

# Searching for the Stochastic Gravitational-Wave Background

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# **A stochastic gravitational-wave background (SGWB) is created by the superposition of individually undetectable signals.**



The individual contributions expected from the collection of BNS, NSBH, and BBH mergers. While uncertainties on the energy density due to BNS and NSBH are due to Poisson uncertainties in their merger rates, our forecast for the SGWB due to BBHs includes systematic uncertainties associated with their imperfectly known mass distribution. (Right): Estimate of the total gravitational-wave background (green), as well as our current experimental sensitivity (red)







# Component separation for stochastic gravitational-wave background

based on: arXiv:2310.05823, arXiv:2106.09593, arXiv:1904.05056

In our usual analysis, we assume a fiducial model for the spectral shape and perform the optimal filtering .

 $\tilde{\mathcal{Q}}(f)$   $\boldsymbol{\alpha}$ *γI*\* *ft*,*<sup>u</sup> H*(*f*)  $P_{\mathcal{F}_1}(t;f)P_{\mathcal{F}_2}(t;f)$ and de-weight the frequencies that correspond to large detector noise. weight the frequencies which agree with the expected signal spectrum.

 $(f, \Omega) = H(f) \mathcal{P}(\Omega)$ ̂ ̂

SGWB energy density

We usually assume

We often choose a power-law functional form for the SGWB template spectrum

$$
\Omega_{\text{GW}}(f,\hat{\Omega}) = \frac{2\,\pi^2}{3H_0^2}\,f^3\,\mathcal{P}(f,\hat{\Omega})
$$

# **A stochastic gravitational-wave background (SGWB) is created by the superposition of individually undetectable signals.**

$$
H(f) = \left(f/f_{\text{ref}}\right)^{\alpha}
$$



\* see G. Boileau+ PRD 103, 103529 (2021) for a nice summary of all the proposed methods

- Separating the contribution of these sources to the total observed background.
- If we filter the data for each GWB component separately: we overestimate the amplitude of each GWB component and underestimate the error bars.
	- since other GWB component is also contributing to the correlated signal.
- We need to go beyond the single component analysis, to better extract the amplitudes of individuals GWB components.
	- Many methods have been proposed to disentangle these components.\* • these are appropriate for estimating the parameters associated with different
		- SGWB models.
	- Parida et. al [JCAP 04, 024 (2016)] proposed a method to jointly estimate the GWB components.
		- joint analysis takes into account the covariance between the spectral shapes.
	- We extended this to the astrophysical SGWB and anisotropic SGWB





SGWB energy density

We usually assume

amplitude of the SGWB intensity

spectral shape of the background  $\;\; H_{\alpha}(f) \quad$  where  $\alpha$  is the spectral index

 $(f, \Omega) = H(f) \mathcal{P}(\Omega)$ ̂ ̂

$$
\Omega_{\text{GW}}(f, \hat{\Omega}) = \frac{2\pi^2}{3H_0^2} f^3 \mathcal{P}(f, \hat{\Omega})
$$

$$
\mathscr{P}(f,\hat{\Omega}) = \sum_{\alpha} H_{\alpha}(f) \mathscr{P}^{\alpha}(\hat{\Omega})
$$

### **Injection Studies - Isotropic background**











### combined source coupling matrix

### **Injection Studies**



![](_page_10_Figure_0.jpeg)

![](_page_10_Figure_3.jpeg)

![](_page_10_Picture_4.jpeg)

 $-1.95152e-51$ 

 $6.50911e-51$ 

![](_page_10_Figure_7.jpeg)

11

![](_page_10_Picture_9.jpeg)

### **Injection Studies**

95% confident UL - Injection

![](_page_11_Figure_5.jpeg)

![](_page_11_Figure_6.jpeg)

- Simulated 1000 noise realizations.
- Perform the injection study considering each noise realization.
- 

• Recovered the source using single-index and joint-index multi-component separation methods.

 $\alpha = 0$   $\alpha = 2/3$   $\alpha = 3$ 

![](_page_12_Figure_3.jpeg)

## **Upper limit from injection study**

• Produced UL sky maps corresponding to the injection (injection strength is set to be close to the detectable limit).

- 
- Histogram shows the difference between upper limit sky map and the injected sky map.

even if the detectors are not sensitive enough to detect SGWB, the joint-index multi-component estimator provides safer upper limits when one cannot ignore the existence of more than one component.

# $1e - 51$

![](_page_13_Figure_0.jpeg)

	$\Omega_0 = 1 \times 10^{-6}$	$\Omega_{2/3} = 1 \times 10^{-6}$	$\Omega_2 = 1 \times 10^{-6}$	$\Omega_3 = 1 \times 10^{-6}$	$\Omega_4 = 1 \times 10^{-6}$
$\alpha=\{0\}$	$(1.9420 \pm 0.0008) \times 10^{-5}$				
$\alpha = \{2/3\}$		$(1.8066 \pm 0.0006) \times 10^{-5}$			
$\alpha = \{2\}$			$(1.2519 \pm 0.0002) \times 10^{-5}$ -		
$\alpha = \{3\}$				$(5.6650 \pm 0.0008) \times 10^{-6}$ -	
$\alpha = \{4\}$					$(1.2262 \pm 0.0002) \times 10^{-6}$
$\alpha = \{0, 2/3, 2, 3\}$	$(-1.525\pm0.001)\times10^{-4}$	$(2.045 \pm 0.002) \times 10^{-4}$	$(-7.061 \pm 0.004) \times 10^{-5}$	$(1.7369 \pm 0.0005) \times 10^{-5}$	
$\alpha = \{0, 2/3, 2, 4\}$	$(1.18 \pm 0.01) \times 10^{-5}$	$(-1.30\pm0.01)\times10^{-5}$	$(5.68 \pm 0.02) \times 10^{-6}$		$(1.054 \pm 0.0003) \times 10^{-6}$
$\alpha = \{0, 2/3, 3, 4\}$	$(-1.50\pm0.08)\times10^{-6}$	$(4.18 \pm 0.07) \times 10^{-6}$		$(1.202 \pm 0.004) \times 10^{-6}$	$(9.898 \pm 0.005) \times 10^{-7}$
$\alpha=\{0,2,3,4\}$	$(1.82 \pm 0.03) \times 10^{-6}$		$(1.29 \pm 0.02) \times 10^{-6}$	$(9.44 \pm 0.07) \times 10^{-7}$	$(1.0027 \pm 0.0006) \times 10^{-6}$
$\alpha = \{2/3, 2, 3, 4\}$		$(2.21 \pm 0.03) \times 10^{-6}$	$(6.6 \pm 0.3) \times 10^{-7}$	$(1.063 \pm 0.009) \times 10^{-6}$	$(9.970 \pm 0.007) \times 10^{-7}$
	$\alpha = \{0, 2/3, 2, 3, 4\}$ $(1.0 \pm 0.2) \times 10^{-6}$	$(1.0\pm0.3)\times10^{-6}$	$(1.00 \pm 0.07) \times 10^{-6}$	$(1.00 \pm 0.02) \times 10^{-6}$	$(1.0000 \pm 0.0009) \times 10^{-6}$

CBC r-mode r-mode

![](_page_13_Picture_7.jpeg)

### **Summary**

• If we filter the data for each GWB component separately, we overestimate the amplitude of each

• We have shown that estimates and the upper limits can get severely biased in the single-index

• Joint analysis accurately separates and estimates backgrounds with different spectral shapes and

![](_page_14_Picture_0.jpeg)

- GWB component and underestimate the error bars.
- analysis when the actual signal is close to more than the detectable limit.
- different sky distributions with no major bias.
- 
- methods are general and should be easily translated to LISA and PTA bands.

• The upper limits set by the joint analysis are safer, though less strict than the individual analysis.

• While the results shown in this presentation are in the context of ground-based detectors, the

• If we filter the data for each GWB component separately, we overestimate the amplitude of each

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• Joint analysis accurately separates and estimates backgrounds with different spectral shapes and

![](_page_15_Picture_0.jpeg)

- GWB component and underestimate the error bars.
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• While the results shown in this presentation are in the context of ground-based detectors, the

![](_page_15_Picture_12.jpeg)

![](_page_15_Figure_14.jpeg)

thank you!

### **Summary**

SGWB energy density  $\Omega_{\text{GW}}(f, \mathcal{S})$ 

- amplitude of the SGWB intensity  $\mathscr{P}(f,$ 
	- spectral shape of the background
- ⟨ ℐ⟩ <sup>=</sup> *<sup>τ</sup>*<sup>∑</sup> Usual Cross-Spectral Density (CSD)

likelihood function  $p(\mathscr{C}^{\mathscr{I}} | \mathscr{P}_{\mathscr{U}}^{\alpha})$ 

clean map

$$
\hat{\Omega} = \frac{2 \pi^2}{3H_0^2} f^3 \mathcal{P}(f, \hat{\Omega})
$$

$$
\hat{\Omega} = \sum_{\alpha} H_{\alpha}(f) \mathcal{P}^{\alpha}(\hat{\Omega})
$$

 $H_{\alpha}(f)$ where  $\alpha$  is the spectral index

$$
\mathcal{P}_u^{\alpha} \propto \exp \left[ -\frac{1}{2} (\mathcal{C}^{\mathcal{J}} - \langle \mathcal{C}^{\mathcal{J}} \rangle)^* \mathcal{N}^{-1} (\mathcal{C}^{\mathcal{J}} - \langle \mathcal{C}^{\mathcal{J}} \rangle) \right]
$$

$$
\hat{\mathscr{P}}_u^{\alpha} = \Gamma^{-1} \cdot \mathbf{X}
$$

### **Preliminaries**

# (1) (2)

### (3)

### (4)

$$
\mathcal{I}\rangle = \tau \sum_{\alpha} H_{\alpha}(f) \gamma_{ft,u}^I \mathcal{P}_{u}^{\alpha}
$$

(5)

 $\Gamma\equiv\Gamma^{\alpha\beta}_{\mu\nu}$  $\frac{\alpha p}{\mu u'} = \sum$ *Ift Hα*(*f*)*Hβ*(*f*)  $P_{\mathscr{I}_1}(t;f) \, P_{\mathscr{I}_2}(t;f)$ *γI*\* *ft*,*<sup>u</sup> γ<sup>I</sup> ft*,*u*′

### Fisher information matrix

![](_page_18_Figure_7.jpeg)

For a three spectral index case, we can write the convolution equation as

ML solution of the convolution equation

 $\mathbf{X} \equiv X_u^{\alpha}$  $=$   $\sum$ *Ift γI*\* *ft*,*u Hα*(*f*)  $P_{\mathcal{F}_1}(t;f)P_{\mathcal{F}_2}(t;f)$  $C^I(t;f)$ 

$$
\hat{\mathcal{P}}_{u}^{\alpha} = \left[ C_{uu'}^{\alpha\beta} \right]^{-1} \cdot X_{u'}^{\beta}
$$

### **Preliminaries**

### (1)

### (2)

### (3)

![](_page_18_Picture_13.jpeg)

	$\tilde{\Omega}_0 = 1 \times 10^{-6}$	$\Omega_{2/3} = 1 \times 10^{-6}$	$\hat{\Omega}_2 = 1 \times 10^{-6}$	$\tilde{\Omega}_3 = 1 \times 10^{-6}$	$\tilde{\Omega}_4 = 1 \times 10^{-6}$
$\alpha = \{0\}$	$(1.9420 \pm 0.0008) \times 10^{-5}$				
$\alpha = \{2/3\}$		$(1.8066 \pm 0.0006) \times 10^{-5}$			
$\alpha = \{2\}$			$(1.2519 \pm 0.0002) \times 10^{-5}$ -		
$\alpha = \{3\}$				$(5.6650 \pm 0.0008) \times 10^{-6}$	
$\alpha = \{4\}$					$(1.2262 \pm 0.0002) \times 10^{-6}$
$\alpha=\{0,2/3\}$	$(-1.5621 \pm 0.0004) \times 10^{-4}$	$(1.3276 \pm 0.0003) \times 10^{-4}$			
$\alpha=\{0,2\}$	$(-3.595\pm0.001)\times10^{-5}$		$(2.1765 \pm 0.0004) \times 10^{-5}$ -		
$\alpha = \{0,3\}$	$(-9.516 \pm 0.009) \times 10^{-6}$			$(6.205 \pm 0.001) \times 10^{-6}$	
$\alpha=\{0,4\}$	$(7.665 \pm 0.008) \times 10^{-6}$				$(1.1886 \pm 0.0002) \times 10^{-6}$
$\alpha=\{2/3,2\}$		$(-3.874 \pm 0.001) \times 10^{-5}$	$(2.7240 \pm 0.0005) \times 10^{-5}$		
$\alpha = \{2/3, 3\}$		$(-9.227 \pm 0.007) \times 10^{-6}$		$(6.513 \pm 0.001) \times 10^{-6}$	
$\alpha = \{2/3, 4\}$		$(6.305 \pm 0.006) \times 10^{-6}$			$(1.1696 \pm 0.0002) \times 10^{-6}$
$\alpha=\{2,3\}$			$(-9.976\pm0.005)\times10^{-6}$	$(8.751 \pm 0.002) \times 10^{-6}$	
$\alpha=\{2,4\}$			$(3.521 \pm 0.003) \times 10^{-6}$		$(1.0827 \pm 0.0002) \times 10^{-6}$
$\alpha=\{3,4\}$				$(1.819 \pm 0.002) \times 10^{-6}$	$(9.141 \pm 0.003) \times 10^{-7}$
$\alpha=\{0,2/3,2\}$	$(2.253 \pm 0.001) \times 10^{-4}$	$(-2.5345\pm0.0009)\times10^{-4}$	$(5.086 \pm 0.001) \times 10^{-5}$		
$\alpha=\{0,2/3,3\}$	$(9.418 \pm 0.007) \times 10^{-5}$	$(-8.525\pm0.005)\times10^{-5}$		$(8.154 \pm 0.002) \times 10^{-6}$	
$\alpha=\{0,2/3,4\}$	$(-2.188\pm0.005)\times10^{-5}$	$(2.278 \pm 0.004) \times 10^{-5}$			$(1.1288 \pm 0.0002) \times 10^{-6}$
$\alpha=\{0,2,3\}$	$(2.918 \pm 0.002) \times 10^{-5}$		$(-2.584\pm0.001)\times10^{-5}$	$(1.2001 \pm 0.0003) \times 10^{-5}$	
$\alpha=\{0,2,4\}$	$(-9.7\pm0.2)\times10^{-7}$		$(3.839 \pm 0.006) \times 10^{-6}$		$(1.0745 \pm 0.0002) \times 10^{-6}$
$\alpha=\{0,3,4\}$	$(3.35 \pm 0.01) \times 10^{-6}$			$(1.385 \pm 0.002) \times 10^{-6}$	$(9.722 \pm 0.004) \times 10^{-7}$
$\alpha = \{2/3, 2, 3\}$		$(3.603 \pm 0.002) \times 10^{-5}$	$(-3.527\pm0.002)\times10^{-5}$	$(1.3262 \pm 0.0003) \times 10^{-5}$	
$\alpha = \{2/3, 2, 4\}$		$(-1.20\pm0.02)\times10^{-6}$	$(4.076 \pm 0.008) \times 10^{-6}$		$(1.0708 \pm 0.0003) \times 10^{-6}$
$\alpha=\{2/3,3,4\}$		$(2.904 \pm 0.009) \times 10^{-6}$		$(1.255 \pm 0.002) \times 10^{-6}$	$(9.848 \pm 0.004) \times 10^{-7}$
$\alpha=\{2,3,4\}$			$(2.599 \pm 0.008) \times 10^{-6}$	$(5.32 \pm 0.04) \times 10^{-7}$	$(1.0289 \pm 0.0005) \times 10^{-6}$
$\alpha = \{0, 2/3, 2, 3\}$	$(-1.525\pm0.001)\times10^{-4}$	$(2.045 \pm 0.002) \times 10^{-4}$	$(-7.061 \pm 0.004) \times 10^{-5}$	$(1.7369 \pm 0.0005) \times 10^{-5}$	
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### CBC r-mode r-mode

![](_page_19_Picture_6.jpeg)