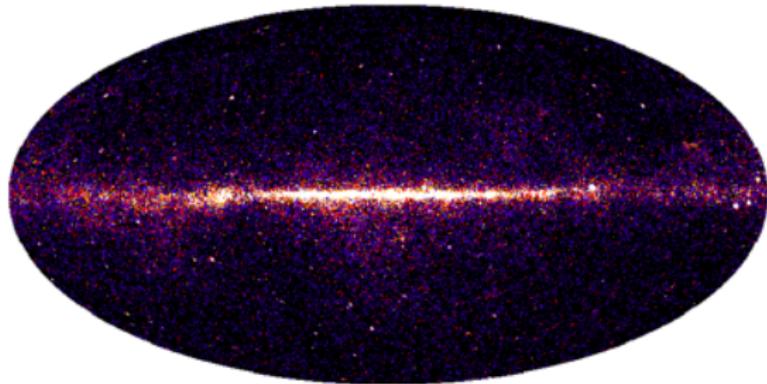
1.6 - 2.5 GeV



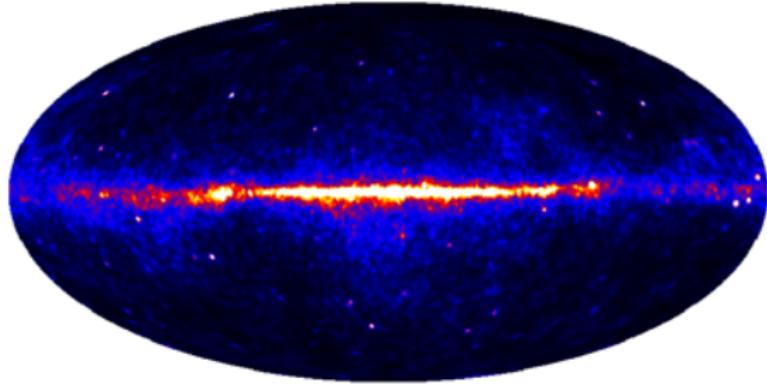
1.6 - 2.5 GeV



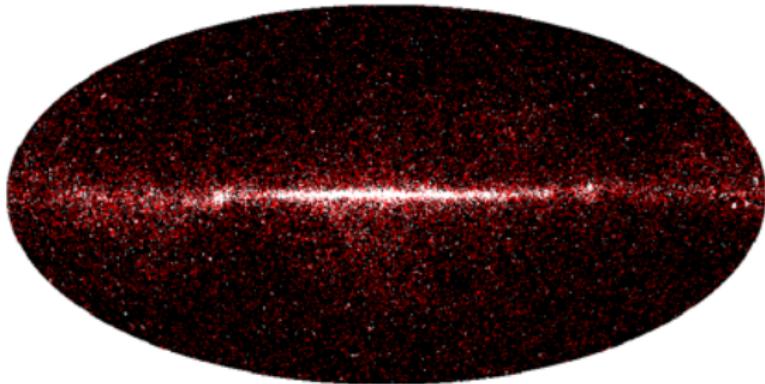
2.5 - 4.0 GeV



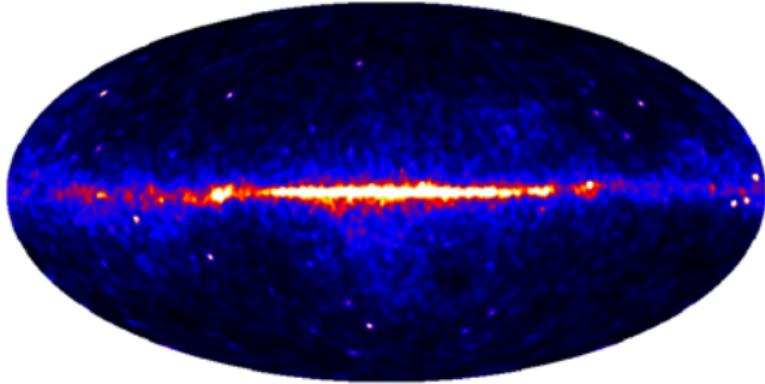
2.5 - 4.0 GeV



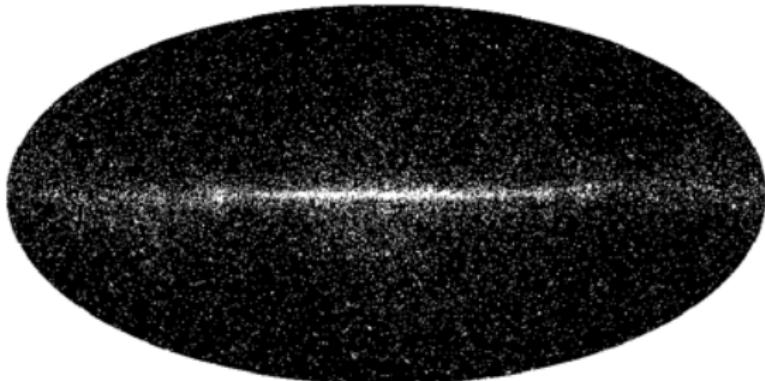
4.0 - 6.3 GeV



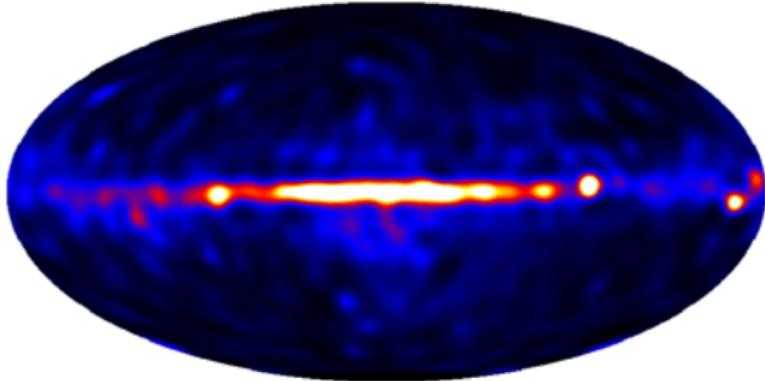
4.0 - 6.3 GeV



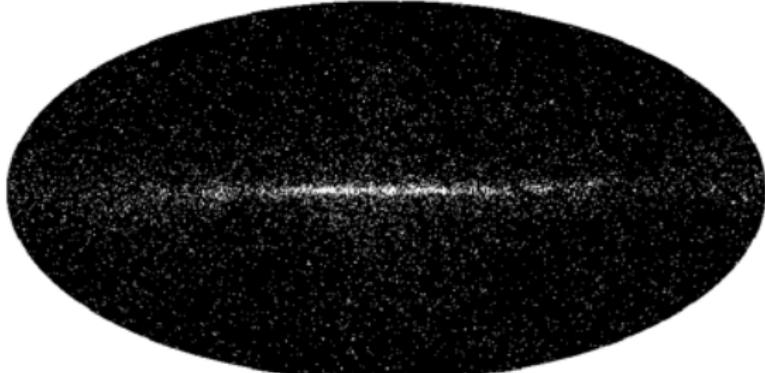
6.3 - 10 GeV



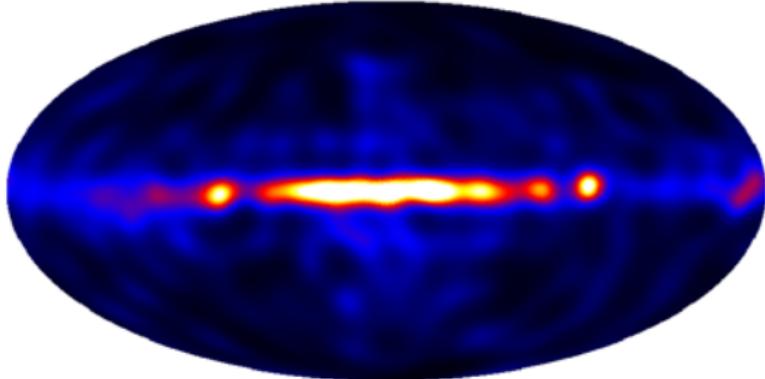
6.3 - 10 GeV



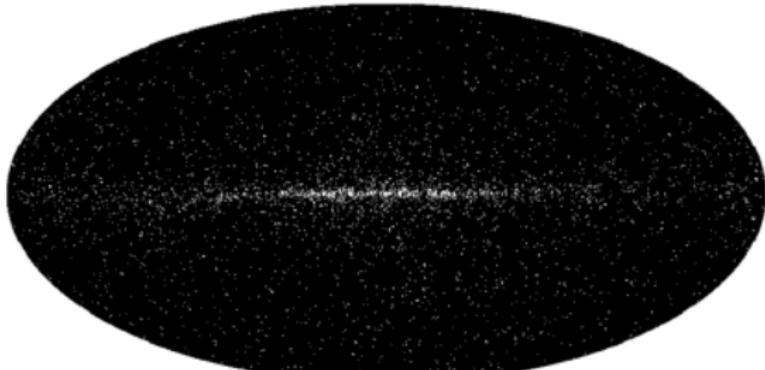
10 - 16 GeV



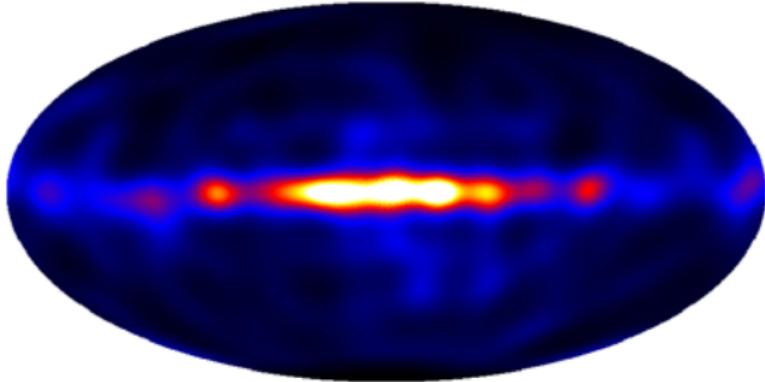
10 - 16 GeV



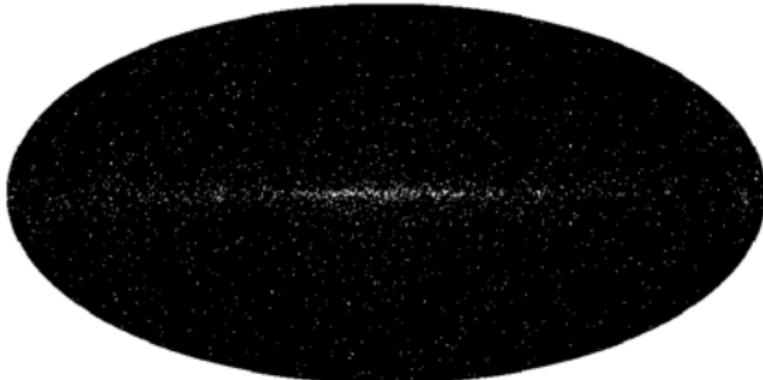
16 - 25 GeV



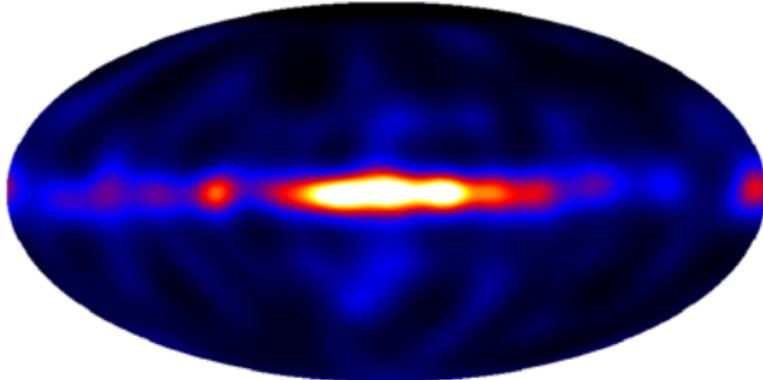
16 - 25 GeV



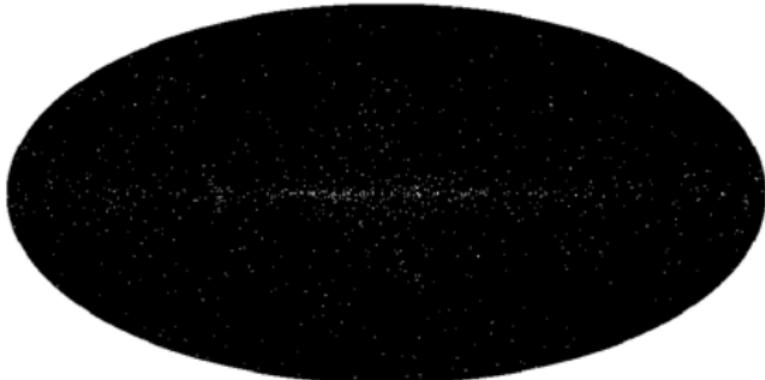
25 - 40 GeV



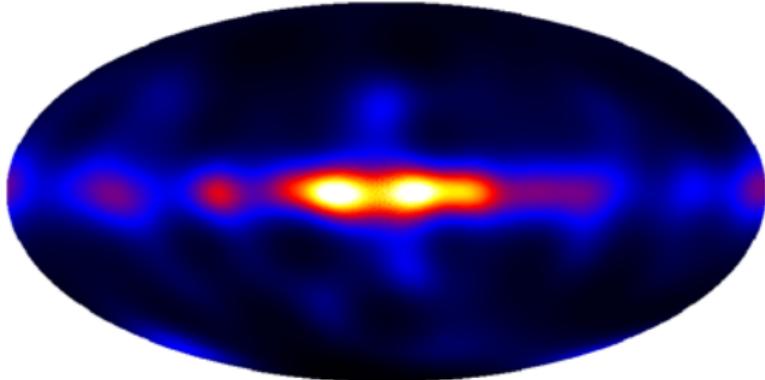
25 - 40 GeV



40 - 63 GeV



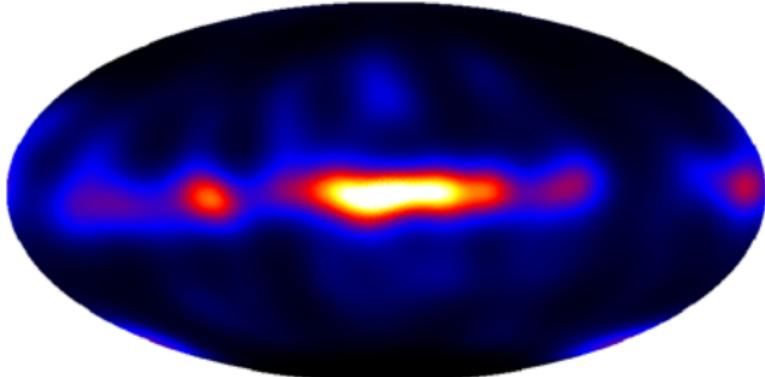
40 - 63 GeV

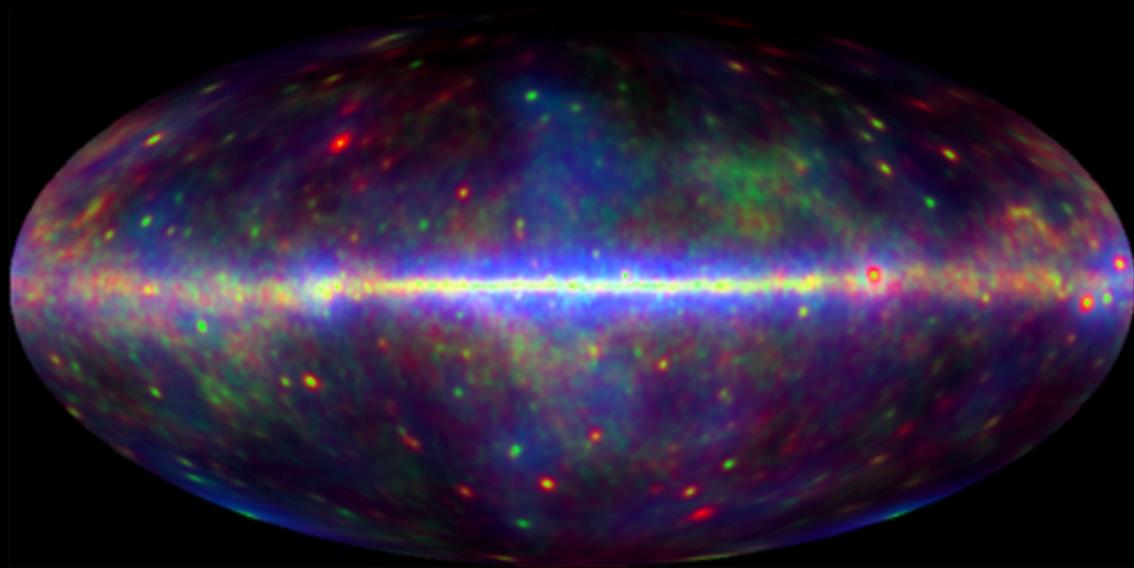


63 - 100 GeV



63 - 100 GeV







Faraday rotation

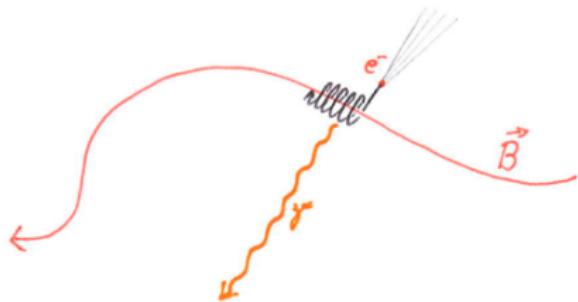
Gamma rays



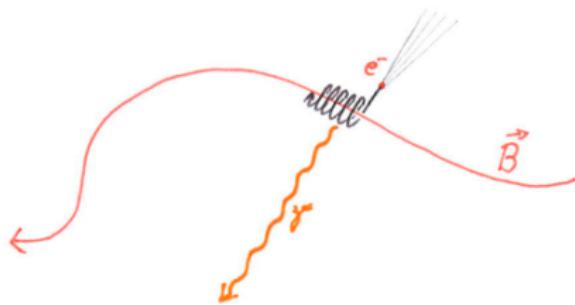
dust

CMB foregrounds

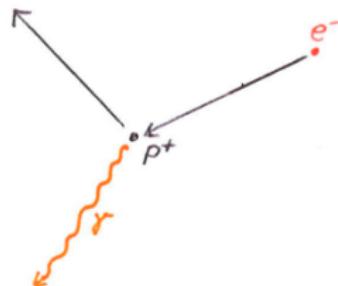
Synchrotron radiation



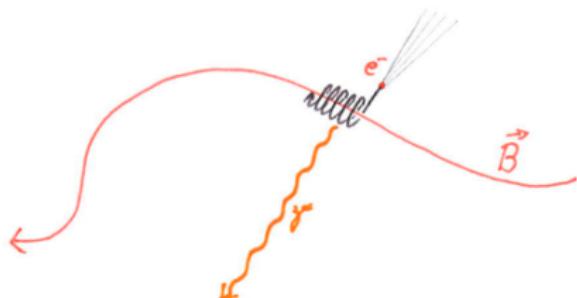
Synchrotron radiation



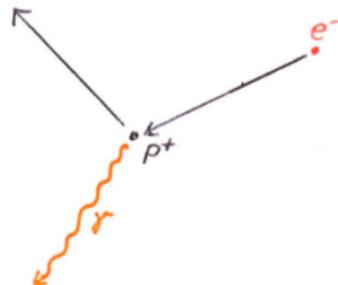
Bremsstrahlung (free-free)



Synchrotron radiation



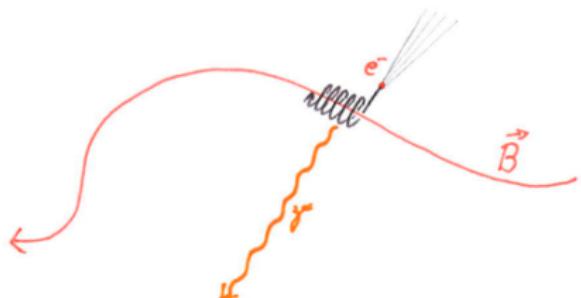
Bremsstrahlung (free-free)



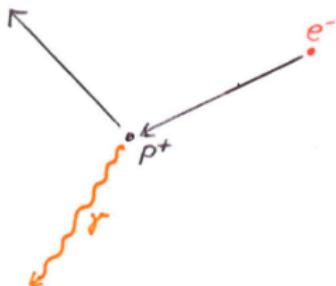
Thermal dust radiation



Synchrotron radiation



Bremsstrahlung (free-free)



Thermal dust radiation



Radiation from rotating dust grains



signal:

data:

- ▶ measurements at different frequencies
- ▶ inhomogeneous noise

response:

- ▶ mixing matrix according to frequency spectra of components

- ▶ different emission mechanisms
- ▶ Gaussian (CMB) and log-normal (foregrounds)
- ▶ cross-correlated

$$\begin{pmatrix} d_1 \\ d_2 \\ d_3 \\ d_4 \\ d_5 \end{pmatrix} = \begin{pmatrix} f_{\text{CMB}}^{\nu_1} & f_{\text{synch}}^{\nu_1} & f_{\text{ff}}^{\nu_1} & f_{\text{dust}}^{\nu_1} \\ f_{\text{CMB}}^{\nu_2} & f_{\text{synch}}^{\nu_2} & f_{\text{ff}}^{\nu_2} & f_{\text{dust}}^{\nu_2} \\ f_{\text{CMB}}^{\nu_3} & f_{\text{synch}}^{\nu_3} & f_{\text{ff}}^{\nu_3} & f_{\text{dust}}^{\nu_3} \\ f_{\text{CMB}}^{\nu_4} & f_{\text{synch}}^{\nu_4} & f_{\text{ff}}^{\nu_4} & f_{\text{dust}}^{\nu_4} \\ f_{\text{CMB}}^{\nu_5} & f_{\text{synch}}^{\nu_5} & f_{\text{ff}}^{\nu_5} & f_{\text{dust}}^{\nu_5} \end{pmatrix} \begin{pmatrix} s_{\text{CMB}} \\ s_{\text{synch}} \\ s_{\text{ff}} \\ s_{\text{dust}} \end{pmatrix} + \begin{pmatrix} n_1 \\ n_2 \\ n_3 \\ n_4 \\ n_5 \end{pmatrix}$$

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- ▶ inhomogeneous noise

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1. Determine mixing matrix.
2. “Invert” equation.

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- ▶ measurements at different frequencies
- ▶ inhomogeneous noise

response:

- ▶ mixing matrix according to frequency spectra of components

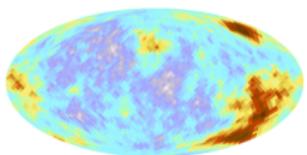
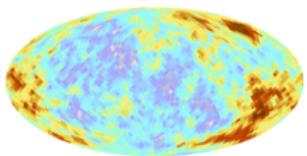
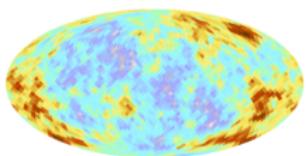
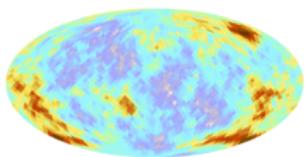
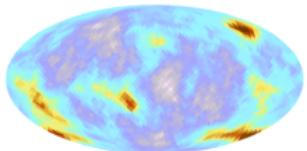
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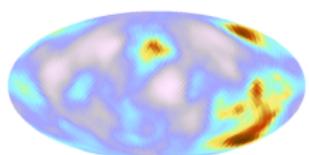
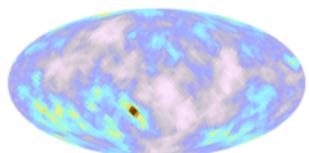
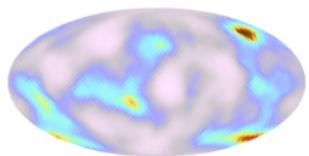
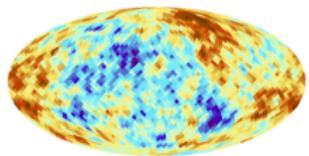
$$\begin{pmatrix} d_1 \\ d_2 \\ d_3 \\ d_4 \\ d_5 \end{pmatrix} = \begin{pmatrix} f_{\text{CMB}}^{\nu_1} & f_{\text{synch}}^{\nu_1} & f_{\text{ff}}^{\nu_1} & f_{\text{dust}}^{\nu_1} \\ f_{\text{CMB}}^{\nu_2} & f_{\text{synch}}^{\nu_2} & f_{\text{ff}}^{\nu_2} & f_{\text{dust}}^{\nu_2} \\ f_{\text{CMB}}^{\nu_3} & f_{\text{synch}}^{\nu_3} & f_{\text{ff}}^{\nu_3} & f_{\text{dust}}^{\nu_3} \\ f_{\text{CMB}}^{\nu_4} & f_{\text{synch}}^{\nu_4} & f_{\text{ff}}^{\nu_4} & f_{\text{dust}}^{\nu_4} \\ f_{\text{CMB}}^{\nu_5} & f_{\text{synch}}^{\nu_5} & f_{\text{ff}}^{\nu_5} & f_{\text{dust}}^{\nu_5} \end{pmatrix} \begin{pmatrix} s_{\text{CMB}} \\ s_{\text{synch}} \\ s_{\text{ff}} \\ s_{\text{dust}} \end{pmatrix} + \begin{pmatrix} n_1 \\ n_2 \\ n_3 \\ n_4 \\ n_5 \end{pmatrix}$$

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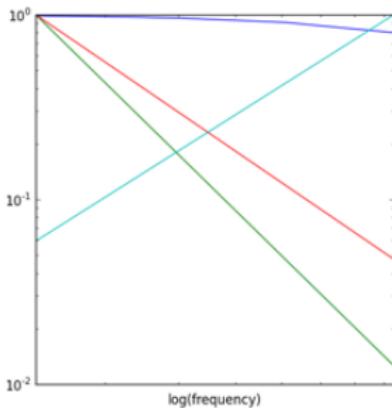
data:



signal:

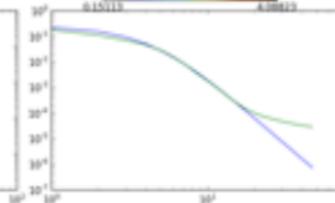
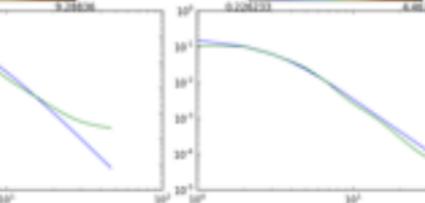
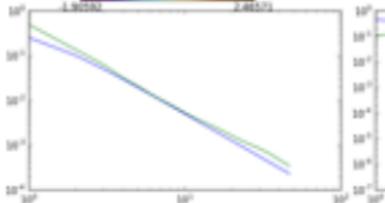
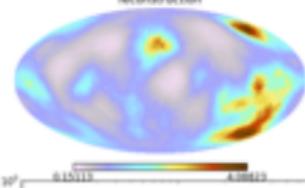
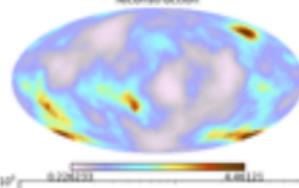
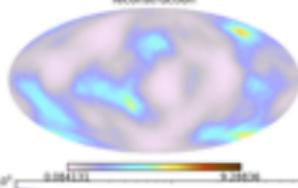
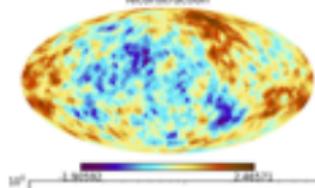
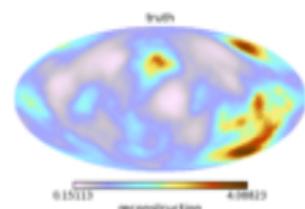
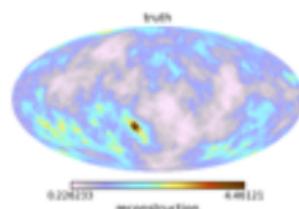
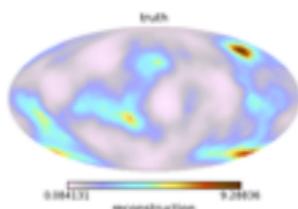
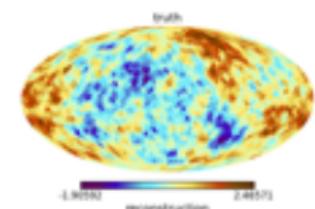


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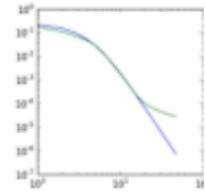
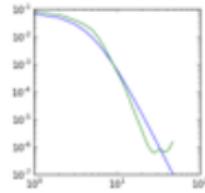
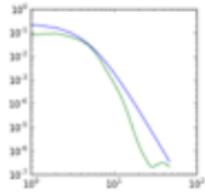
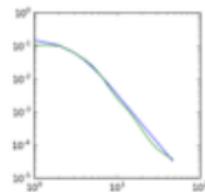
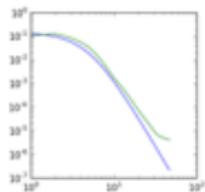
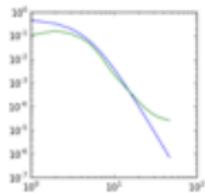




WORK IN PROGRESS

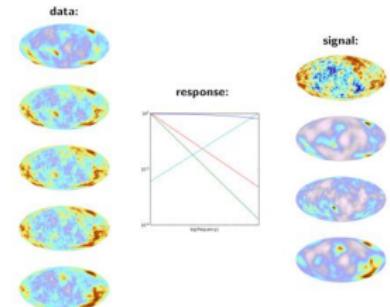
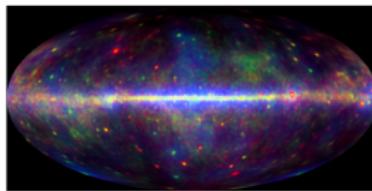
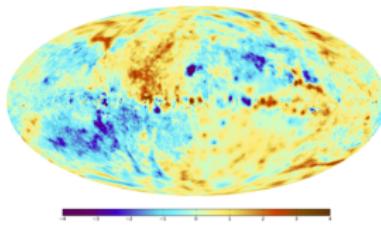


WORK IN PROGRESS

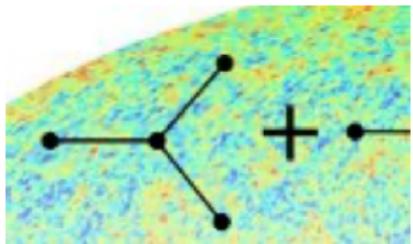


Summary

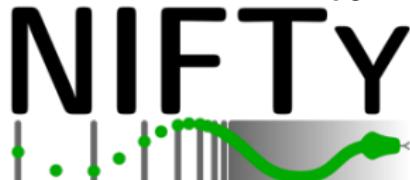
- ▶ Probabilistic inference problems
- ▶ Use correlation structure to interpolate
- ▶ Probabilistic method for dealing with outliers
- ▶ Non-linear response / Non-Gaussian signals can be dealt with



Information Field Theory



Numerical IFT for python



<http://www.mpa-garching.mpg.de/ift/>

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

- ▶ Lecture on IFT next Wednesday (July 10th)
- ▶ NIFTy tutorial next Thursday (July 11th)