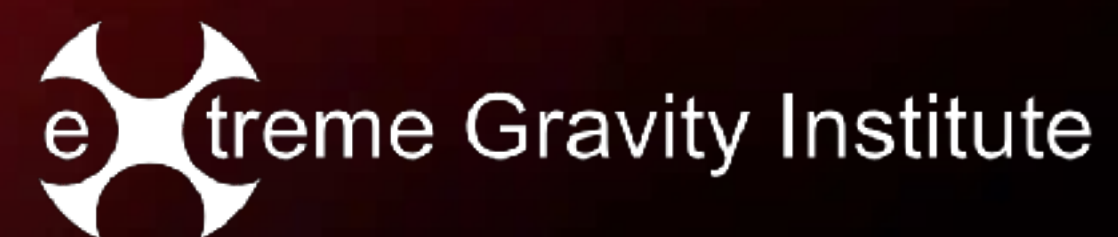


The LISA Global Fit

Neil J. Cornish & Tyson Littenberg



Because of the signal overlaps, a global fit to all the signals has to be performed

PHYSICAL REVIEW D

covering particles, fields, gravitation, and cosmology

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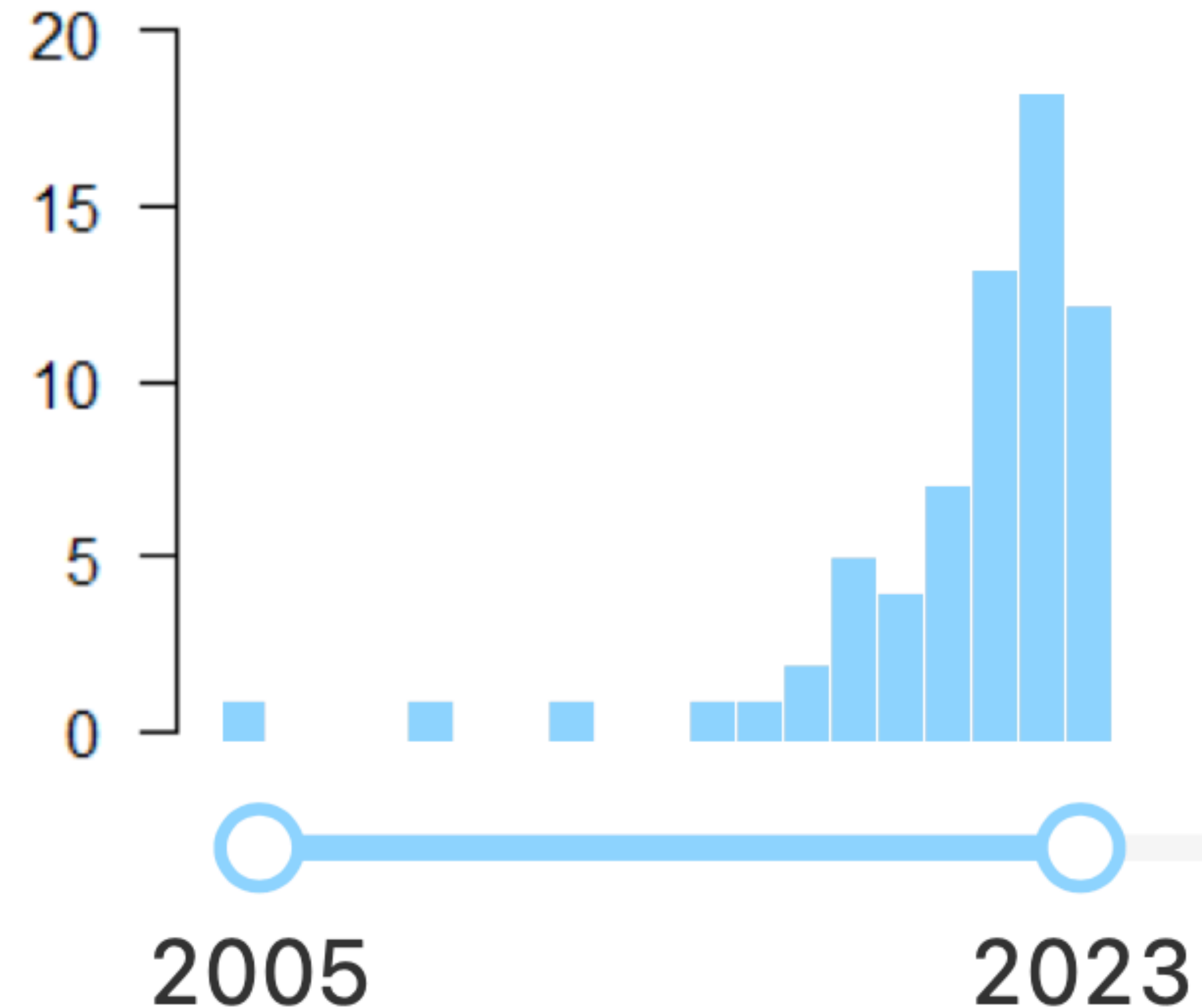
LISA data analysis using Markov chain Monte Carlo methods

Neil J. Cornish and Jeff Crowder
Phys. Rev. D **72**, 043005 – Published 22 August 2005

Article References Citing Articles (85) PDF HTML Export Citation

ABSTRACT

The Laser Interferometer Space Antenna (LISA) is expected to simultaneously detect many thousands of low-frequency gravitational wave signals. This presents a data analysis challenge that is very different to the one encountered in ground based gravitational wave astronomy. LISA data analysis requires the identification of individual signals from a data stream containing an unknown number of overlapping signals. **Because of the signal overlaps, a global fit to all the signals has to be performed** in order to avoid biasing the solution. However, performing such a global fit requires the exploration of an enormous parameter space with a dimension upwards of 50000. Markov Chain Monte Carlo (MCMC) methods offer a very promising solution to the LISA data analysis problem. MCMC algorithms

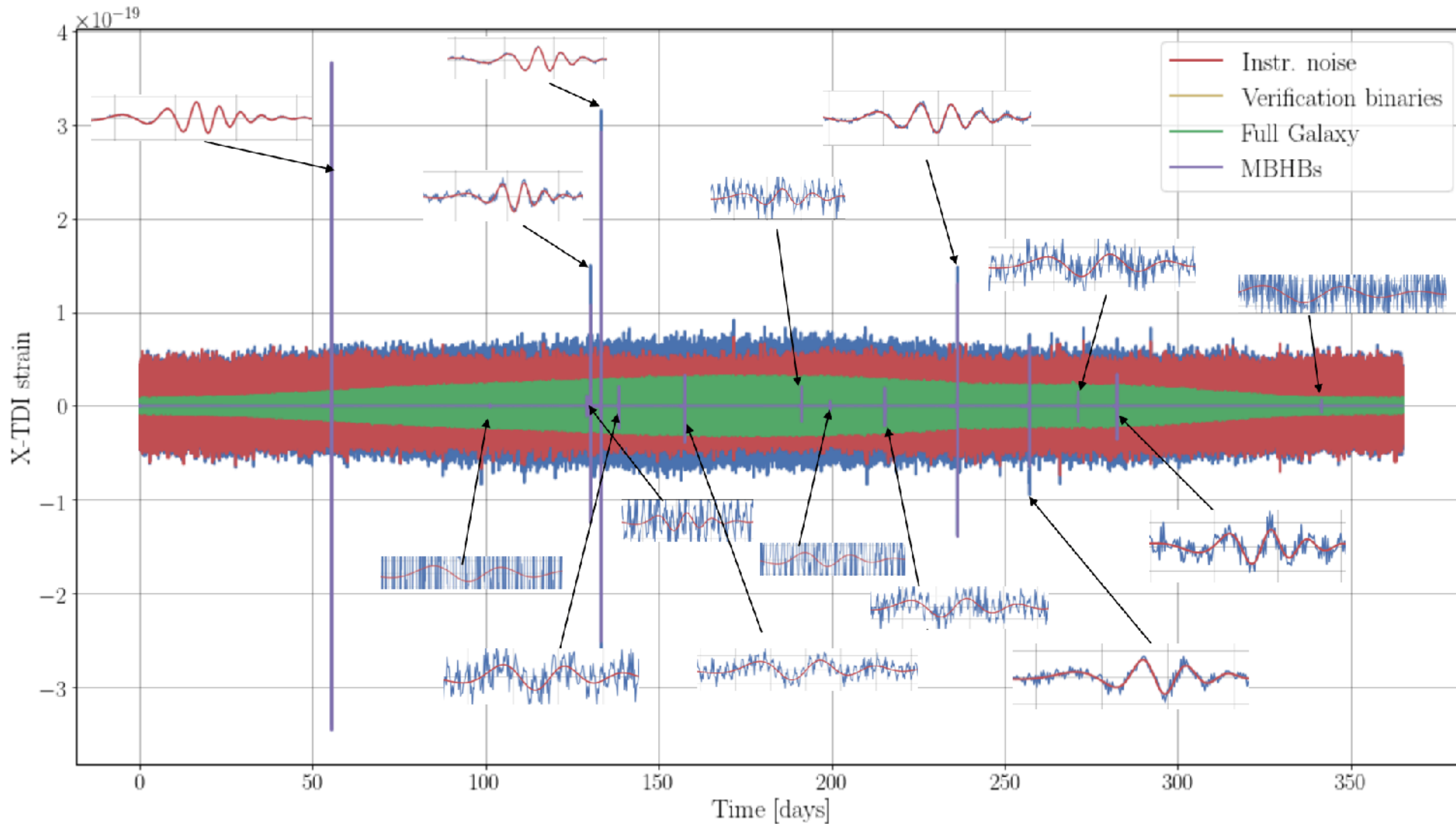


LISA is not LIGO in Space

- Millions of overlapping signals
- Unknown number of detectable sources
- Non-stationary and non-Gaussian noise
- Data gaps and disturbances
- Time varying instrument response
- Complex signals, multiple harmonics



LISA Data Challenge: Sangria Edition



The Global Solution

Likelihood function
for Gaussian noise

$$p(\mathbf{d}|\vec{\lambda}) = \frac{1}{\sqrt{(2\pi)^M \det \mathbf{C}}} e^{-\frac{1}{2}(\mathbf{d}-\mathbf{h}) \cdot \mathbf{C}^{-1} \cdot (\mathbf{d}-\mathbf{h})}$$

$$\mathbf{h} = \sum_{i=1}^N \mathbf{h}_i = \text{GW signal model}$$

N unknown, mix of signal types

\mathbf{C} = noise correlation matrix

Jointly inferred with signal model. Up to M^3 cost to invert

$\vec{\lambda}$ = model parameters

Signal and noise $\mathcal{O}(10^6)$ parameters

The Global Solution

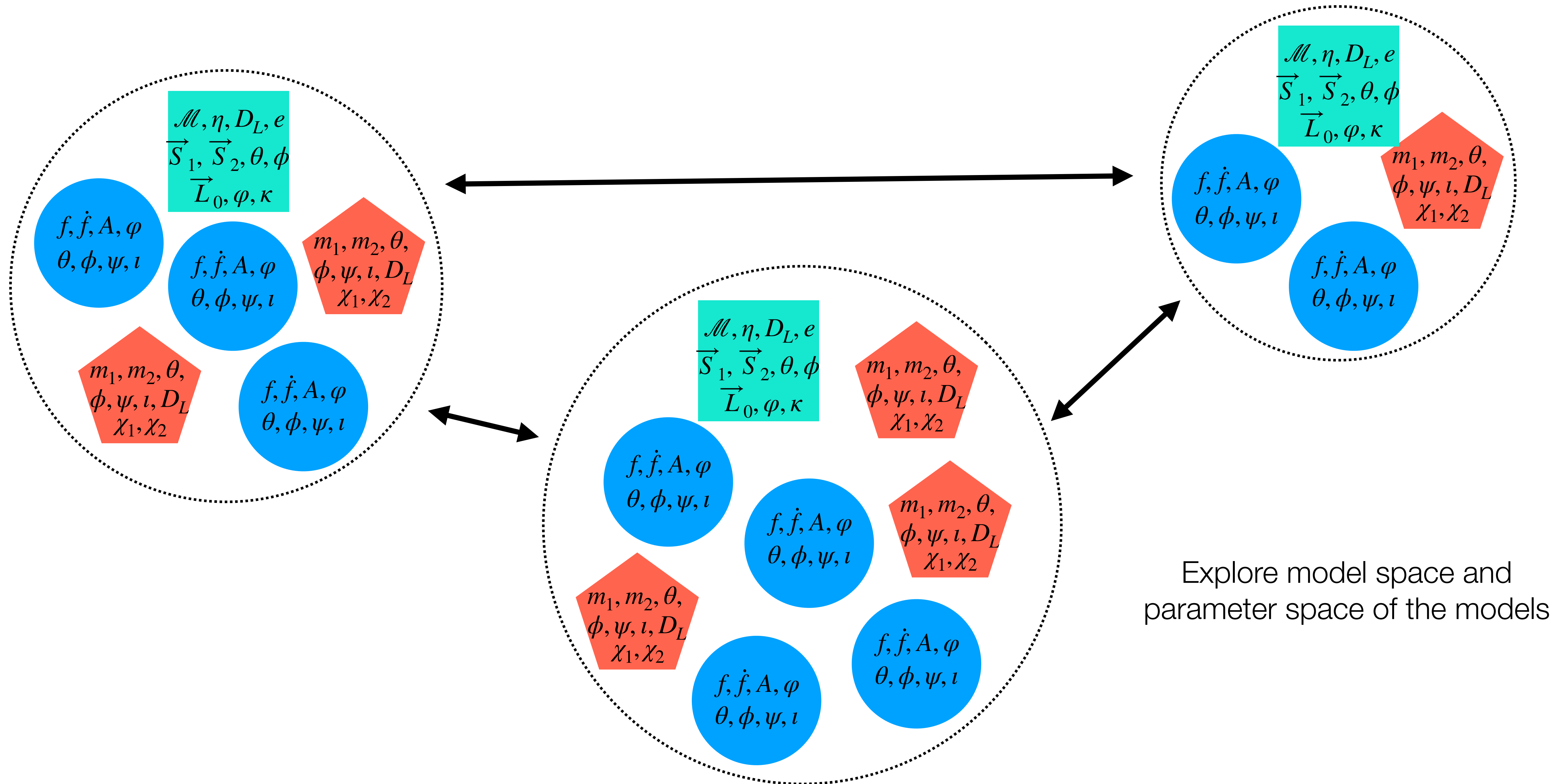
Likelihood function
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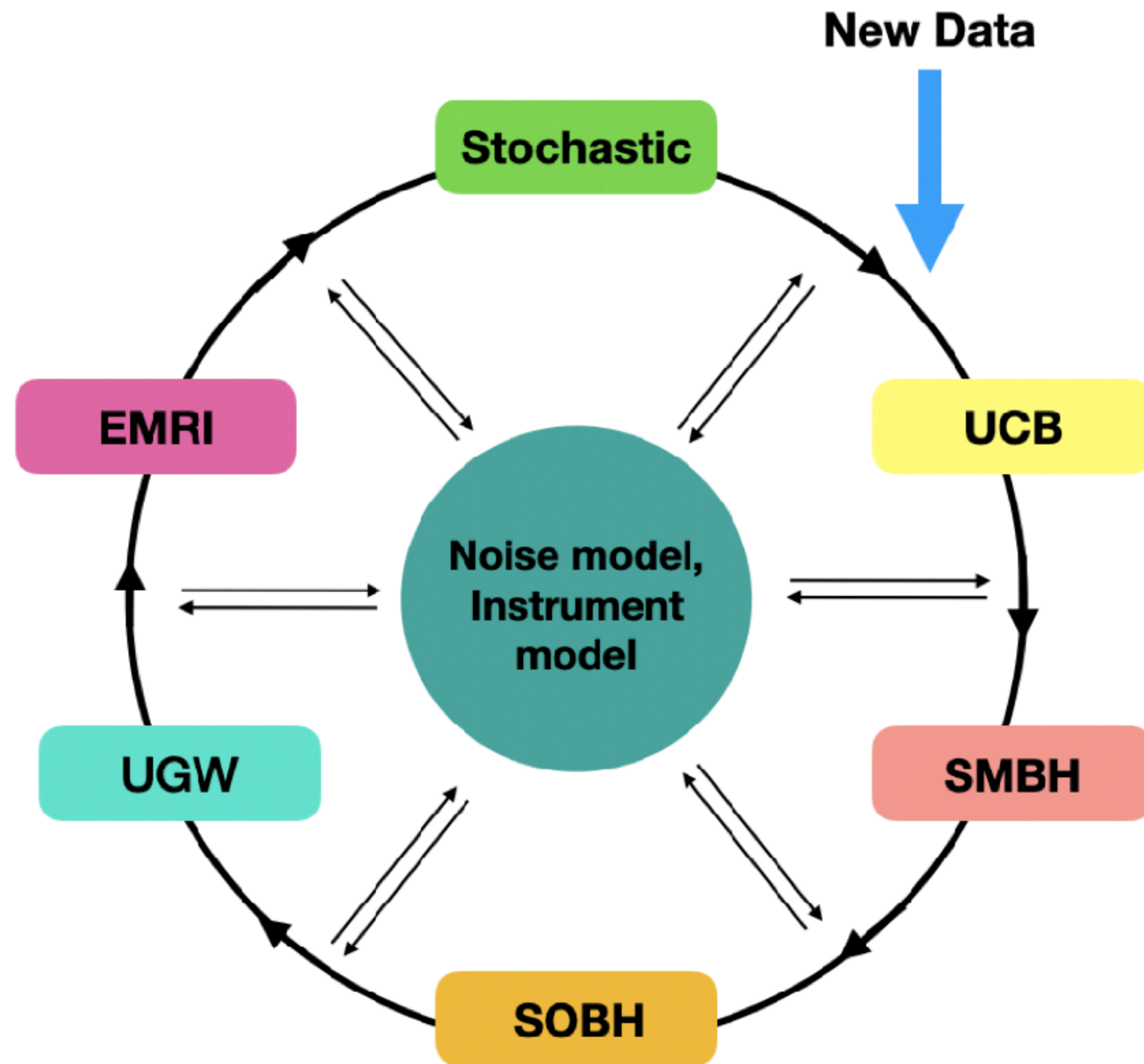
$$\log L = (d|h) - \frac{1}{2}(h|h) = \sum_i^N \log L_i - \frac{1}{2} \sum_{i \neq j} (h_i|h_j)$$

Why we need a global solution

Trans-dimensional Inference



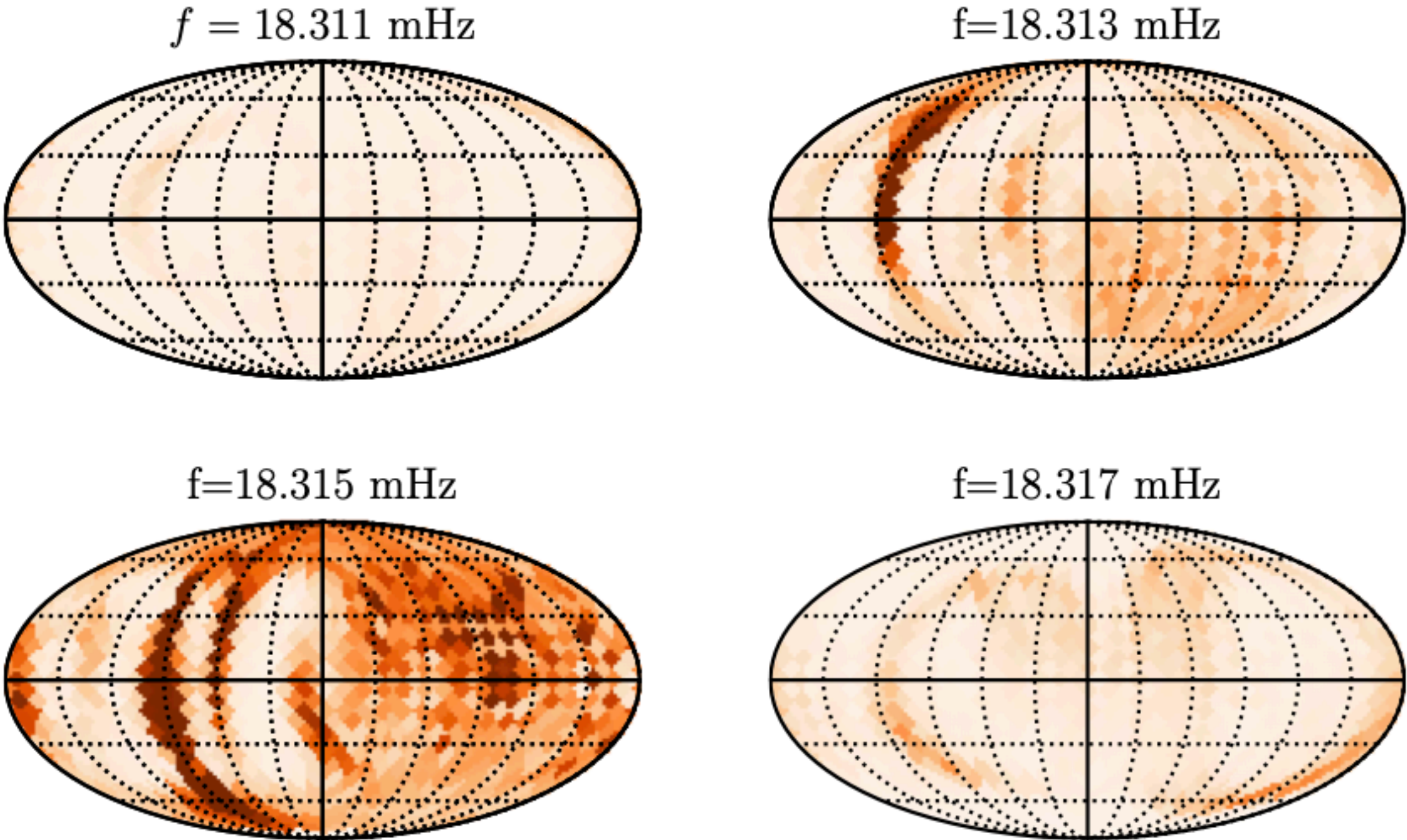
Global Fit via Blocked Sampling



- Transdimensional Markov Chain Monte Carlo (RJMCMC)
- Blocked Metropolis Hastings— update each component of the signal/noise model in circular sweeps
- Only pass residuals - decouples the analysis types
- Update the fit every ~week as new data arrives

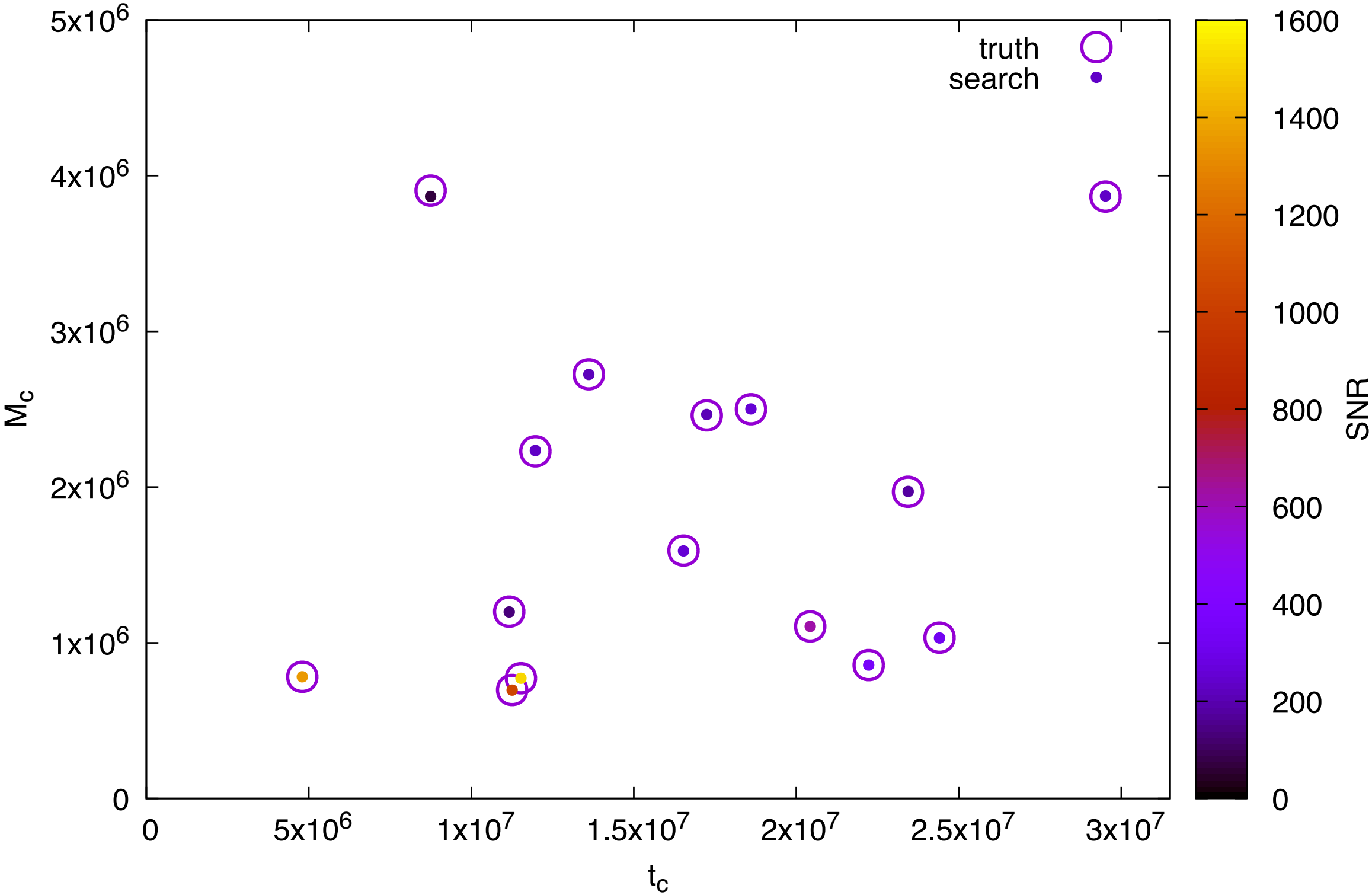
Low latency single-source search results used as proposals in global fit

F-statistic maps for GBs



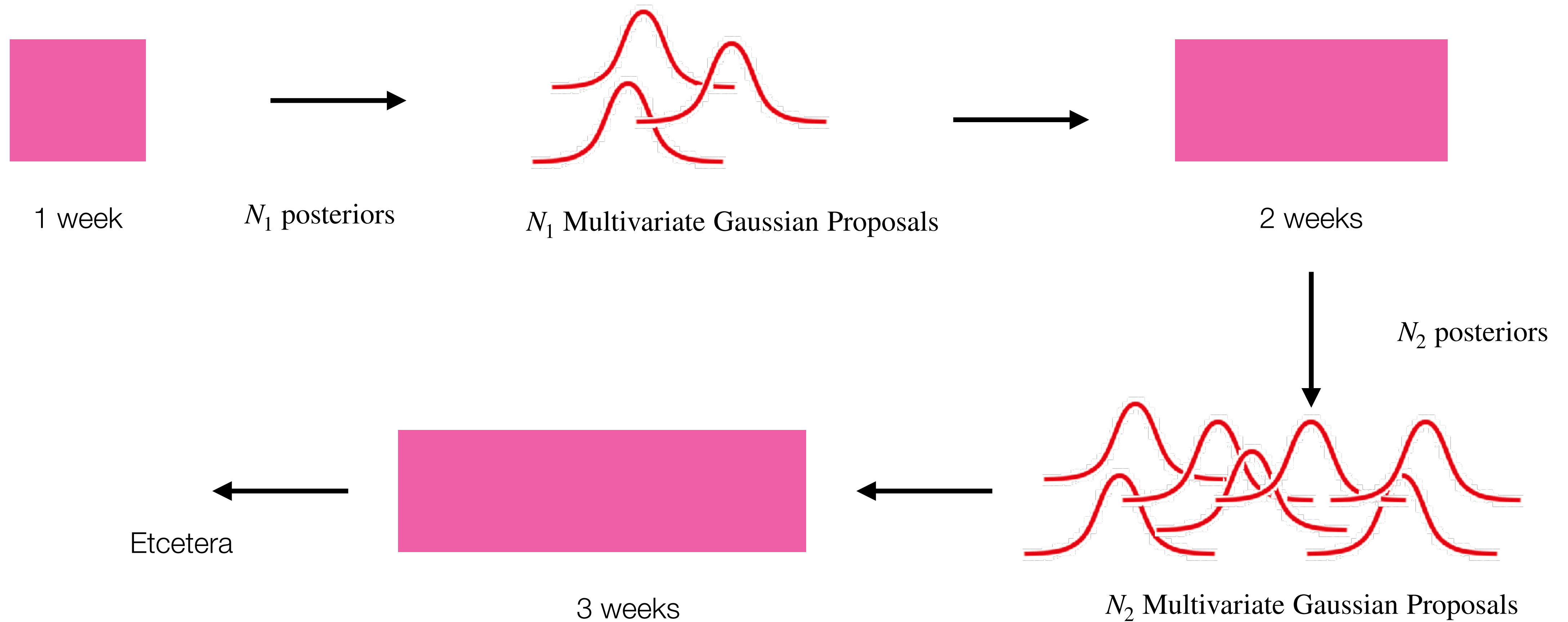
[Littenber, Cornish, Lackeos & Robson, arXiv:2004.08464]

Low latency BH search



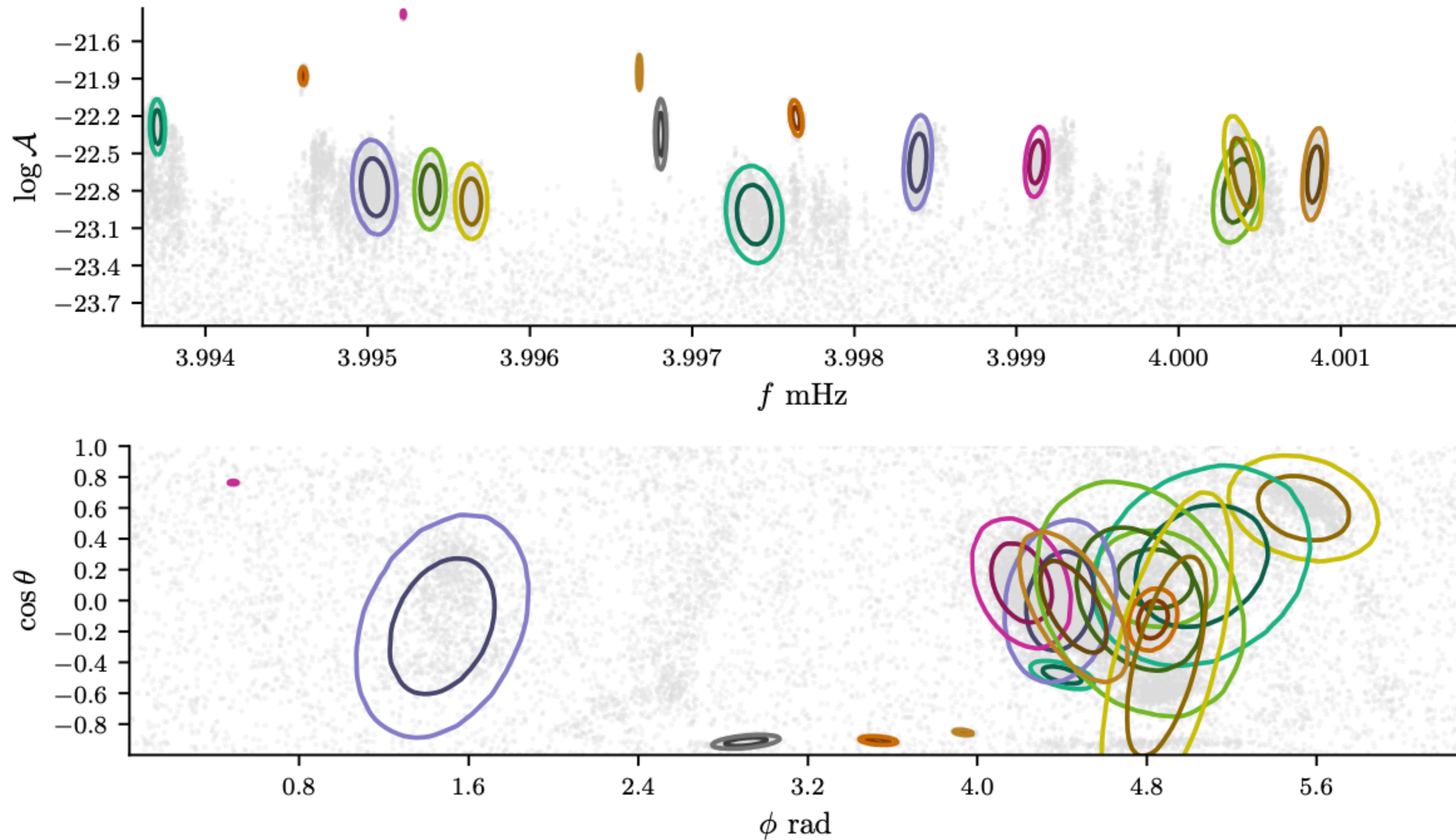
[Cornish, arXiv: 2110.06238]

Building up the solution - "time annealing"



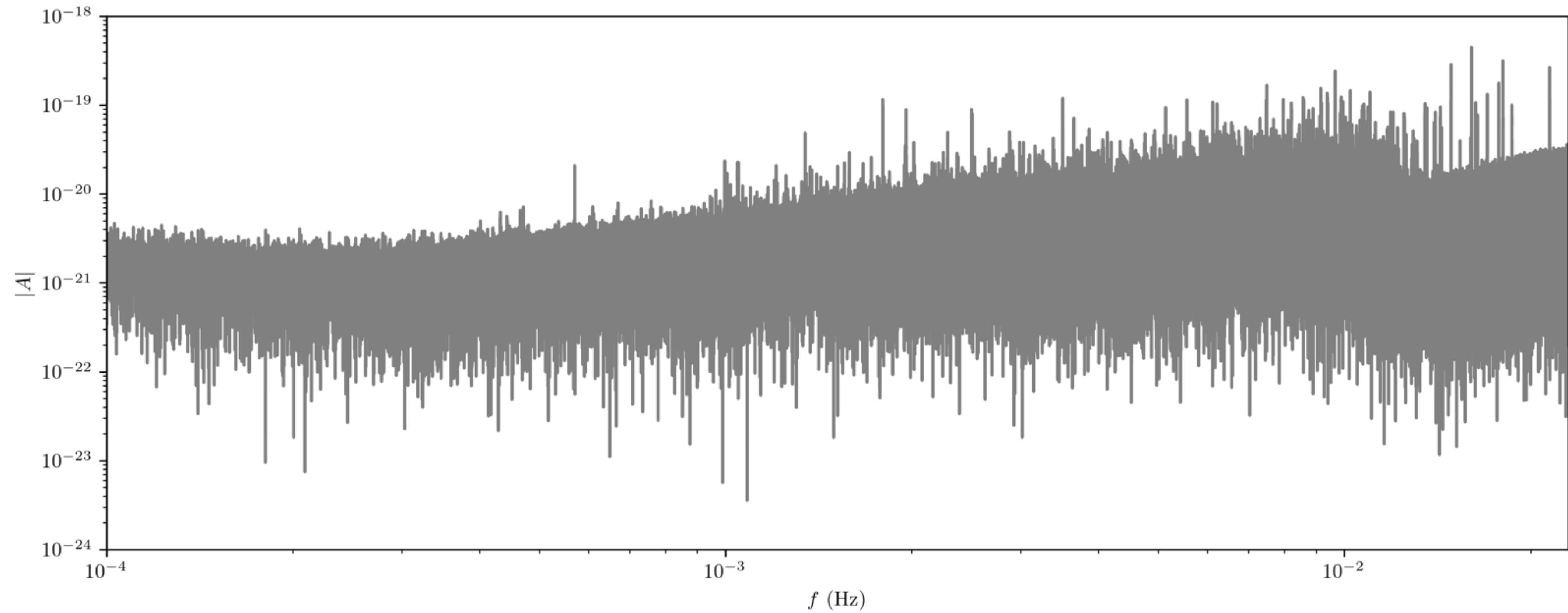
We used 1 month -> 3 months -> 6 months -> 12 months

Building up the solution - “time annealing”



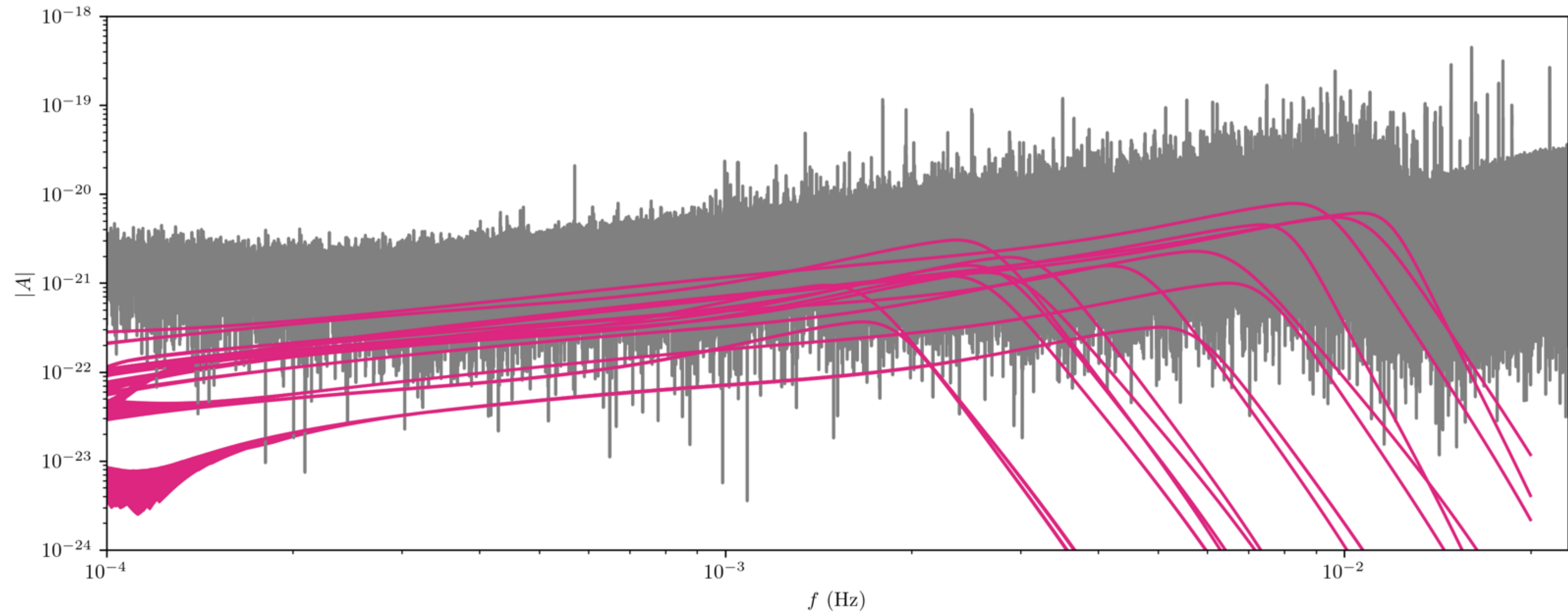
We used 1 month \rightarrow 3 months \rightarrow 6 months \rightarrow 12 months

12 months of Sangria data - A TDI channel

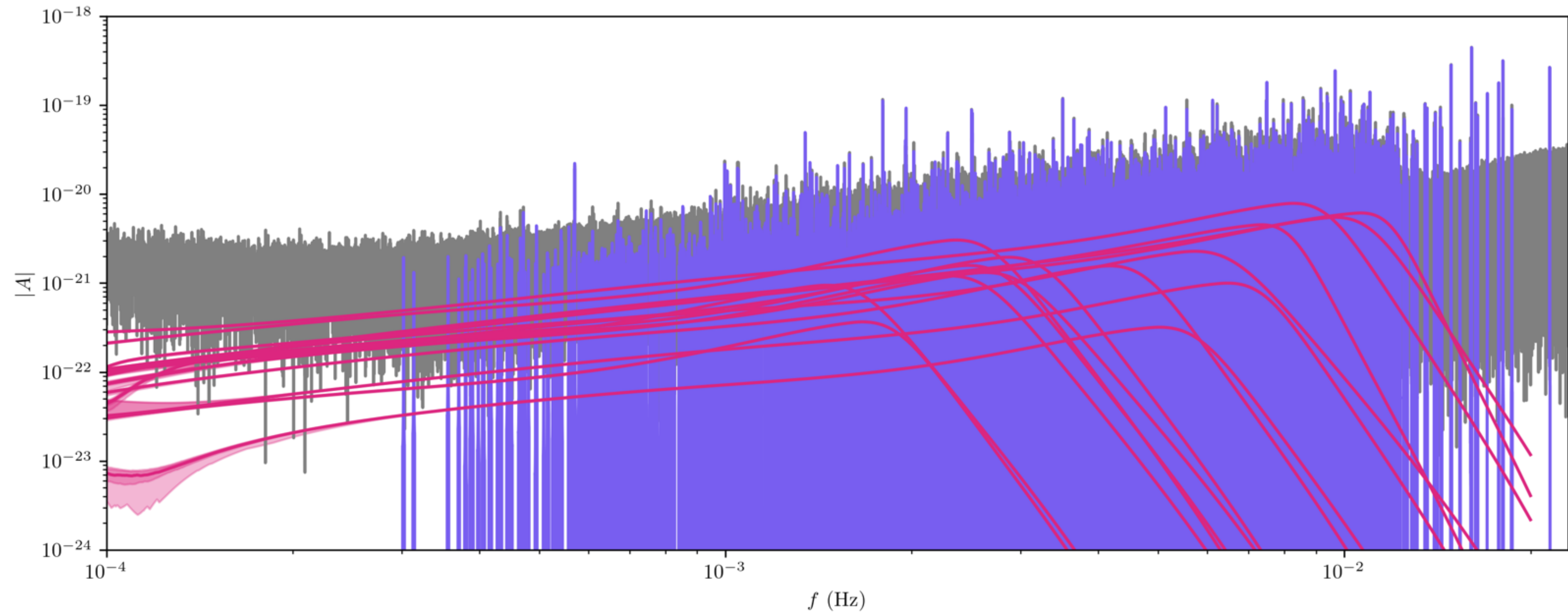


[Litttenberg & Cornish, [arXiv: 2301.03673](https://arxiv.org/abs/2301.03673)]

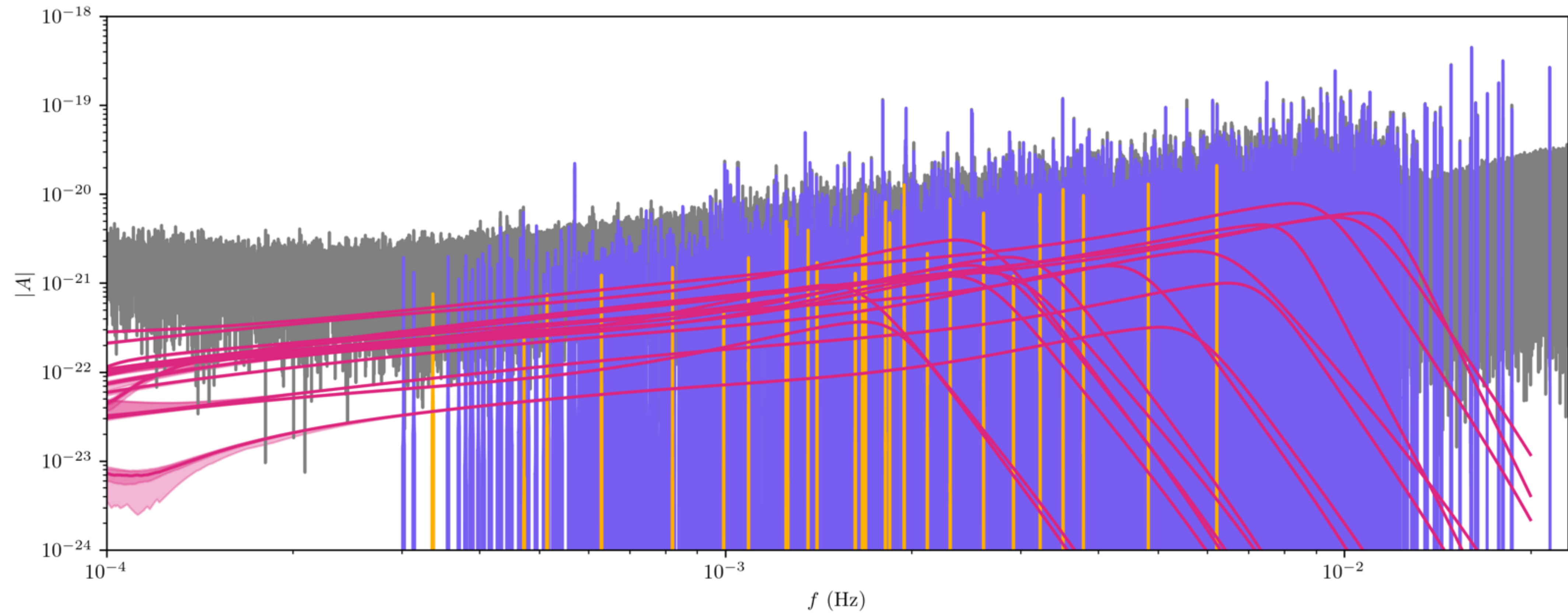
12 months of Sangria data - A TDI channel



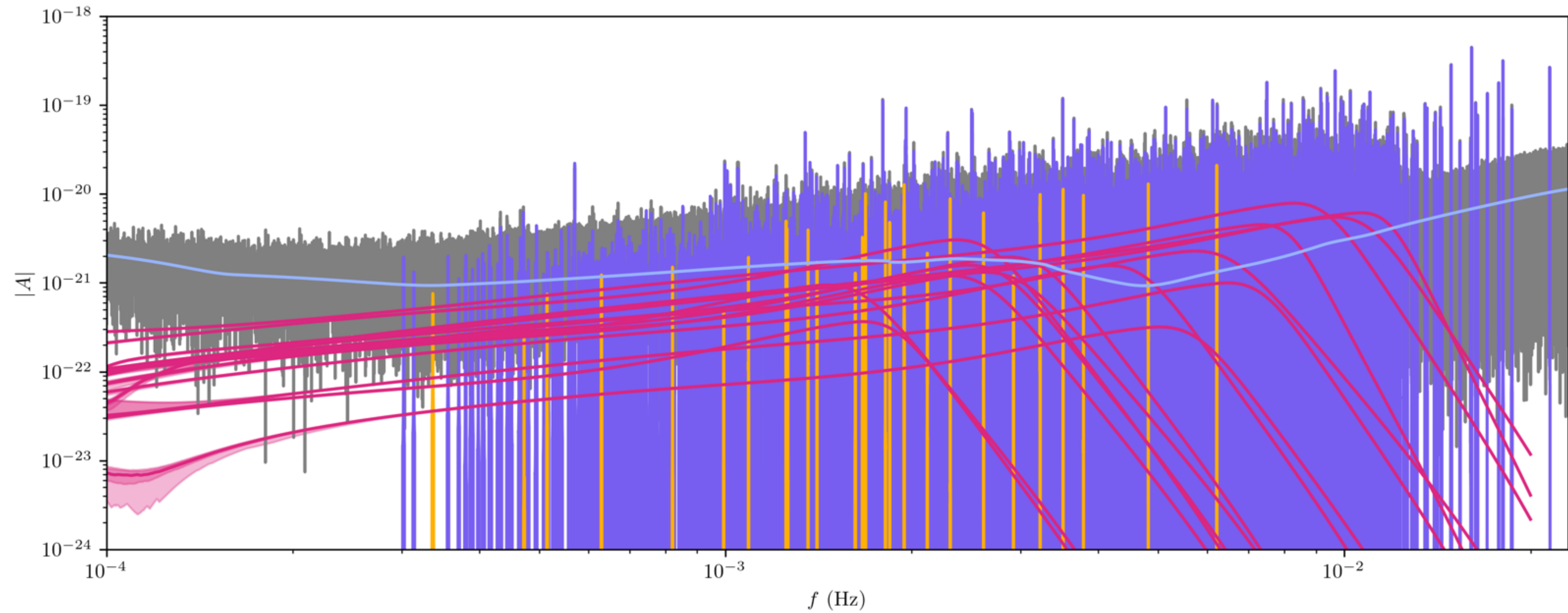
12 months of Sangria data - A TDI channel



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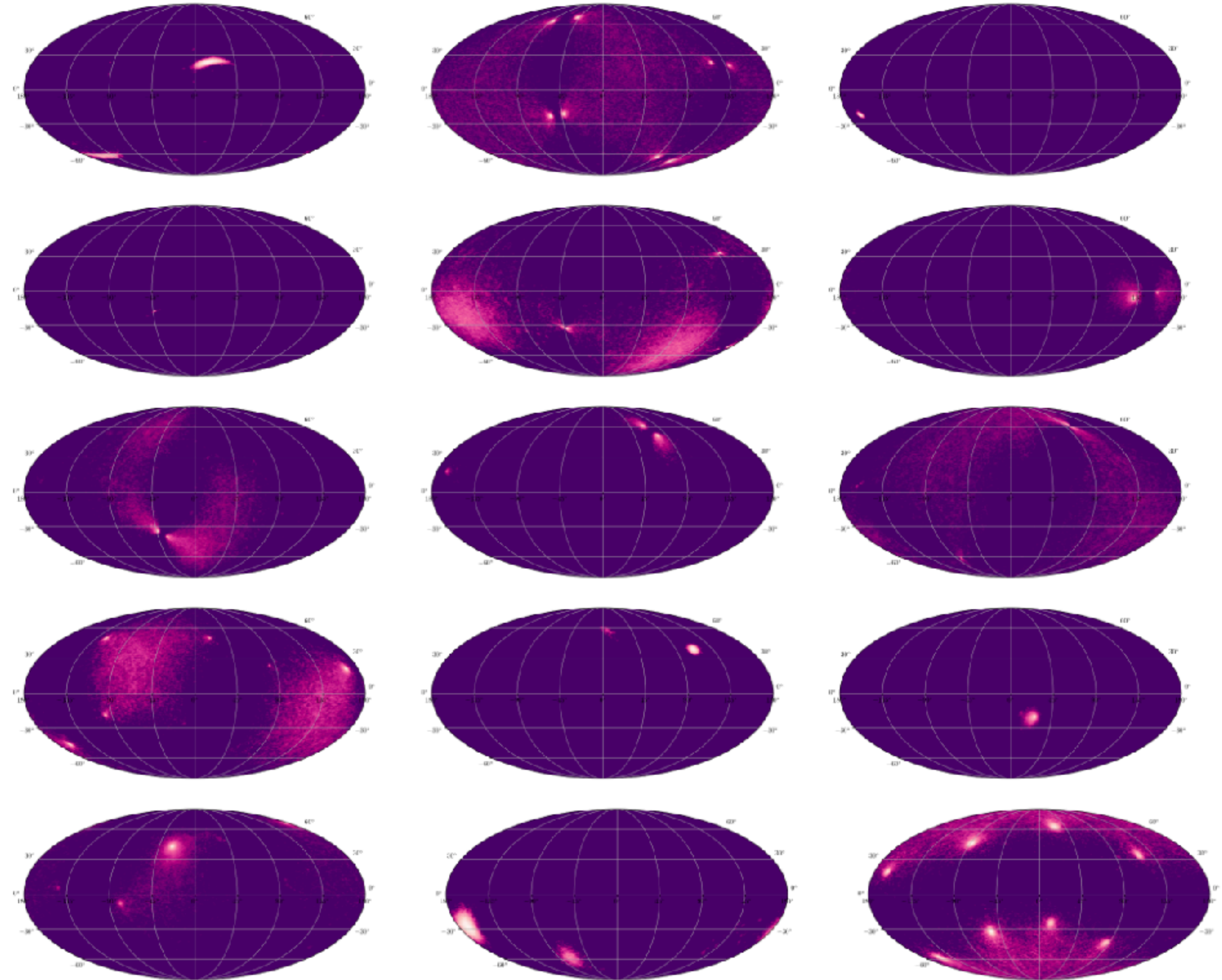
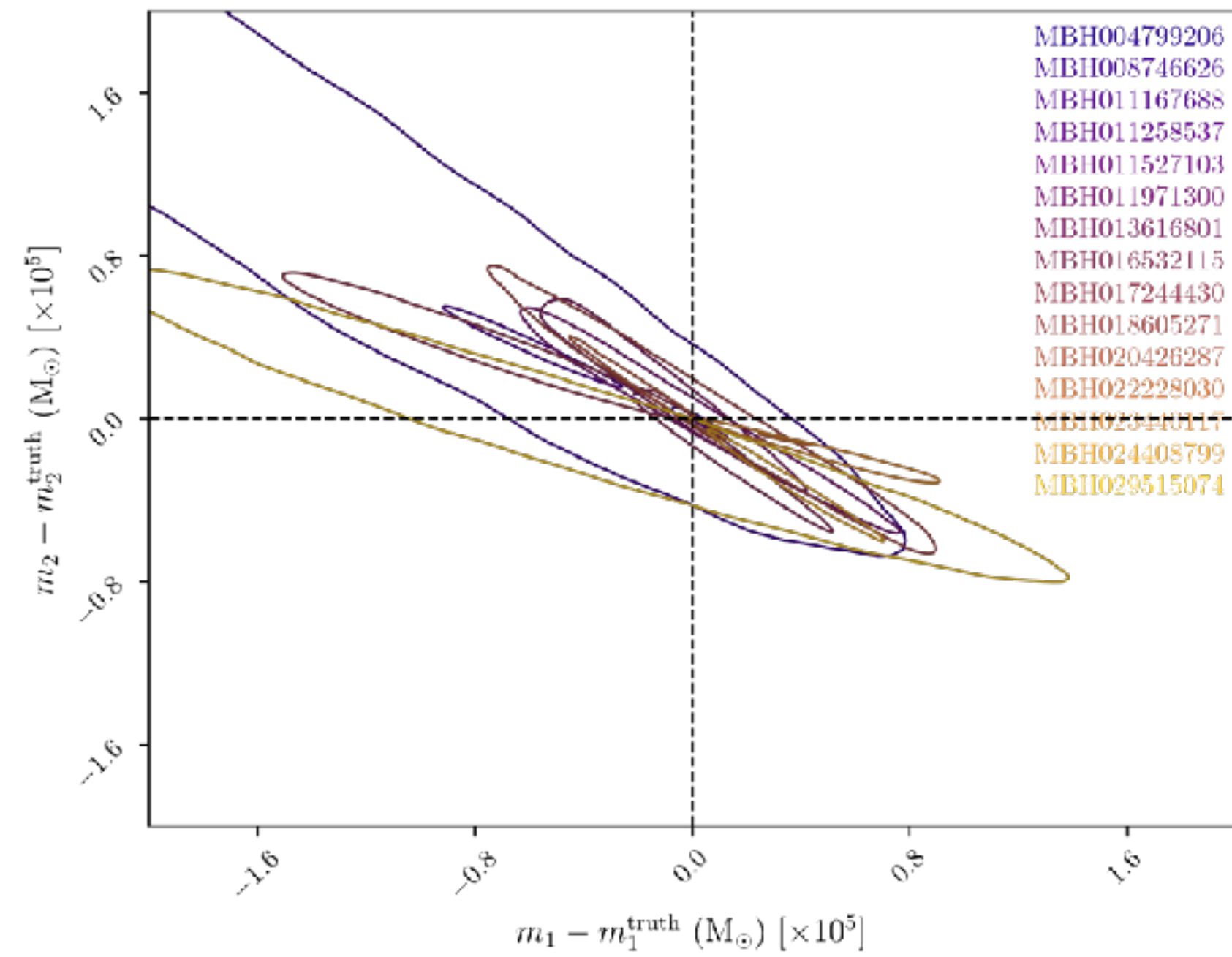
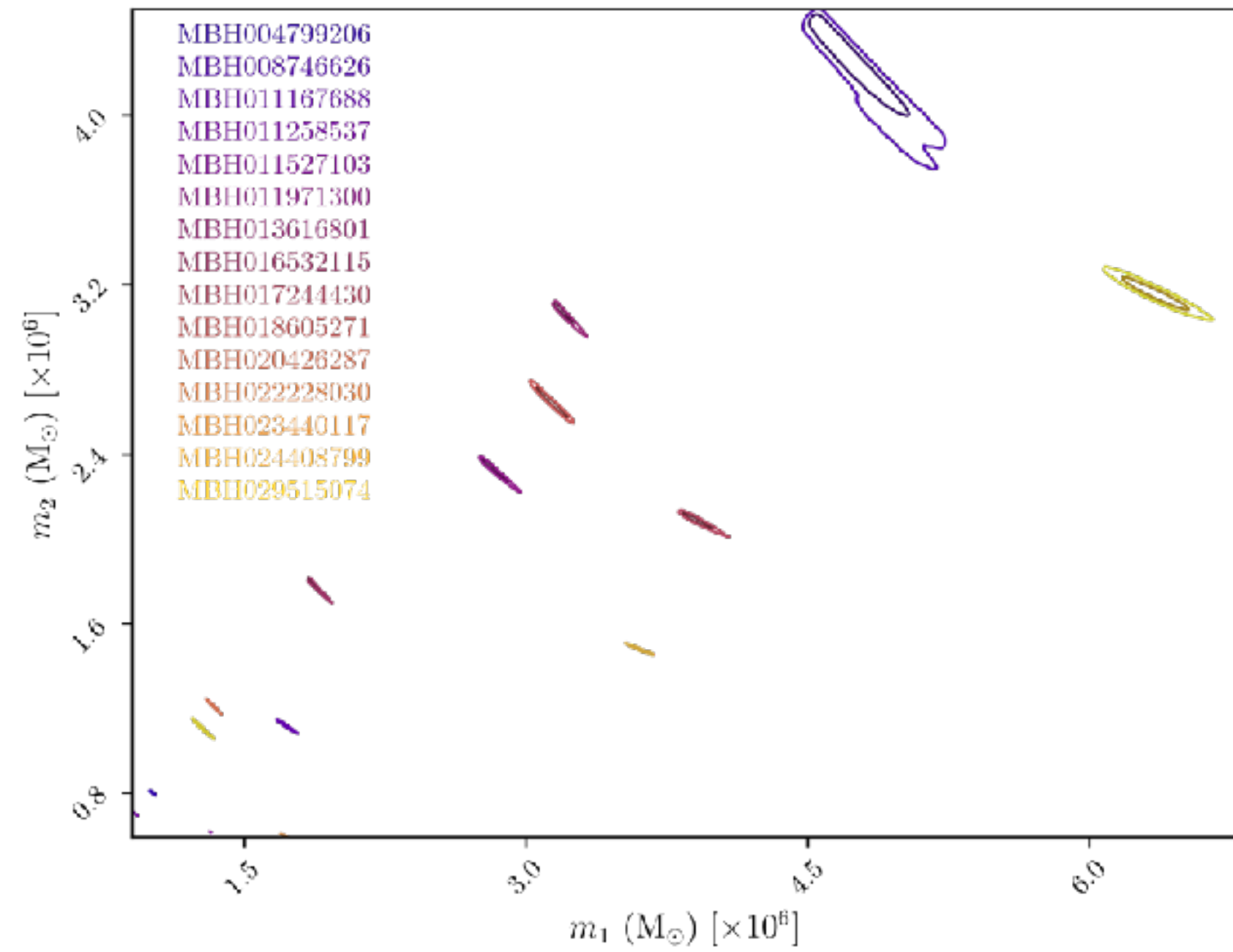


12 months of Sangria data - A TDI channel

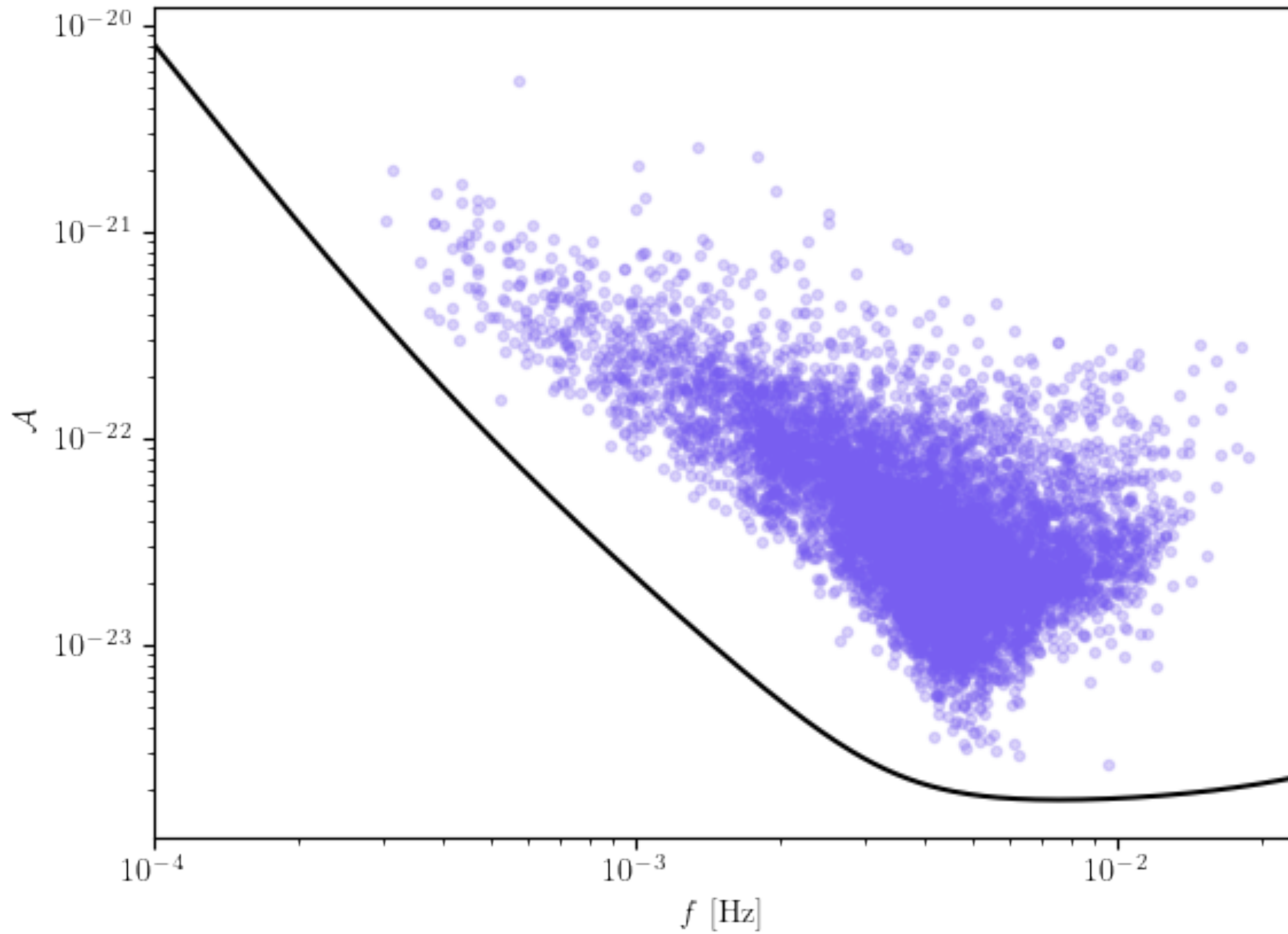


Run time ~ 2 days on AWS, \$12K

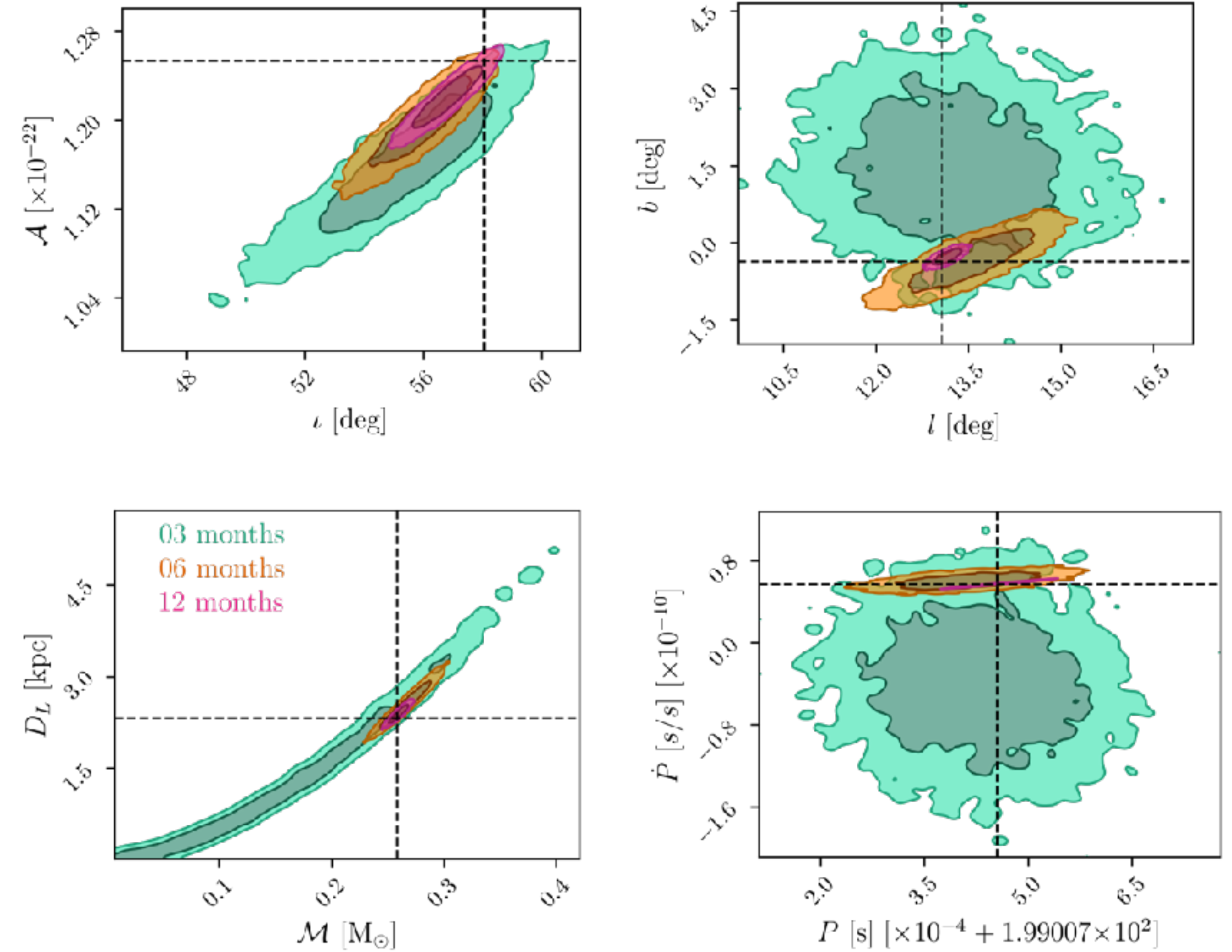
12 months of Sangria data - MBHBs



Sangria data - Galactic Binaries

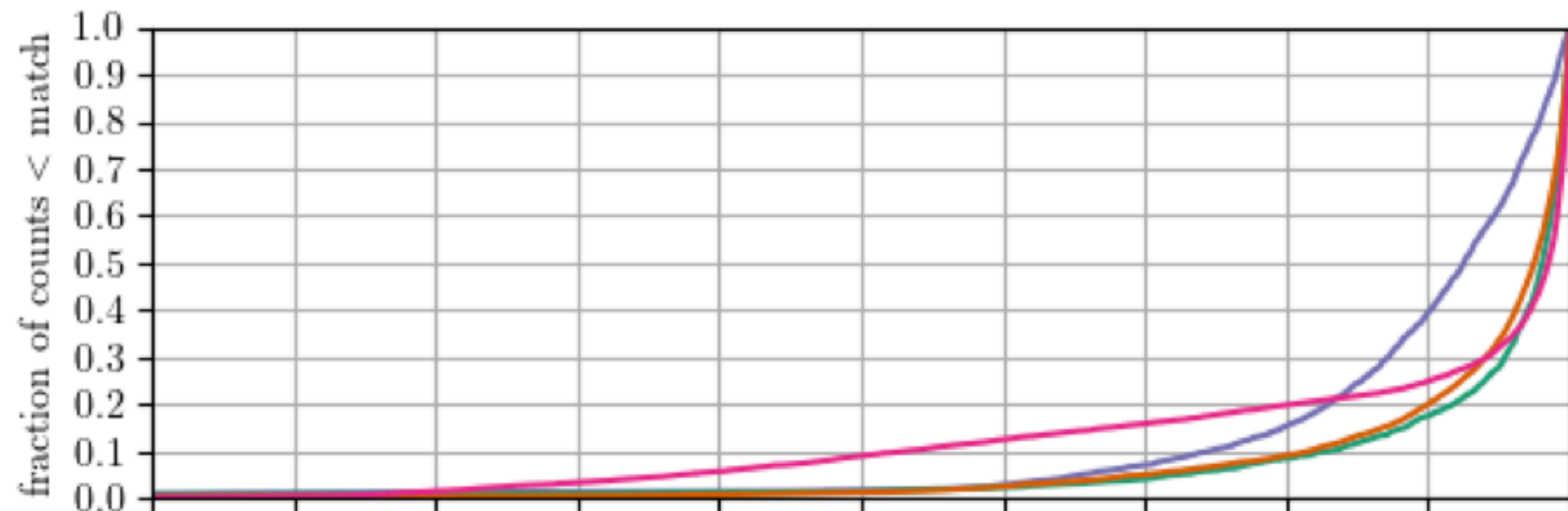


All candidate sources at 12 months

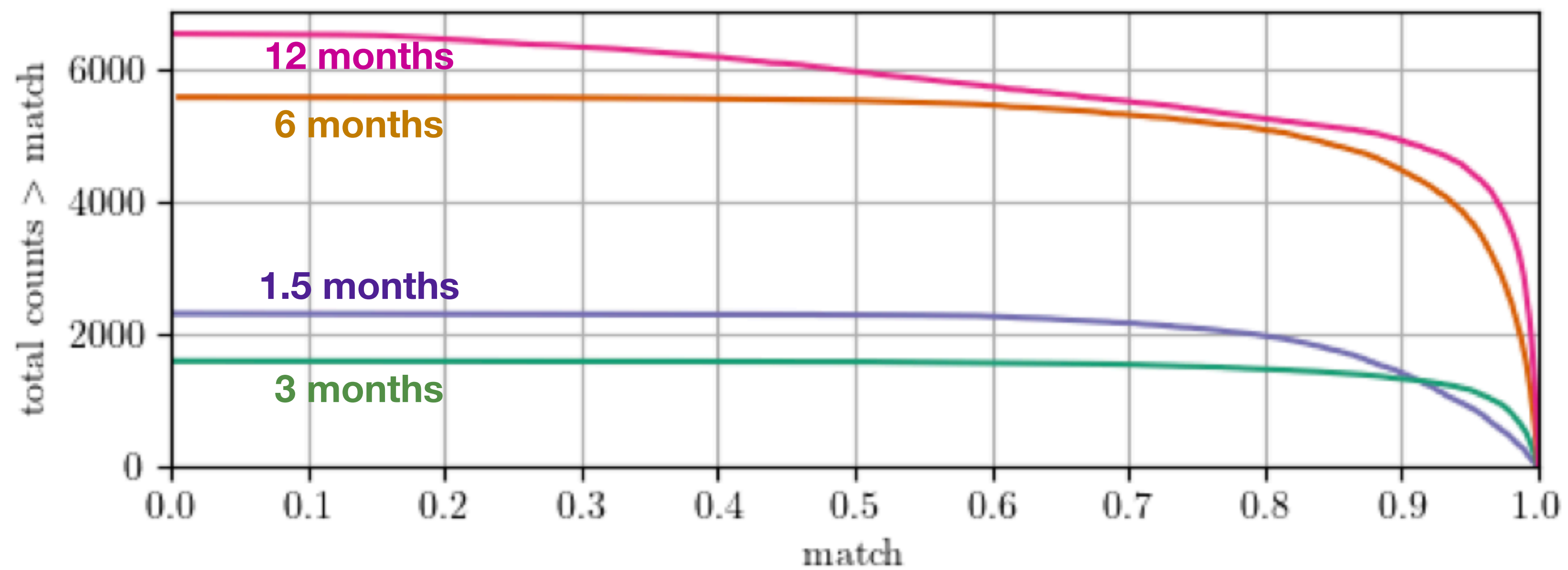


Example of how a source resolves with time

GB matches over time for 90+% confident detections

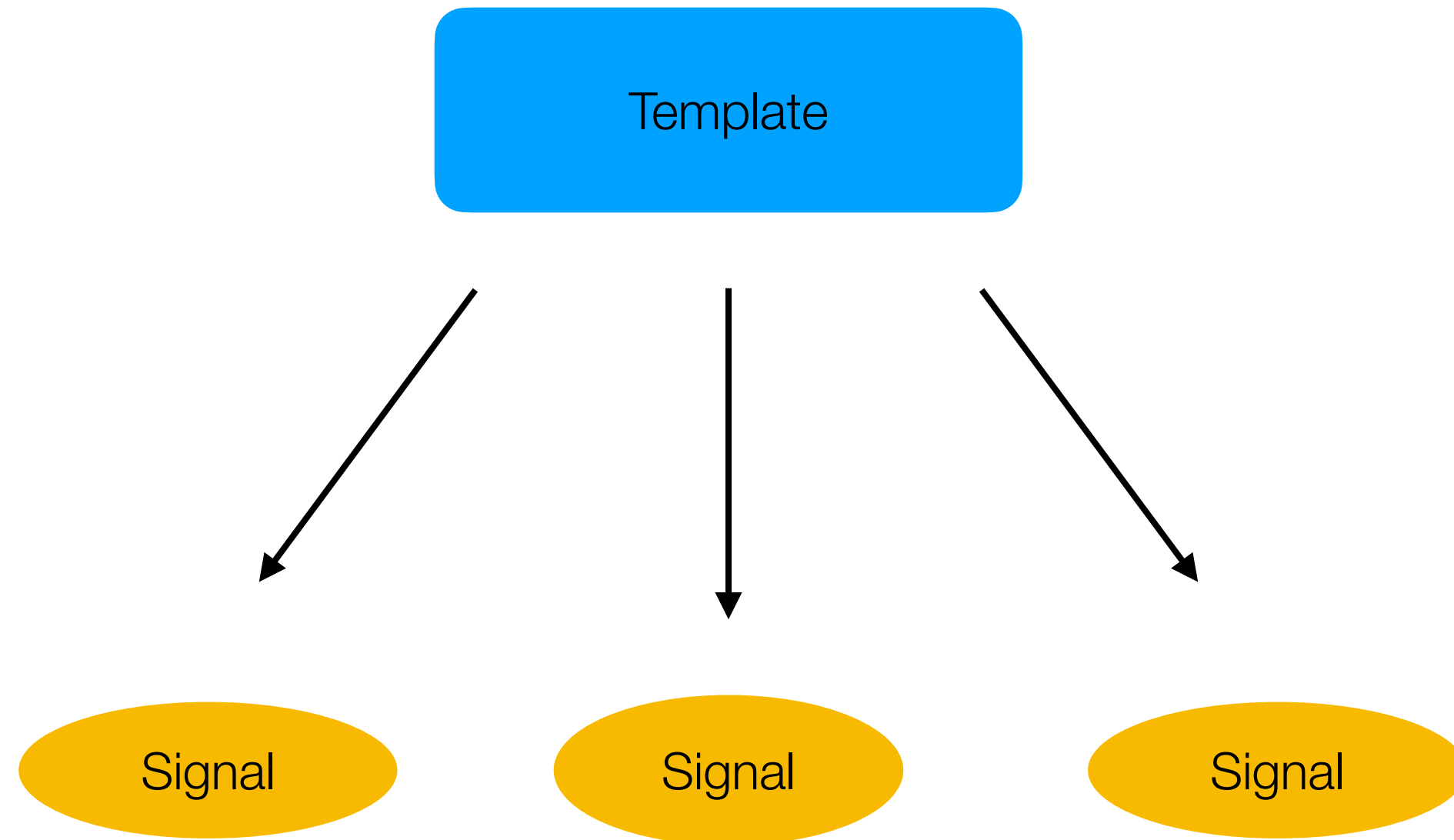


Theory: $M \approx 1 - \frac{D}{2 \text{SNR}^2}$



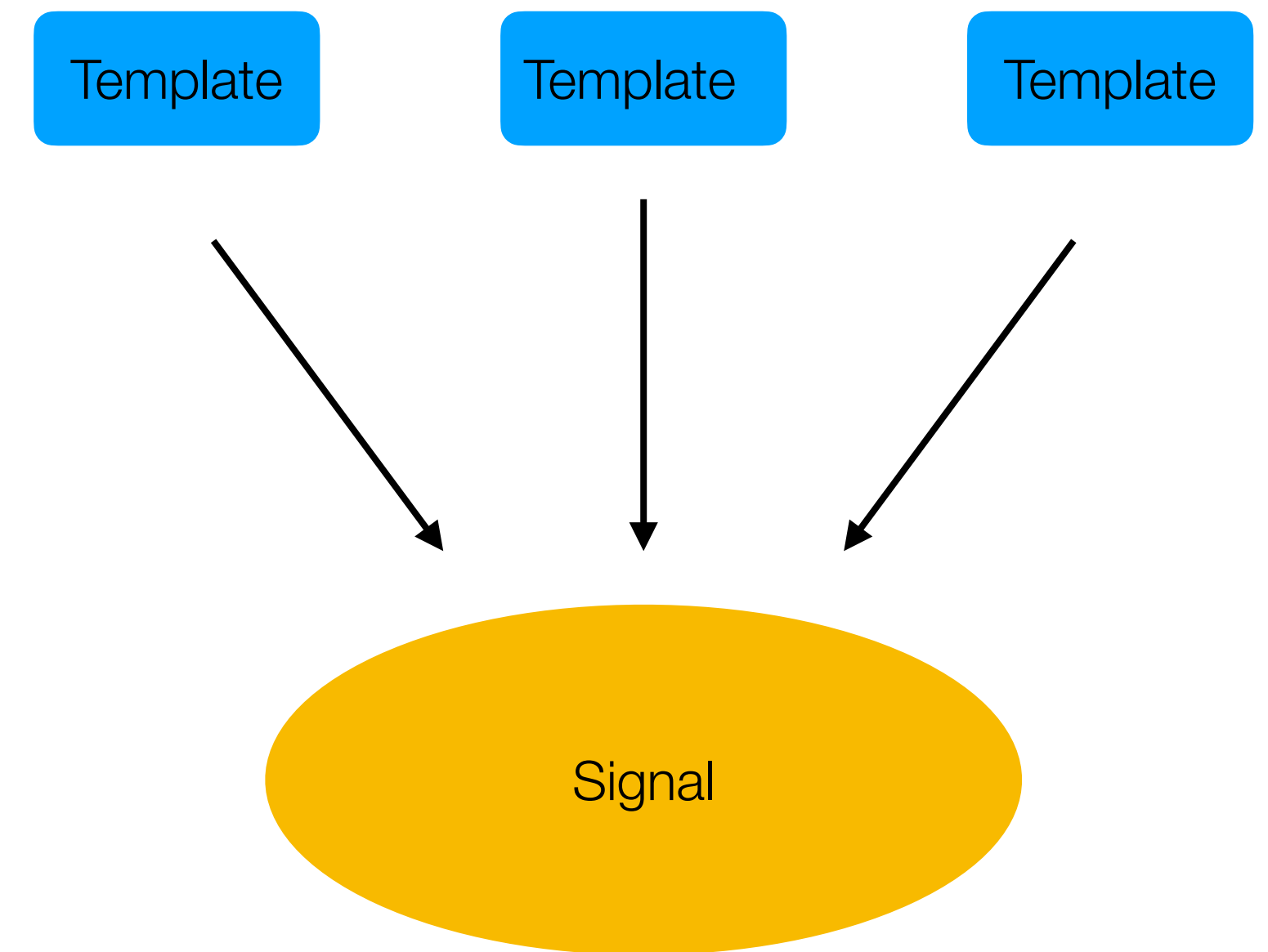
Galactic Binaries - what went wrong at 12 months?

One to Many



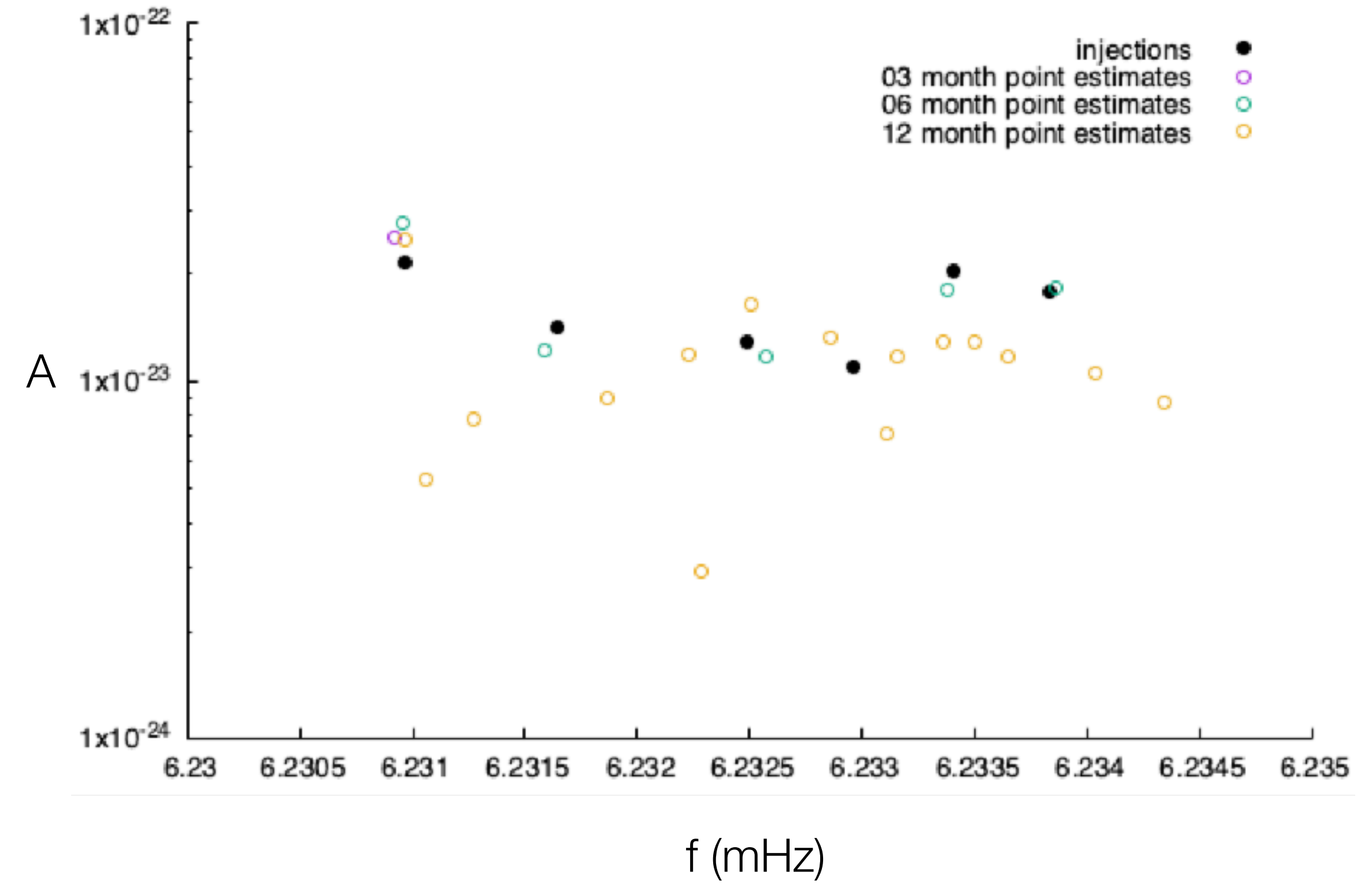
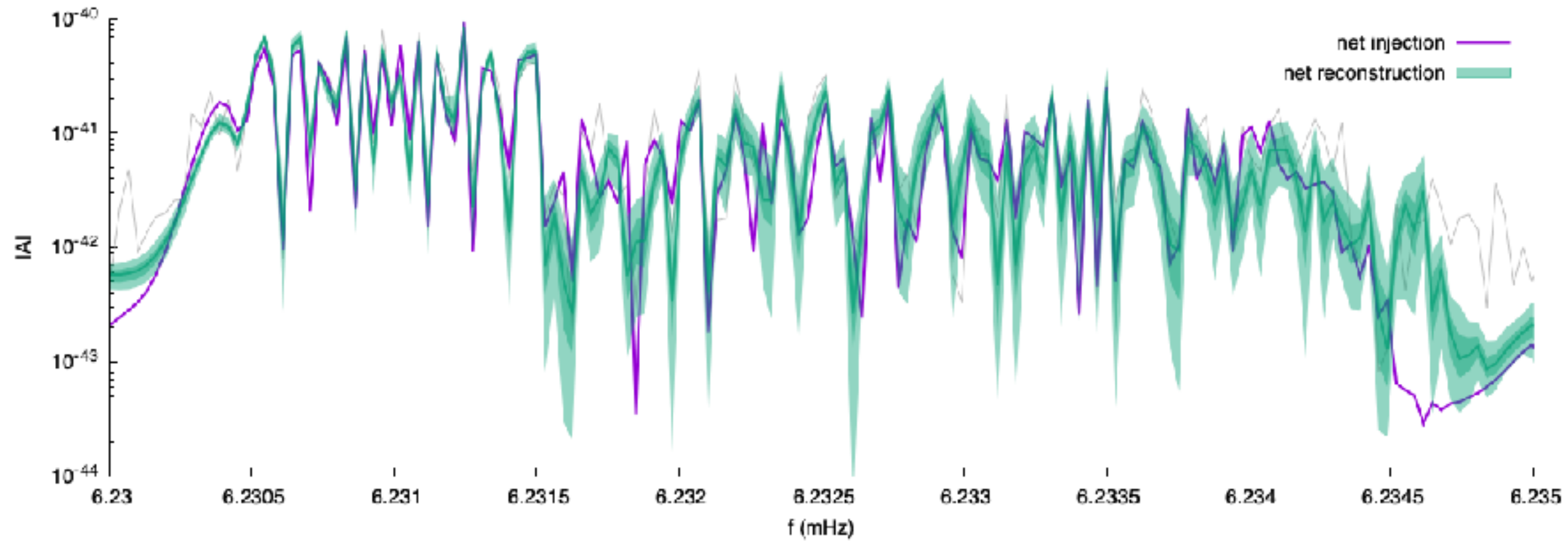
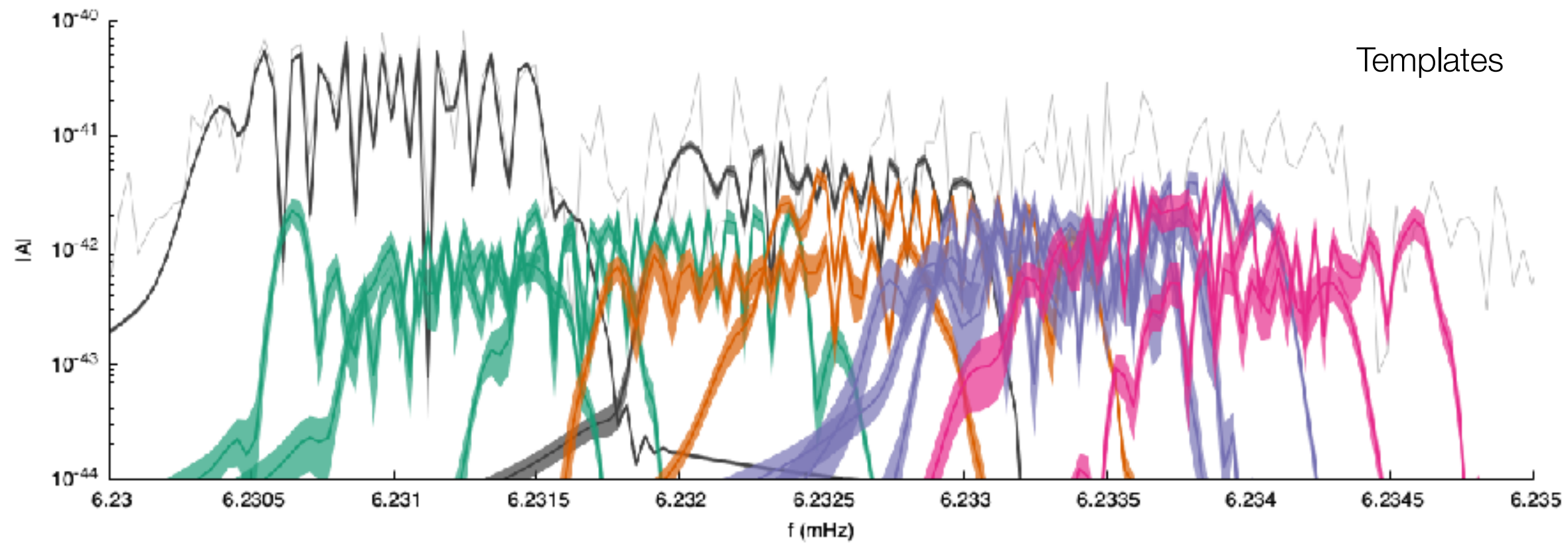
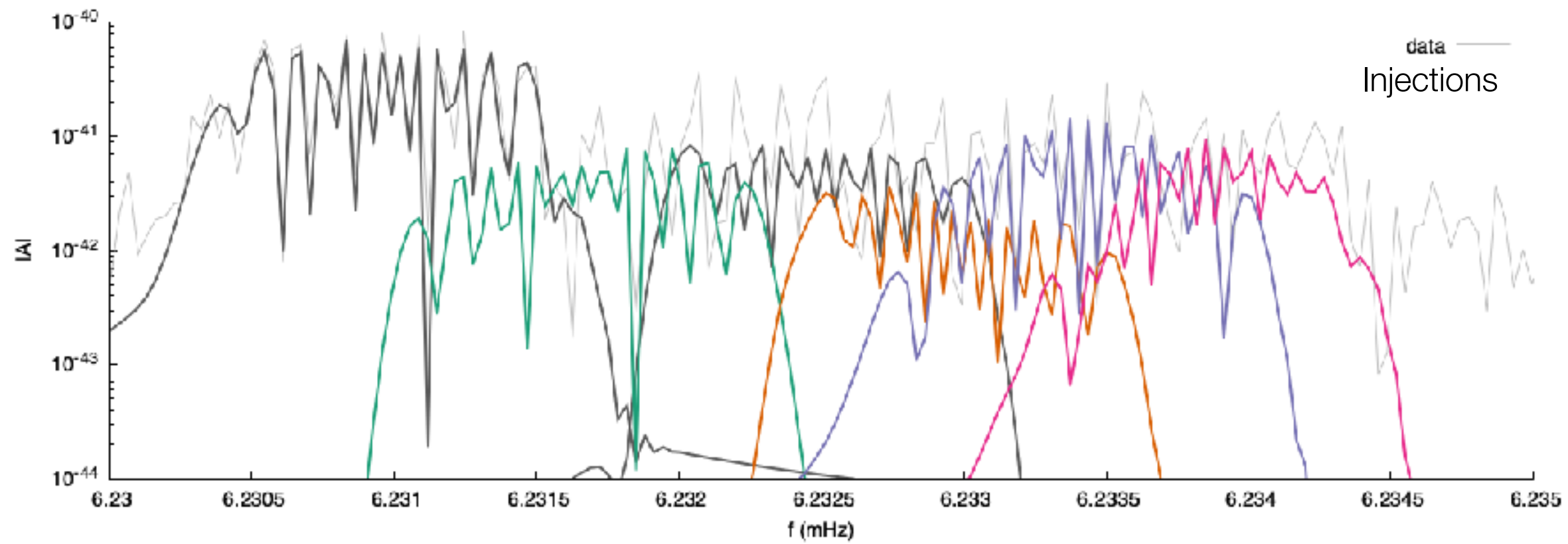
Can be the right answer in a Bayesian sense

Many to One



Never the right answer - poor sampling

Galactic Binaries - what went wrong at 12 months?



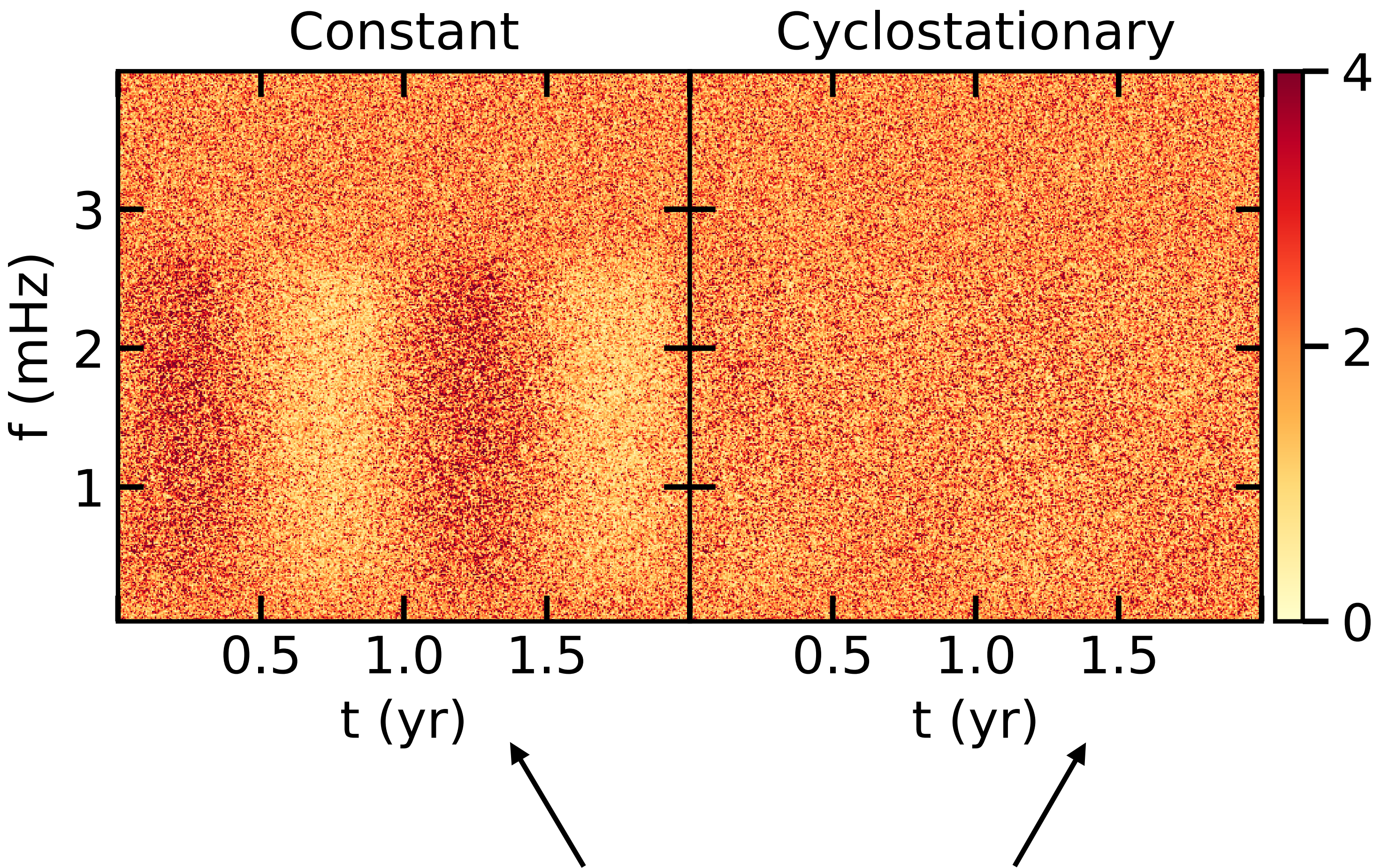
How to do better?

- Better proposals - easiest fix is to increment by smaller amounts in time
- Time-frequency modeling of signals and noise
- Include all three data channels, A, E & T
- Treat the unresolved binaries as a stochastic background (signal), and model the noise component by component
- Include a galaxy shape prior with hyper-parameters for bulge radius, disk radius, disk height etc

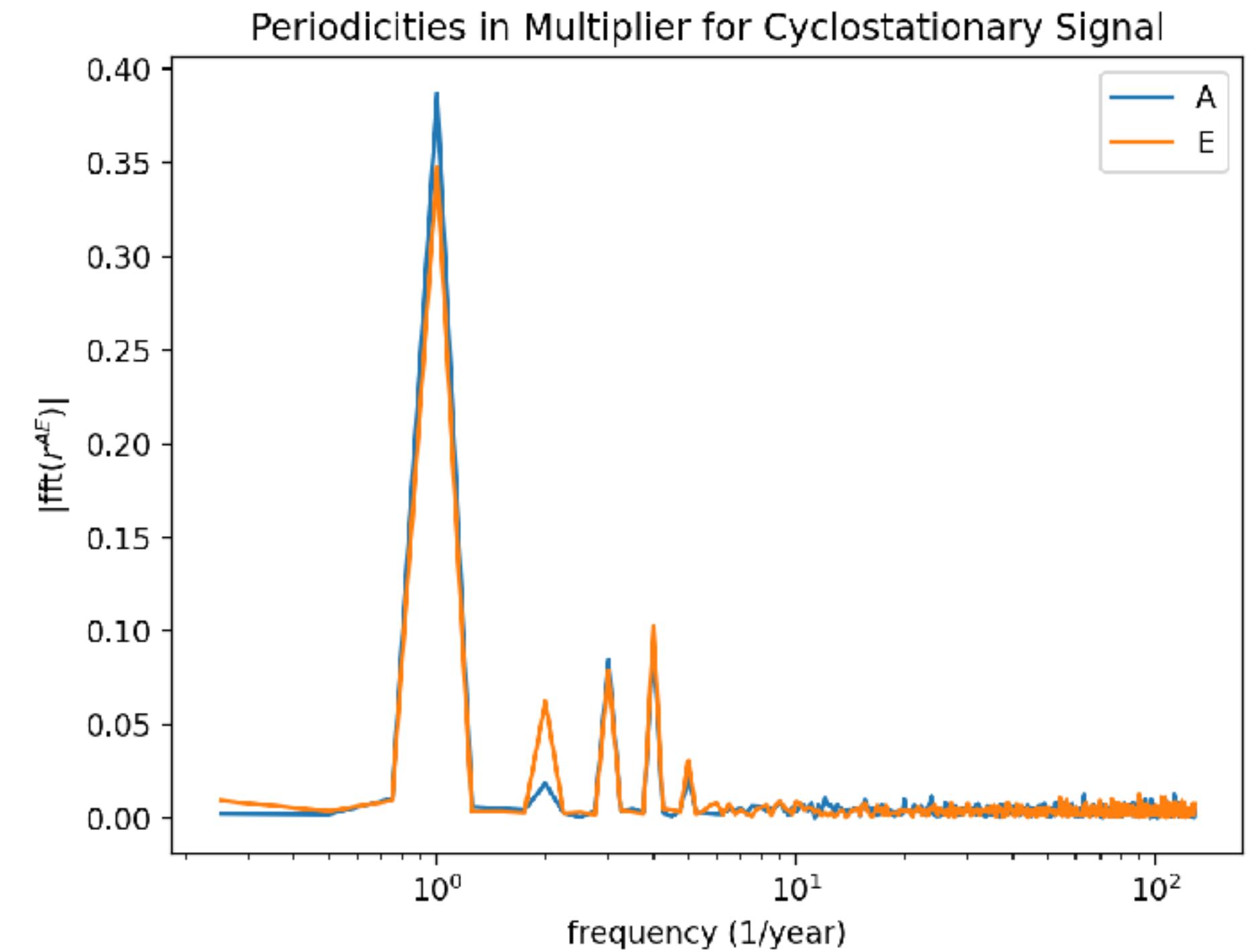
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Galactic Confusion - Cyclostationary Noise



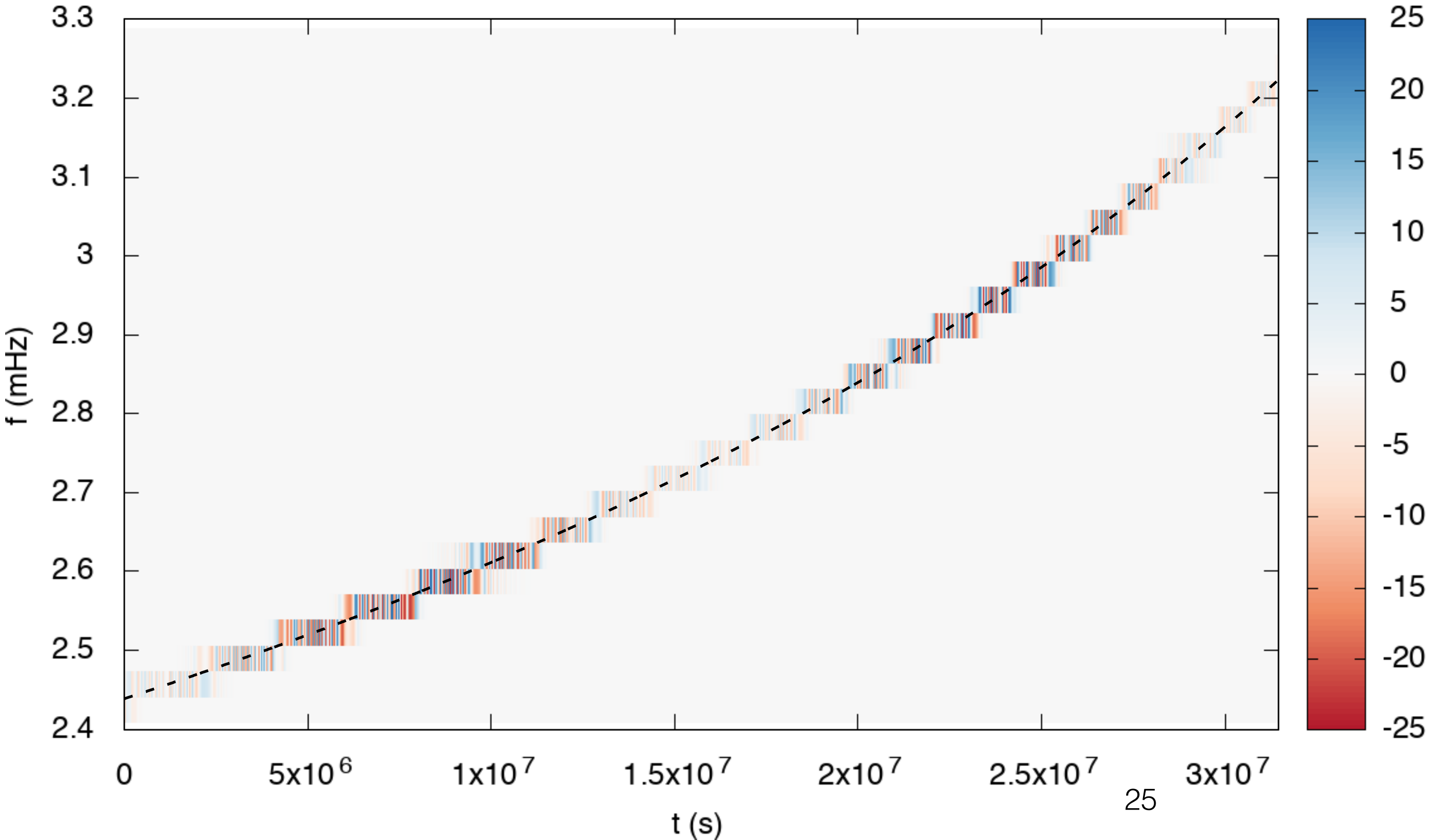
Whitening using constant PSD and dynamic PSD



[Digman & Cornish, [arXiv: 2206.148132](https://arxiv.org/abs/2206.148132)]

Wavelet domain waveforms

$$p(\mathbf{d}|\mathbf{h}) = \frac{1}{\sqrt{\det(2\pi\mathbf{C})}} e^{-\frac{1}{2}(\mathbf{d}-\mathbf{h})^\dagger \mathbf{C}^{-1}(\mathbf{d}-\mathbf{h})}$$



Fast wavelet transforms of the signals for computational efficiency

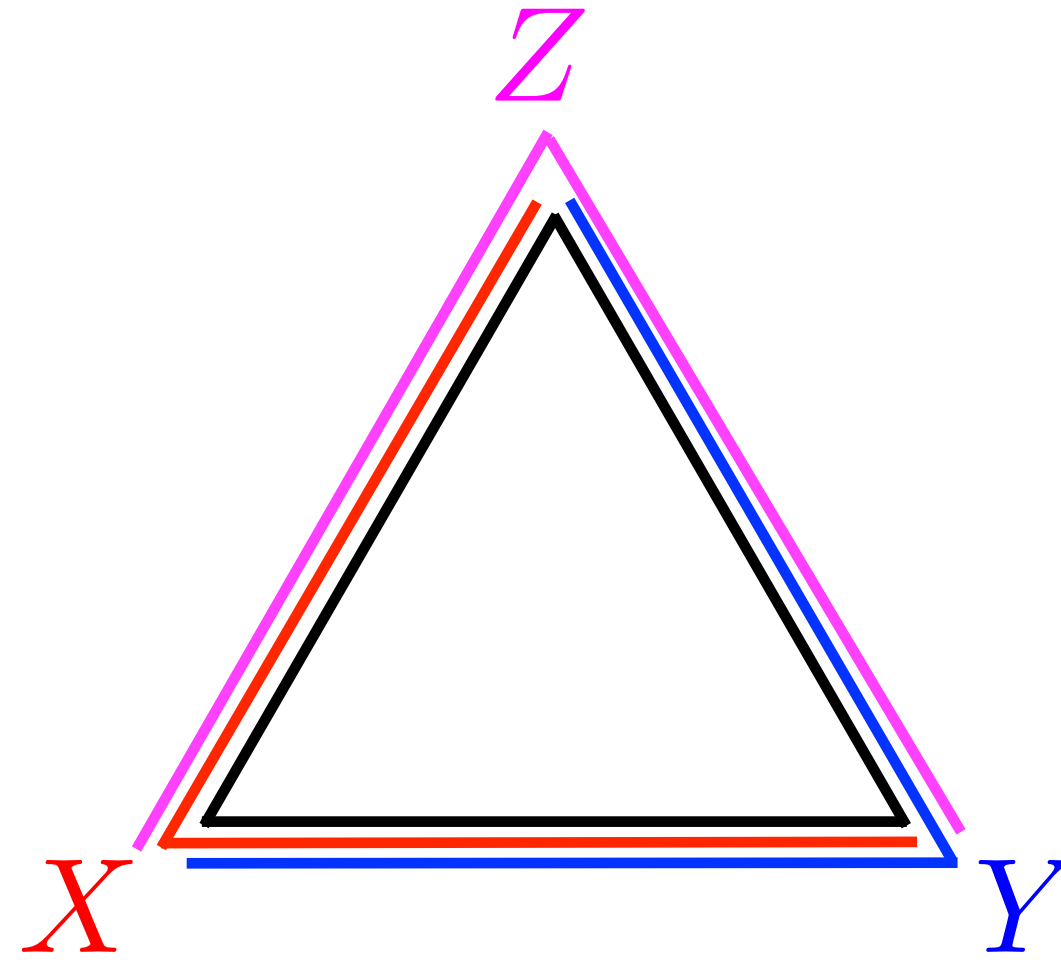
Faster than frequency domain, \sqrt{N} scaling

[Cornish, Phys Rev D **102**, 124038 (2020)]

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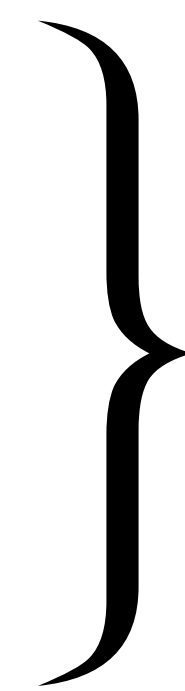
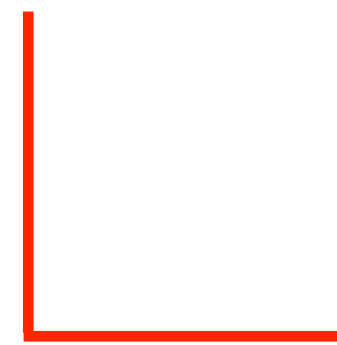
Multiple Data LISA Channels



A

$$S_+ = \frac{\sqrt{3}}{2} X$$

\Rightarrow



Sensitive to GWs

E

$$S_\times = \frac{1}{2} (X + 2Y)$$

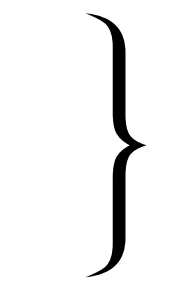
\Rightarrow



T

$$S_\odot = \frac{1}{3} (X + Y + Z)$$

\Rightarrow



Insensitive to GWs

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Galaxy shape prior with hyper-parameters

$$\rho(x, y, z) = \rho_0 \left(A_b e^{-r^2/R_b^2} + (1 - A_b) e^{-\sqrt{x^2 + y^2}/R_d} \operatorname{sech}^2(z/Z_d) \right)$$

