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Neural Posterior Estimation for BBH Foreground Subtraction in 3G SGWB Searches (a manuscript is in preparation)

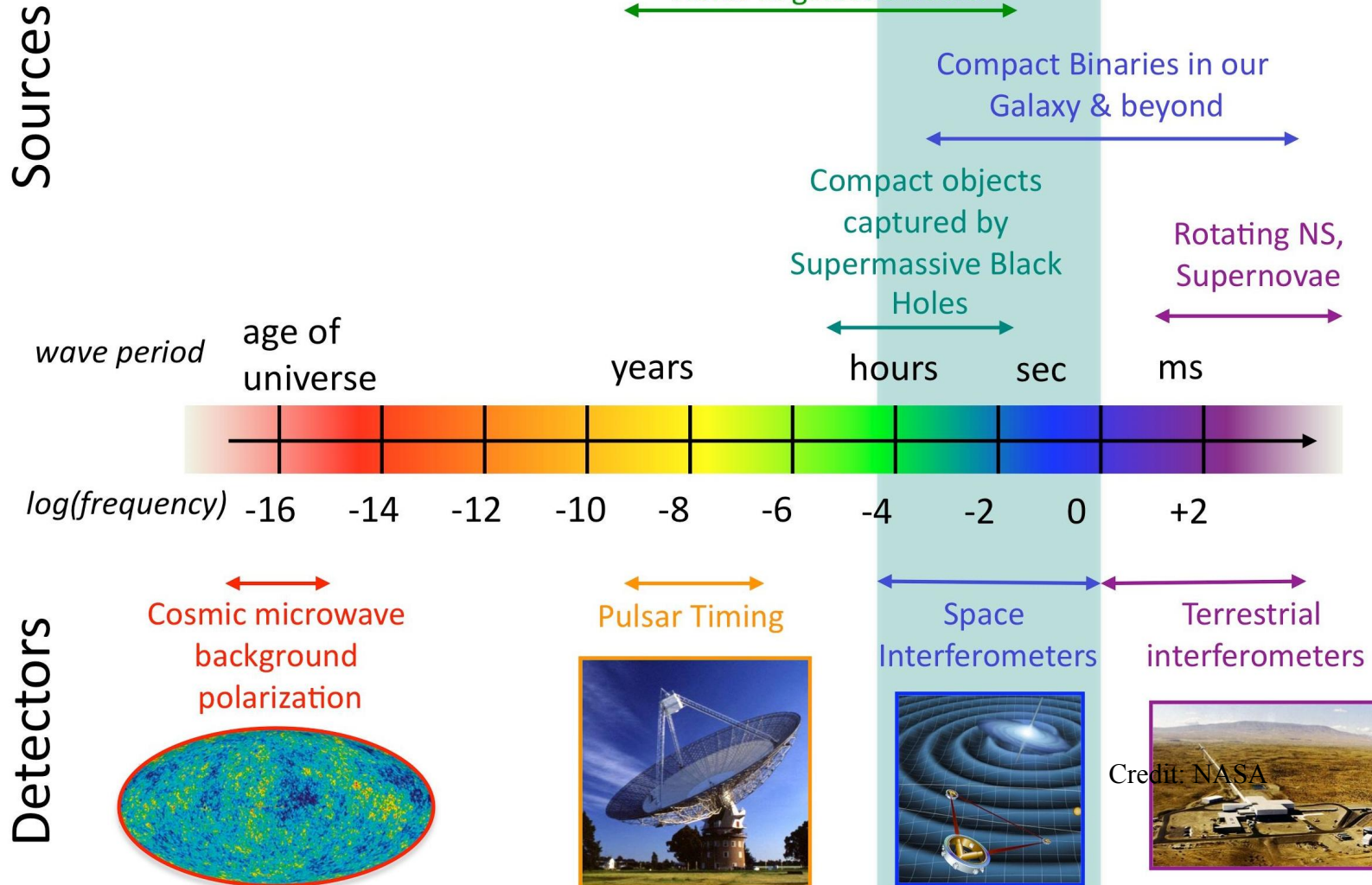
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Institute of Mechanics, Chinese Academy of Sciences

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The Gravitational Wave Spectrum

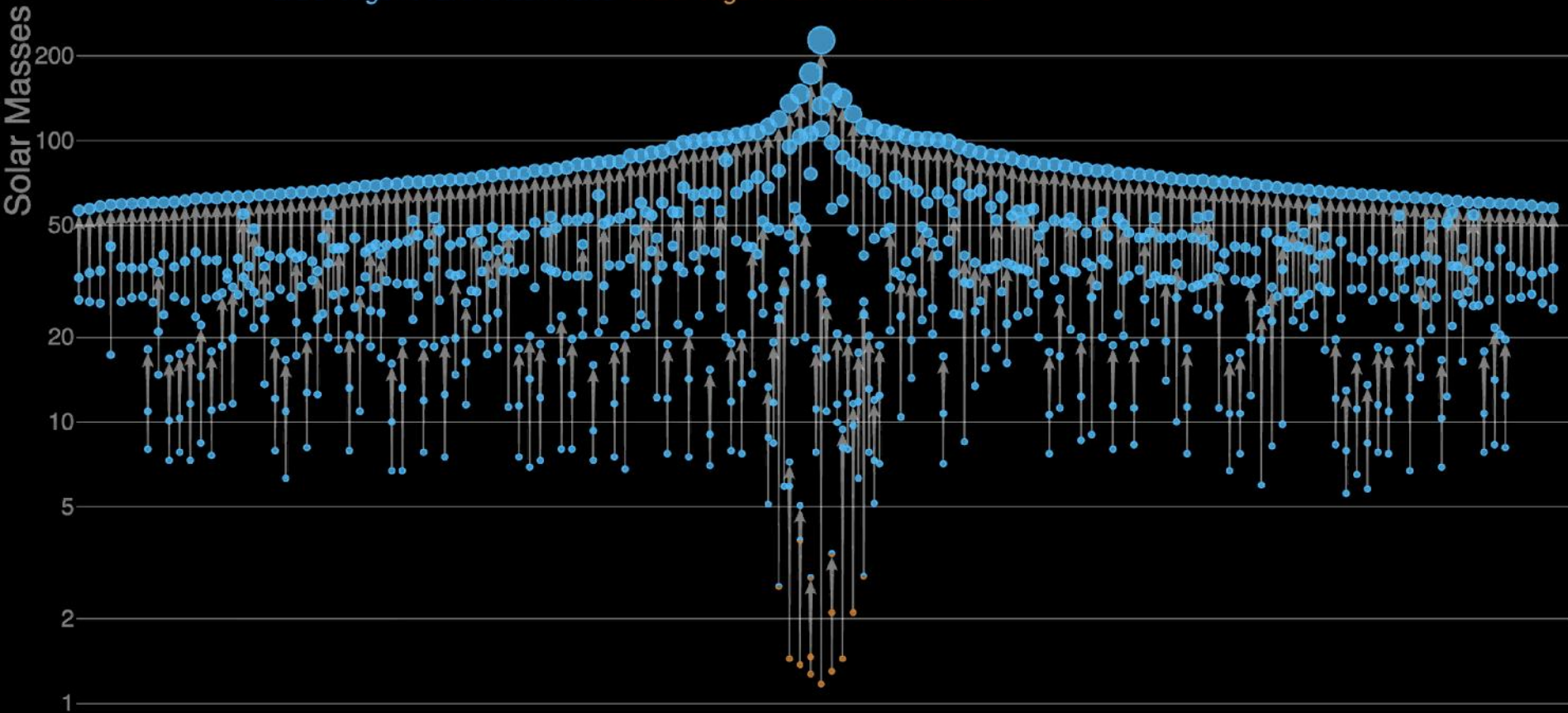


Credit: NASA



Masses in the Stellar Graveyard

LIGO-Virgo-KAGRA Black Holes LIGO-Virgo-KAGRA Neutron Stars



LIGO-Virgo-KAGRA | Aaron Geller | Northwestern



- CBC foreground

$$\Omega_{\text{GW}}(f) := \frac{1}{\rho_{\text{crit}}} \frac{d\rho_{\text{GW}}}{d \ln f} = \frac{4\pi^2}{3H_0^2} f^3 H(f)$$

$$H(f) = \frac{2}{\langle F_+^2 \rangle + \langle F_\times^2 \rangle} \frac{1}{T} \sum_i |h(f)|_i^2 = \frac{5}{T} \sum_i |h(f)|_i^2$$

$$h(f) = F_+(\theta, \phi, \psi) h_+(f) + F_\times(\theta, \phi, \psi) h_\times(f)$$

- Foreground subtraction

Step 1: $\delta h(f) = h(f) - h^{\text{ML}}(f)$

Step 2: $\delta_1^{\text{rfn}} H(f) = \frac{5}{T} \sum \{ |\delta h(f; \Theta)|^2 - \langle |\delta h(f; \Theta^{\text{ML}}(d))|^2 \rangle \}_i$

$$\langle |\delta h(f; \Theta)|^2 \rangle = \langle |h(f; \Theta) - h(f; \Theta^{\text{bst}} | \Theta)|^2 \rangle$$

1, effectiveness of Fisher's method at SNR ~ 10 ?

2, real noise ?

Calculated via Fisher matrix

Pan & Yang (2024), PRD



- CBC foreground with ‘real data’

Fisher

$$\Omega_{\text{GW}}(f) = \frac{4\pi^2}{3H_0^2} f^3 H(f)$$

$$H(f) = \frac{5}{T} \sum_i |h(f)|_i^2$$

$$d_i = h_i + n_i$$

Flow

$$\Omega_{\text{total}}(f) = \frac{4\pi^2}{3H_0^2} f^3 H(f),$$

$$H(f) = \frac{5}{T} \sum_i |d(f)|_i^2$$

- Foreground subtraction

Step 1: $\delta h(f) = h(f) - h^{\text{ML}}(f)$

$\delta d(f) = d(f) - h^{\text{ML}}(f)$

Step 2:

$$\delta_1^{\text{rfn}} H(f) = \frac{5}{T} \sum \{ |\delta h(f; \Theta)|^2 - \langle |\delta h(f; \Theta^{\text{ML}}(d))|^2 \rangle \}_i$$

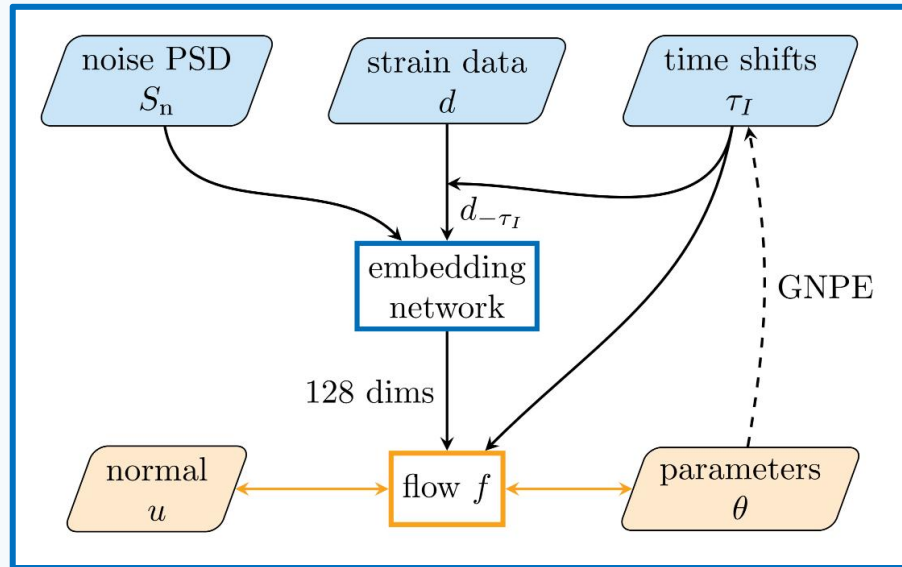
$$\delta_1^{\text{rfn}} H(f) = \frac{5}{T} \sum_i \{ |\delta d(f; \Theta)|^2 - \langle |\delta d(f; \Theta^{\text{ML}}(d))|^2 \rangle \}_i$$

$$\langle |\delta h(f; \Theta)|^2 \rangle = \langle |h(f; \Theta) - h(f; \Theta^{\text{bst}} | \Theta)|^2 \rangle$$

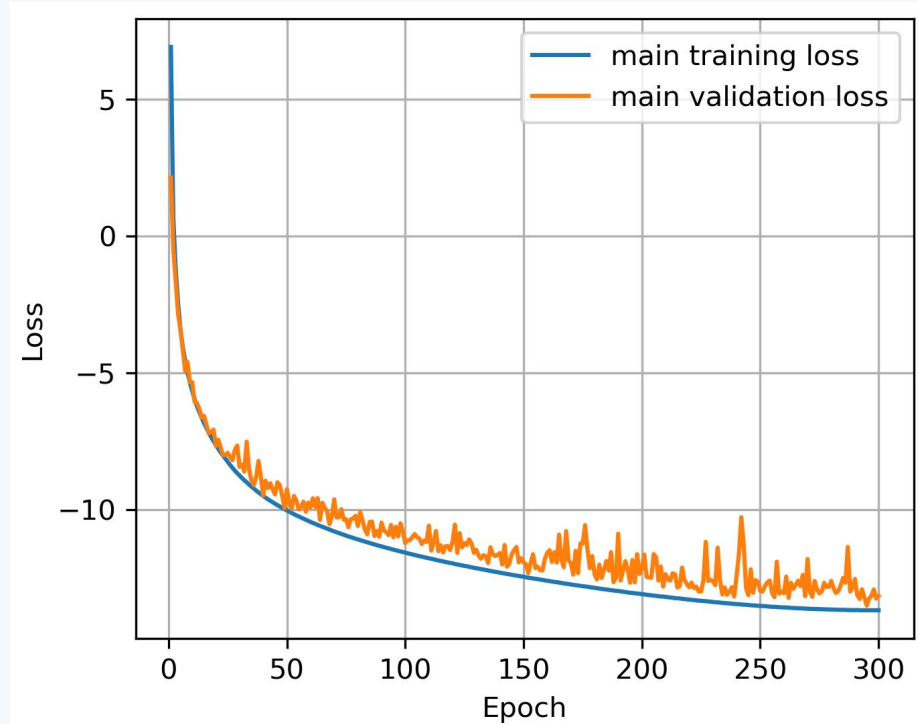
$$\langle |\delta d(f; \Theta)|^2 \rangle = \langle |d(f; \Theta) - h(f; \Theta^{\text{bst}} | \Theta)|^2 \rangle$$



- Fisher is not sufficiently accurate, especially at low SNRs (8–20) and in real noise.
- Bayesian methods are time-consuming for analyses with more than $O(10^4)$ events, often requiring days of inference even on a computing cluster.
- Neural Posterior Estimation (M. Dax, S. Green, et al.), with **5 seconds per inference**.
- Poipulation model: GWTC-4.0 (arXiv:2508.18083), GWTC-5.0 (arXiv:2605.27226)



(Dax et al. PRL, 2021)



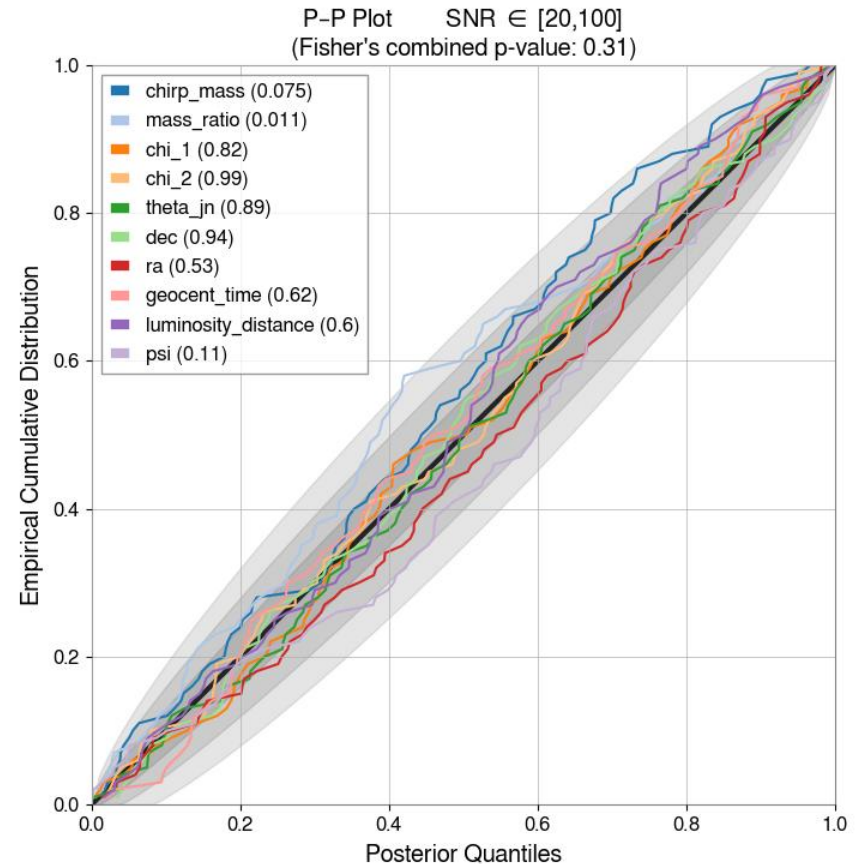
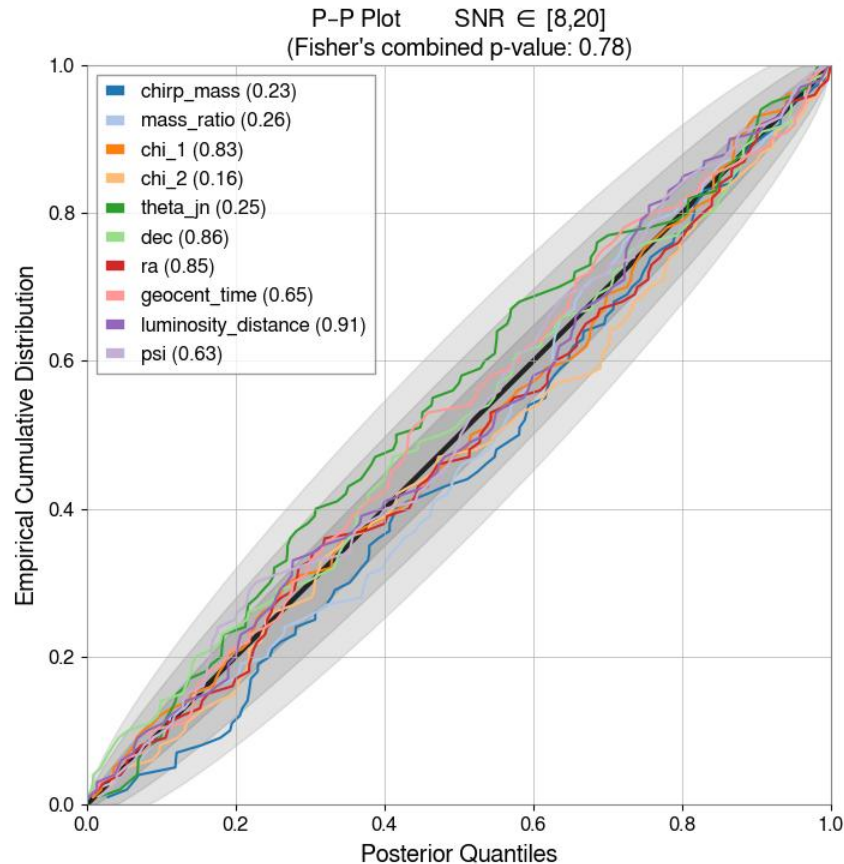
- Twenty million (10^6) mock samples drawn from the LVK-constrained population, assuming CE sensitivity;
- Aligned-spin IMRPhenomD (11D), Trained for 1 epoch in 1 hour, with 5 seconds per inference on RTX 4090 24G.



Probability-Probability (P-P) plot



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P-P plot for 100 injections



Single Event Recovery

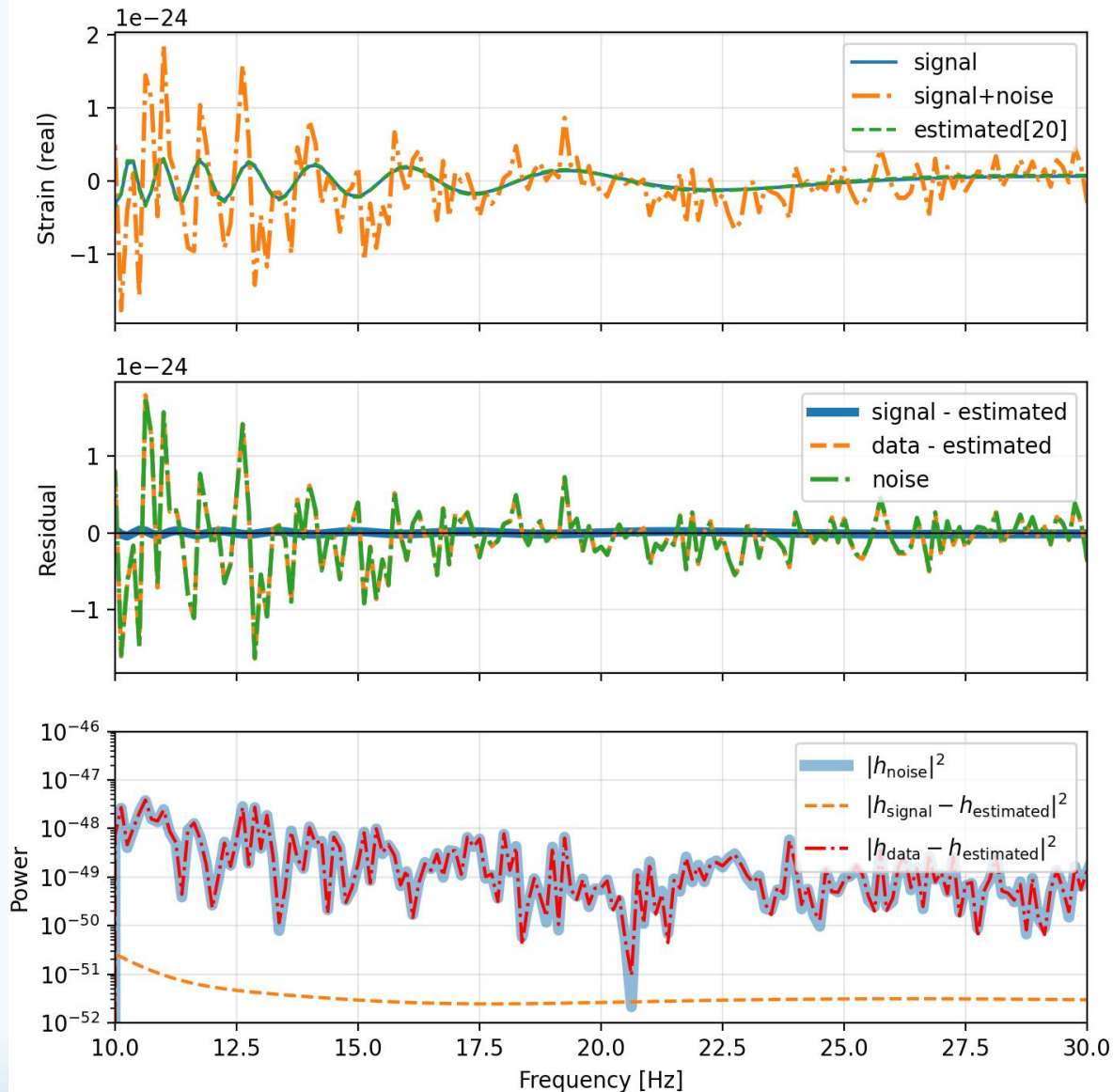


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$$M_c = 48$$

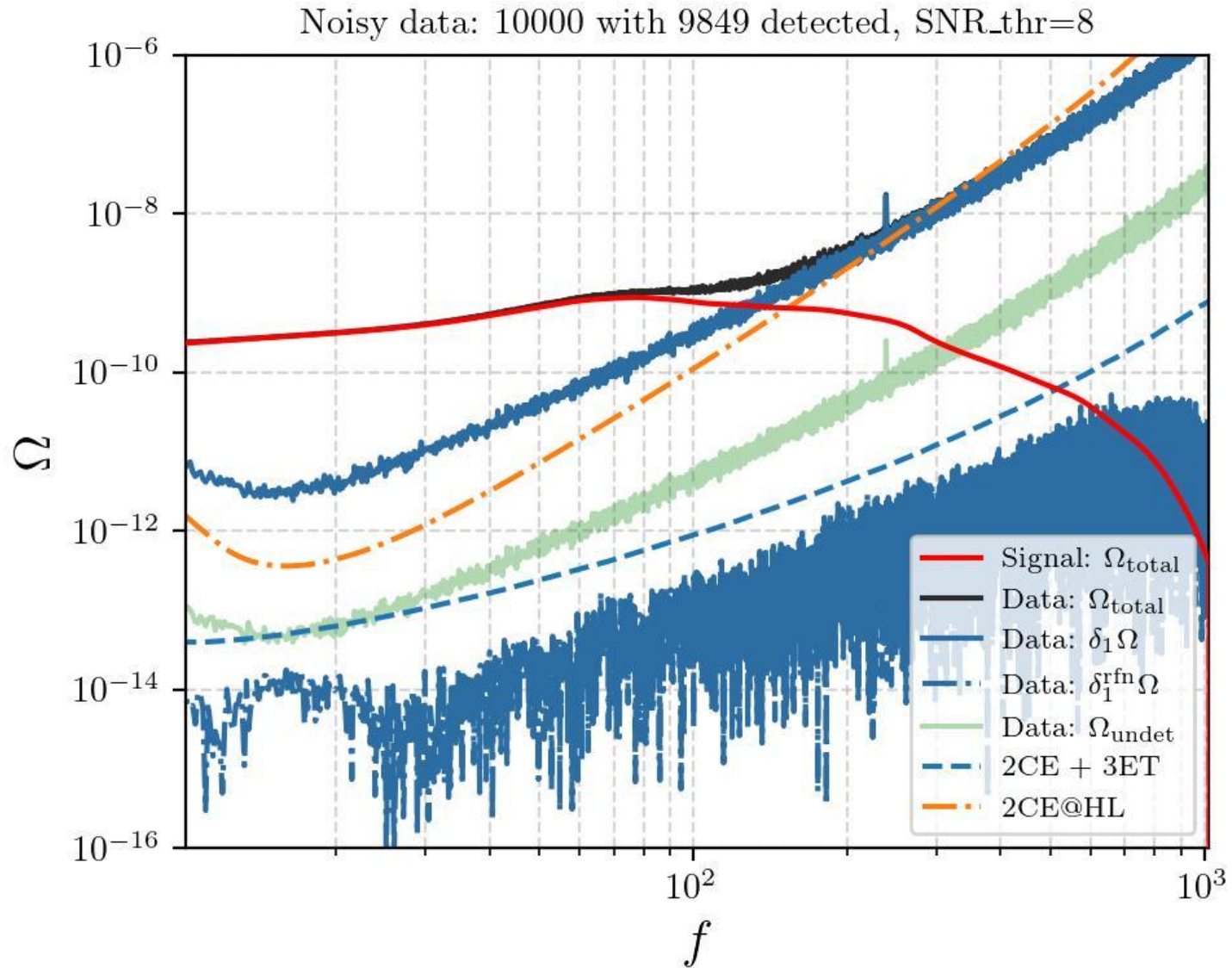
$$\text{SNR} = 10$$

One tricky point is
shifting the phase



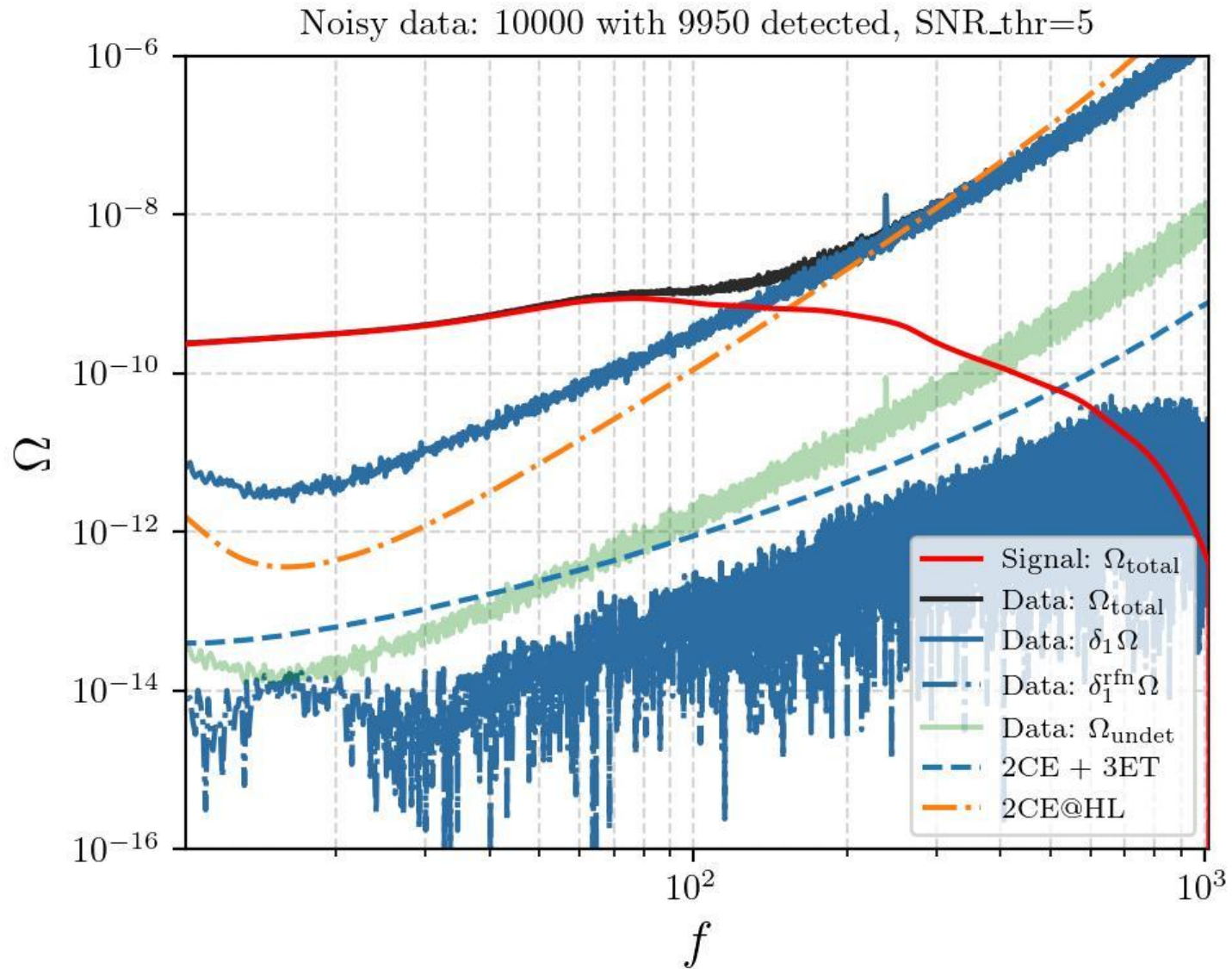


Subtraction Results: Noisy Data





Subtraction Results: Noisy Data





- Neural Posterior Estimation
 - vs. Fisher: more accurate at low SNRs (8–20)
 - vs. Bayesian: in real noise, for aligned-spin IMRPhenomD (11D), training takes 1 hour for 1 epoch, with 5 seconds per inference.
- Science: GW foreground subtraction is sufficiently clean for SGWB detection; the remaining challenge may be resolving low-SNR candidates.



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Thanks for your attention.

Backup slides

