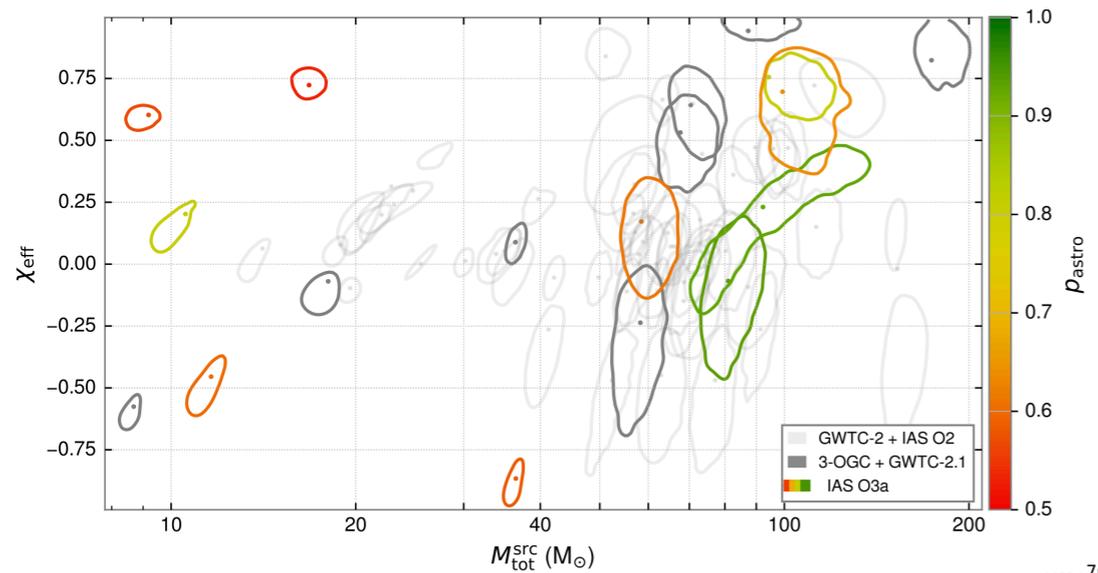
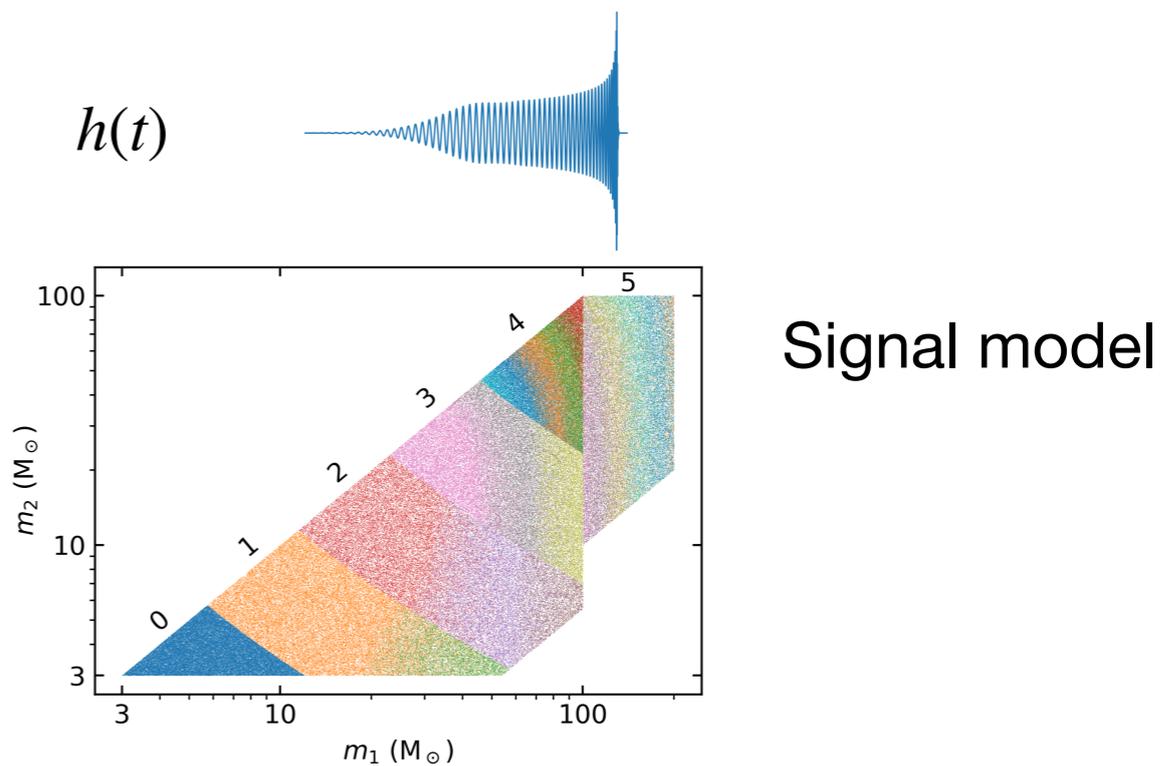
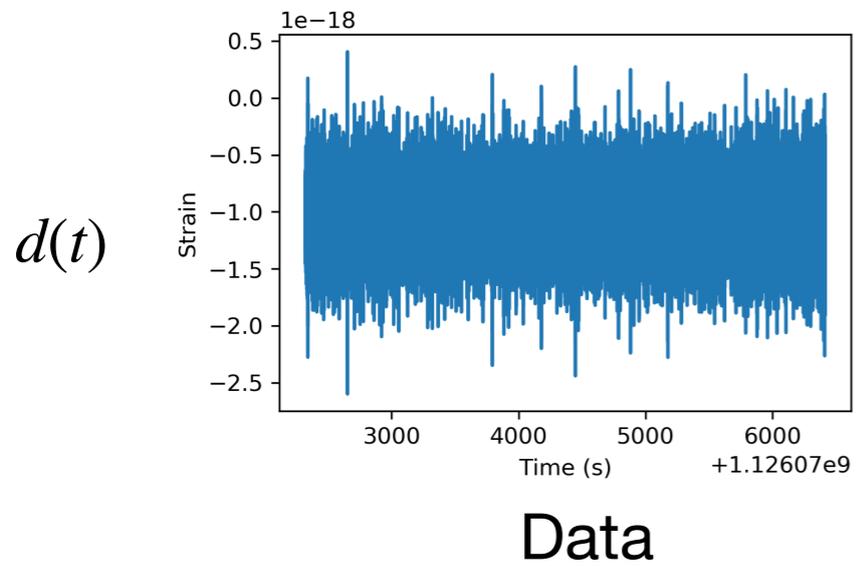


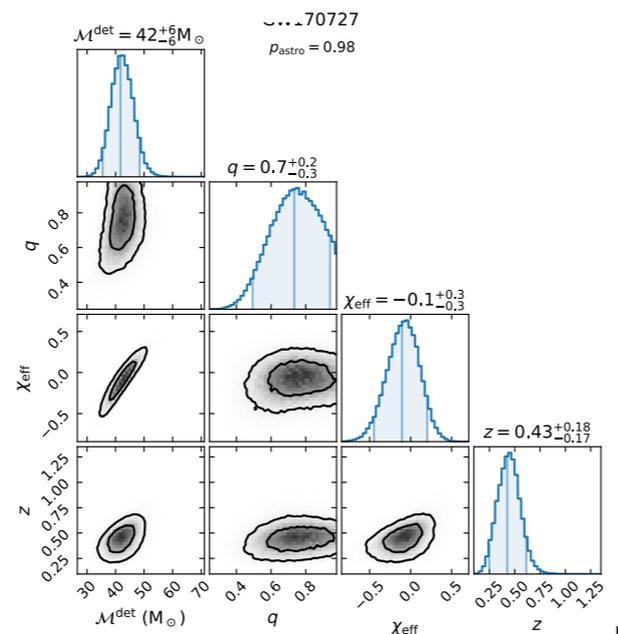
Looking for mergers in gravitational wave data: results, lessons, and future outlook

Tejaswi Venumadhav

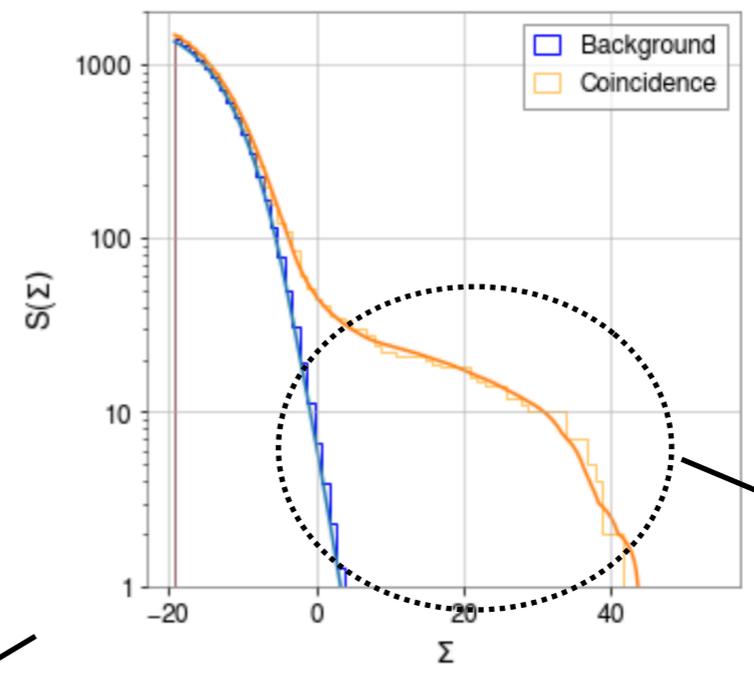
University of California
Santa Barbara



Population inference



Parameter estimation



Summary statistic + Triggers

Candidates

GW from binaries: source properties

Amplitude

$$h_{ij} \sim \frac{1}{R} \ddot{q}_{ij} \sim 10^{-21} \left(\frac{\omega_{GW}}{100 \text{ Hz}} \right)^{2/3} \left(\frac{M_c}{25 M_\odot} \right)^{5/3} \left(\frac{400 \text{ Mpc}}{R} \right)$$

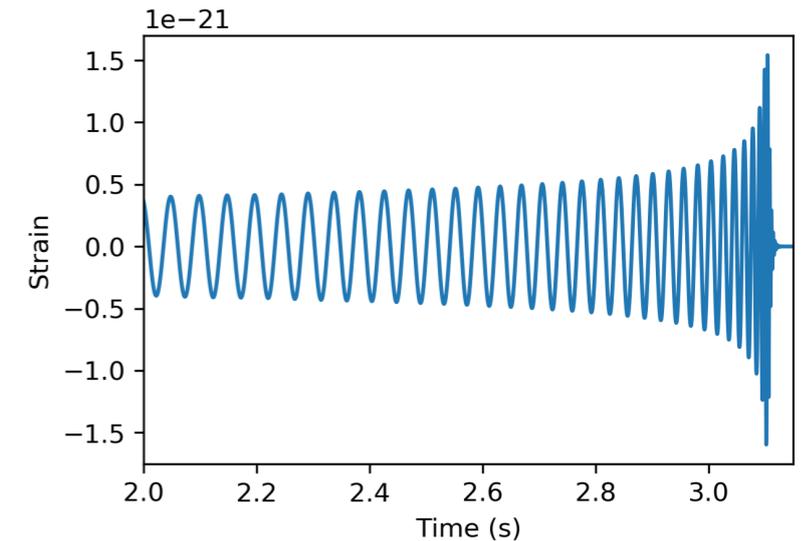
$$M_c = \frac{(m_1 m_2)^{3/5}}{(m_1 + m_2)^{1/5}} \quad \text{“chirp” mass}$$

Frequency

$$\omega_{GW}(t) \sim M_c^{-5/8} (t_0 - t)^{-3/8} \quad \text{“Chirping” behavior}$$

$$t_0 - t_{\min} \sim \frac{1}{M_c^{5/3} \omega_{GW,\min}^{8/3}} \approx 0.5 \text{ s} \left(\frac{25 M_\odot}{M_c} \right)^{5/3} \left(\frac{20 \text{ Hz}}{\omega_{\min}} \right)^{8/3}$$

$$\omega_{\max} \sim 100 \text{ Hz} \left(\frac{50 M_\odot}{M} \right) \quad \text{Cutoff at ISCO}$$

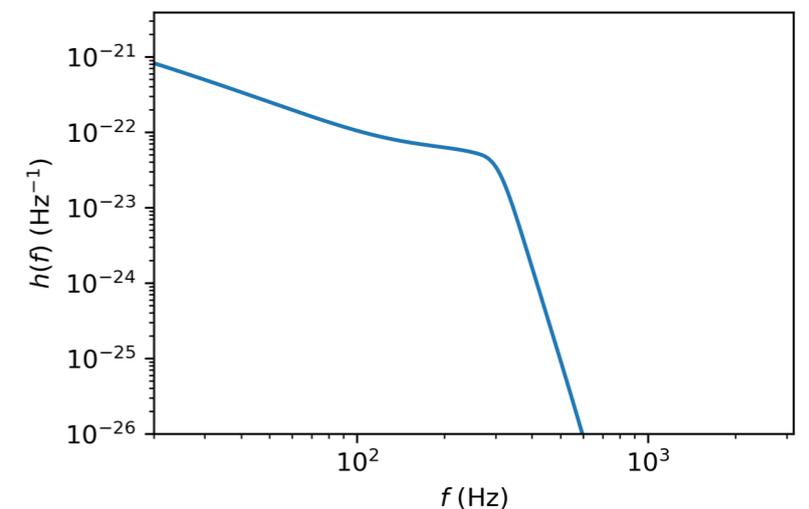


Orbital phase and waveforms

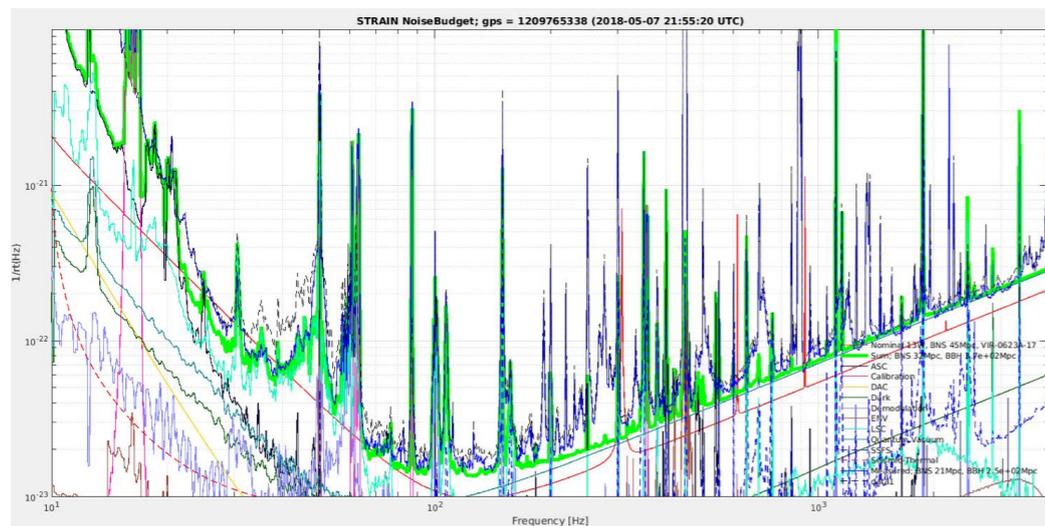
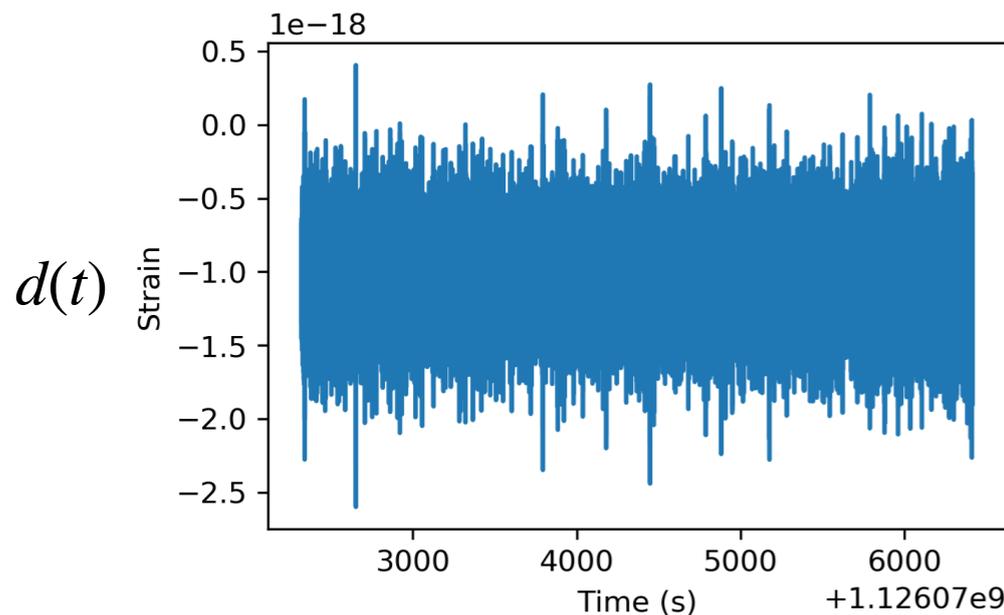
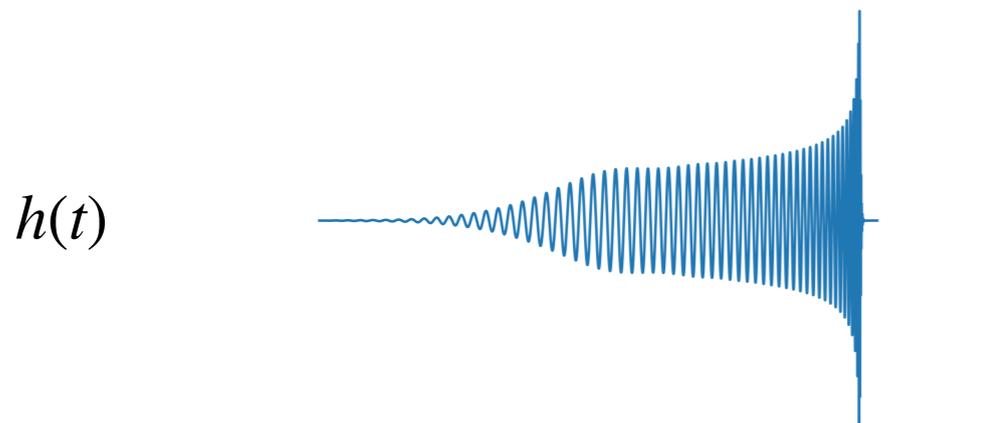
$$\phi(\omega_{GW}) \sim \int \omega_{GW}(t) dt \sim \left(\frac{1}{M_c \omega_{GW}} \right)^{5/3}$$

$$h(\omega_0) = \int dt h(t) e^{-i\omega_0 t} \sim \omega_0^{-7/6} e^{i\phi(\omega_0)}$$

SPA



Gaussian noise model: matched filtering



Credit: Giovanni Losurdo

Simplest case is stationary Gaussian random noise

$$\langle d(t + \tau)d(t) \rangle = C(\tau)$$

FT of $C(\tau)$ is the PSD $S(f) \sim \sigma^2(f) \equiv \langle |d(f)|^2 \rangle$

We have to estimate $S(f)$ ourselves!

Define ('whitening filter') $1/\sigma(f)$ such that filtered data

$$w(f) = \frac{d(f)}{\sigma(f)}$$

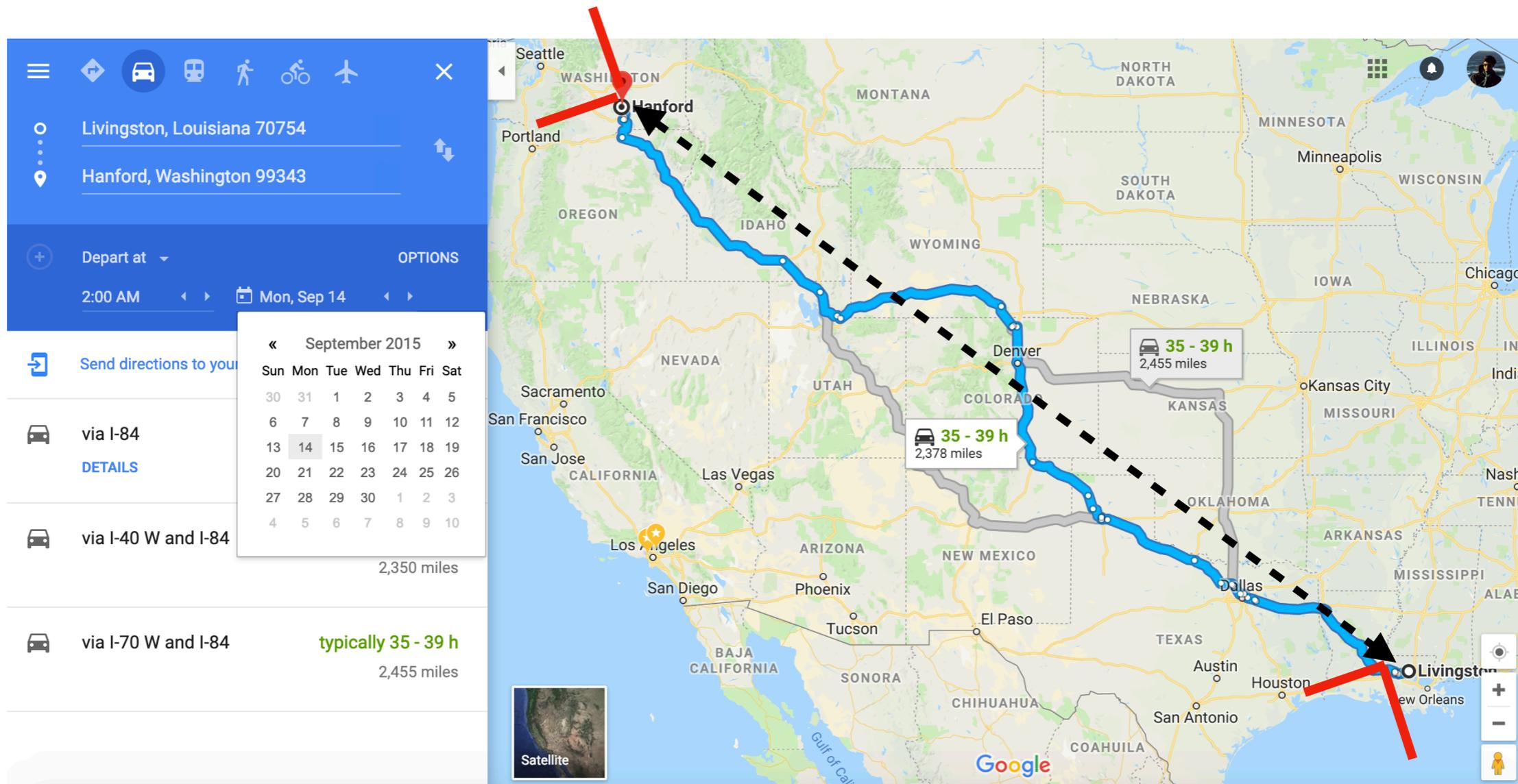
has white noise statistics

The matched-filtering score is the convolution of the whitened data with the whitened template

$$\rho = \frac{\sum_f d(f)h^*(f)}{\sigma^2(f)} \left[\frac{\sum_f |h(f)|^2}{\sigma^2(f)} \right]^{1/2}$$

Behaves like SNR

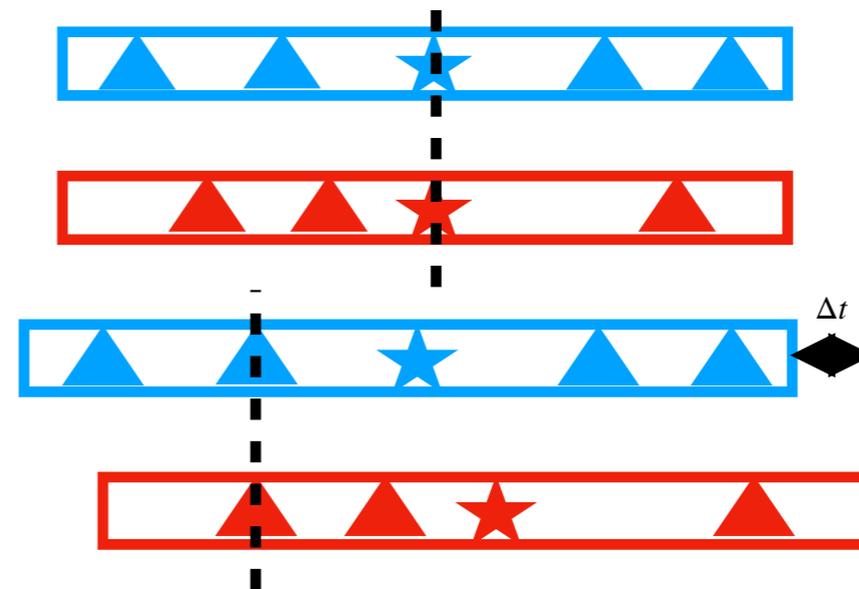
Collect coincident and background events



Shift data stream by $t > 10$ ms
to estimate background

Zero lag

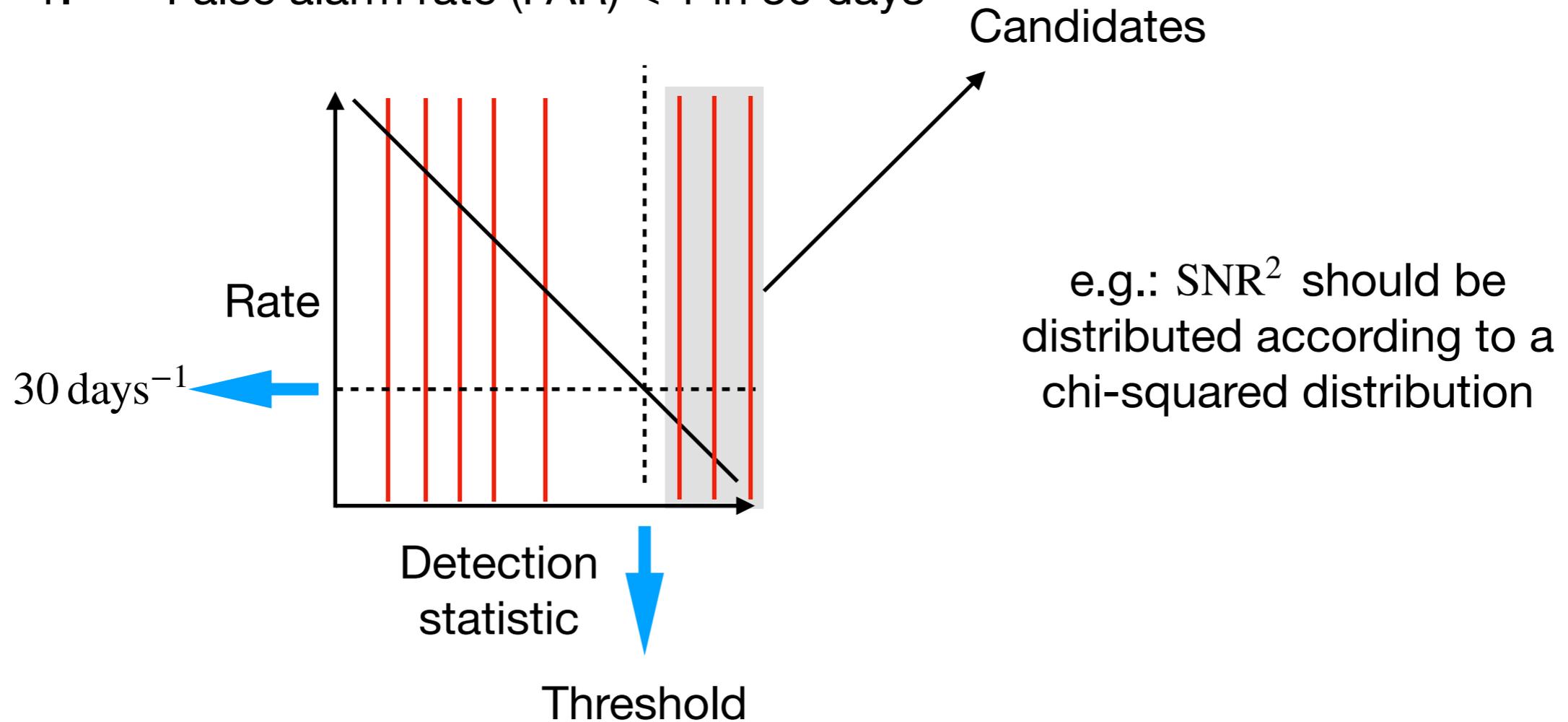
Time slide



Credit: Ian Harry

Detection criteria (O2)

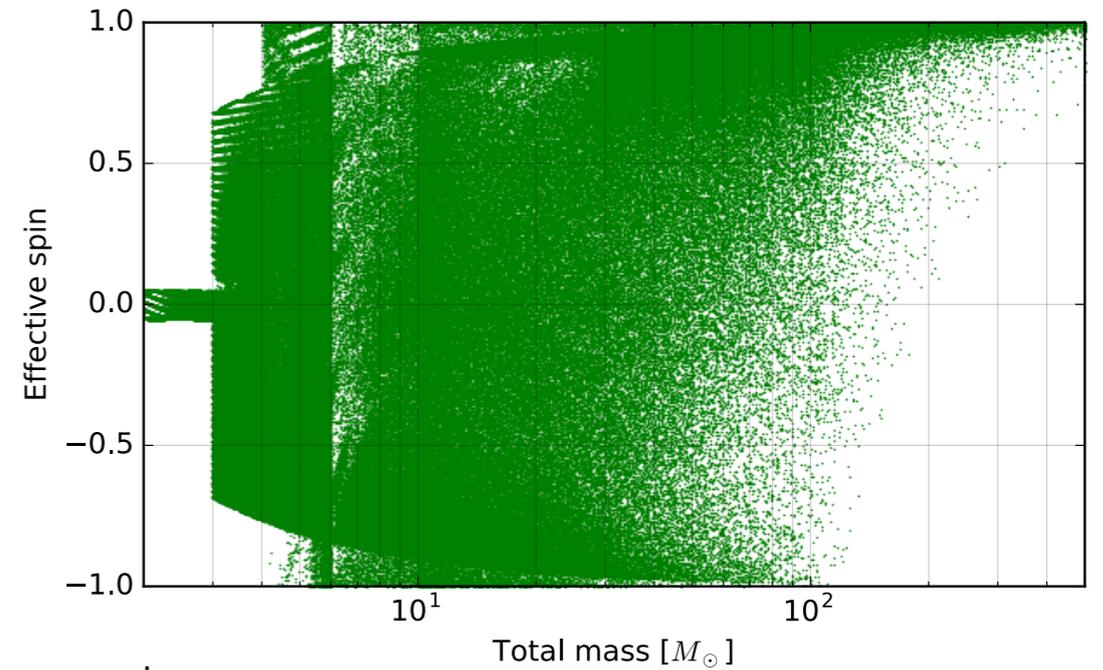
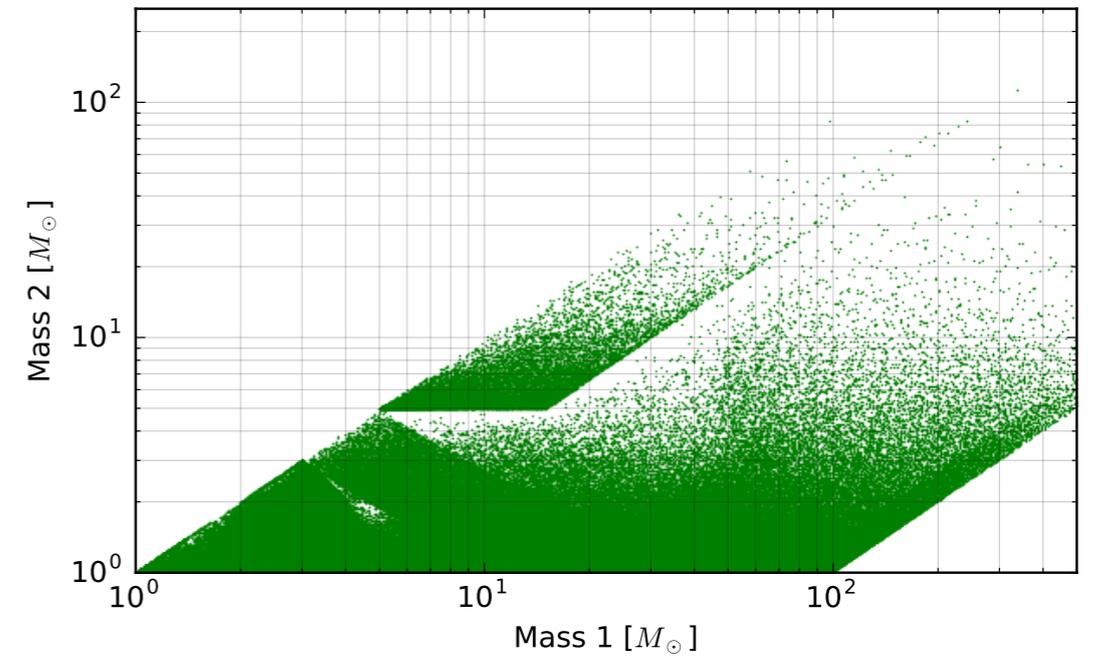
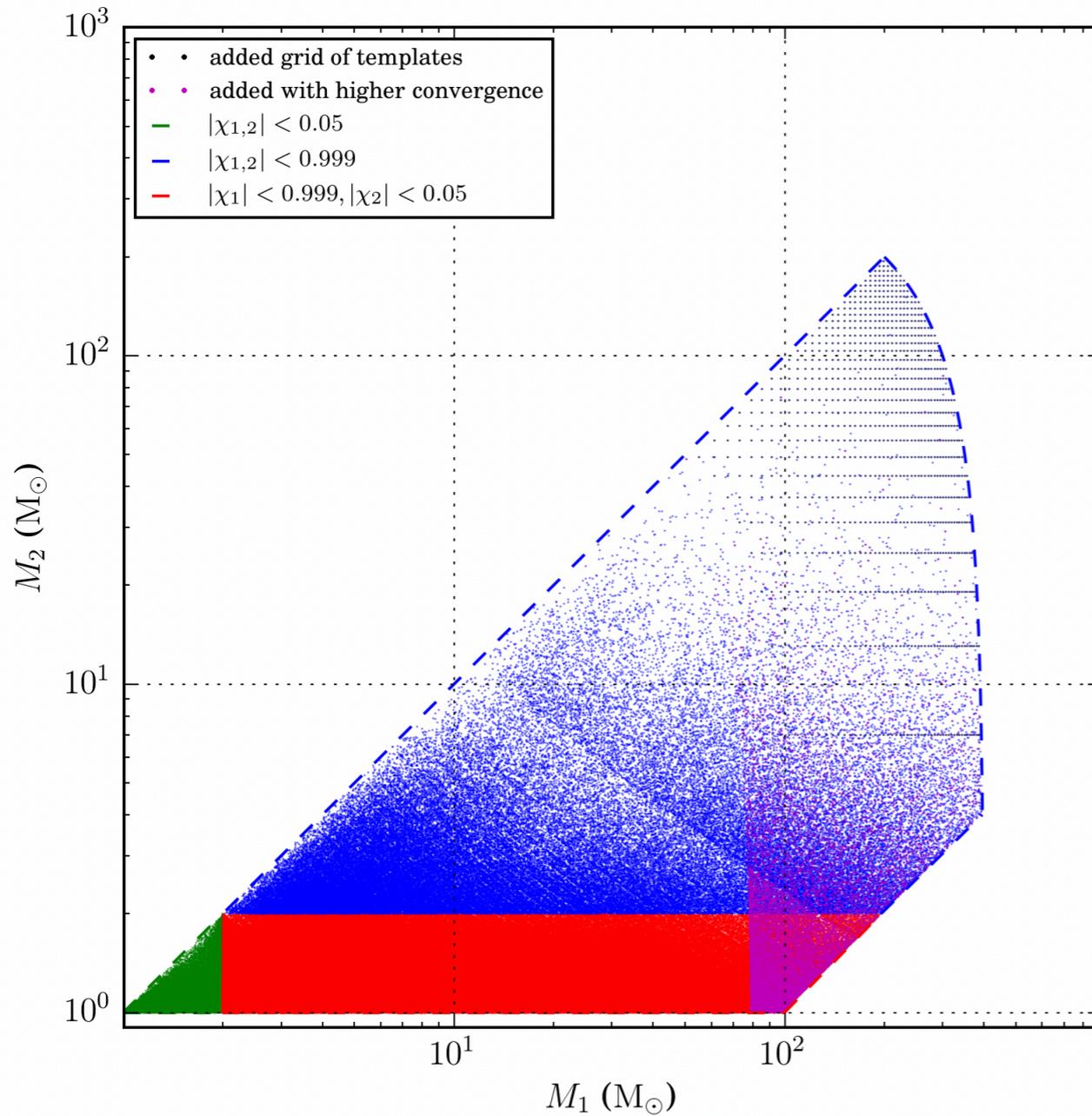
1. False alarm rate (FAR) < 1 in 30 days



2.
$$p_{\text{astro}} = \frac{R(\text{candidate} | \mathcal{S})}{R(\text{candidate} | \mathcal{N}) + R(\text{candidate} | \mathcal{S})} > 0.5$$

Abbott et. al. (2018)
arxiv:1811.12907

Template bank



$$\chi_{\text{eff}} = \frac{m_1 \chi_{1,z} + m_2 \chi_{2,z}}{m_1 + m_2}$$

Canton and Harry (2017)

Sachdev et. al., (2019)
arxiv: 1901.08580

Template bank

Canton and Harry (2017)

$$z(h) = \sum_f \frac{d(f)h^*(f)}{\sigma^2(f)}$$

$$h(f) \sim f^{-7/6} e^{i\phi(f)}$$

$$\phi \sim \frac{1}{(M_c f_{\min})^{5/3}}$$

$$\frac{dN}{d \ln M_c} \sim (M_c f_{\min})^{-5/3}$$

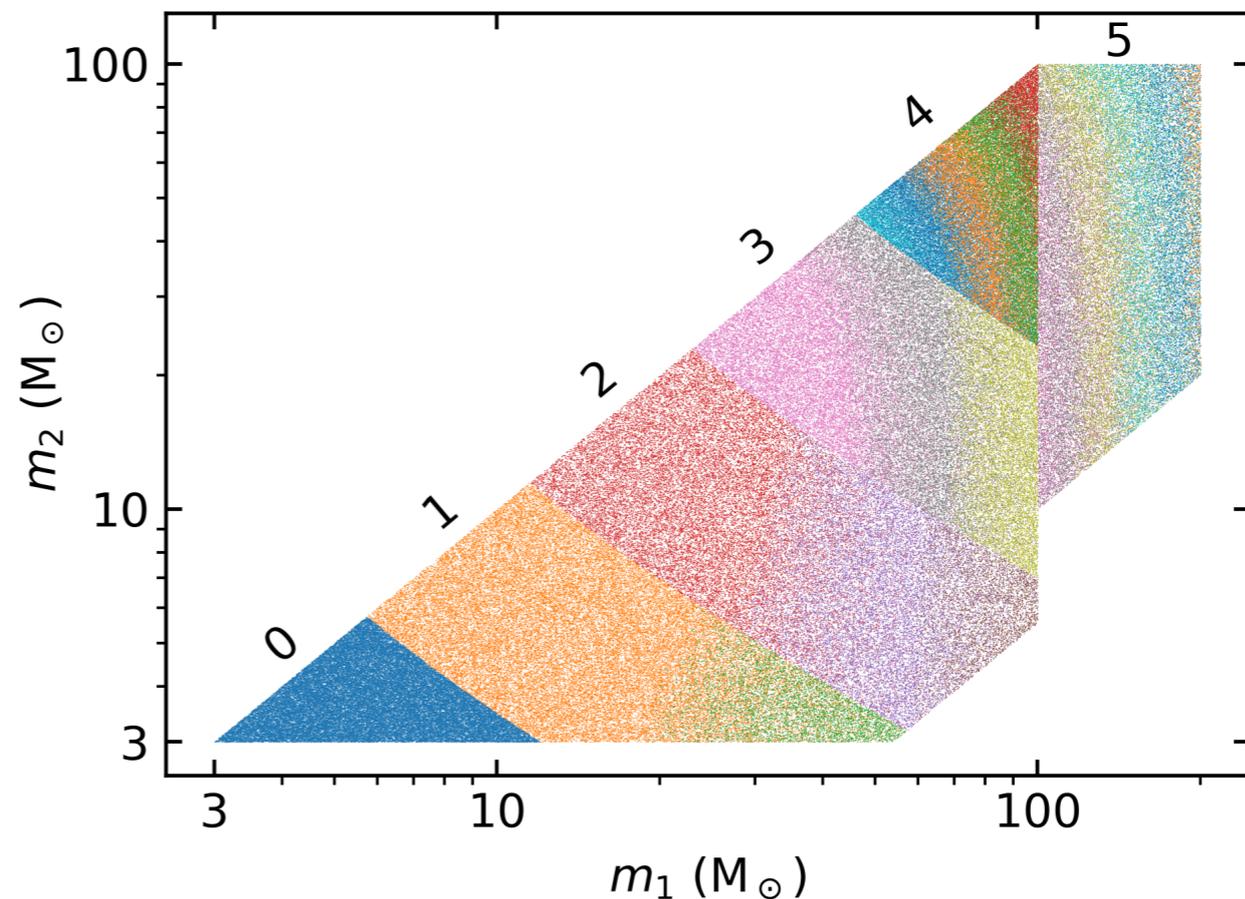
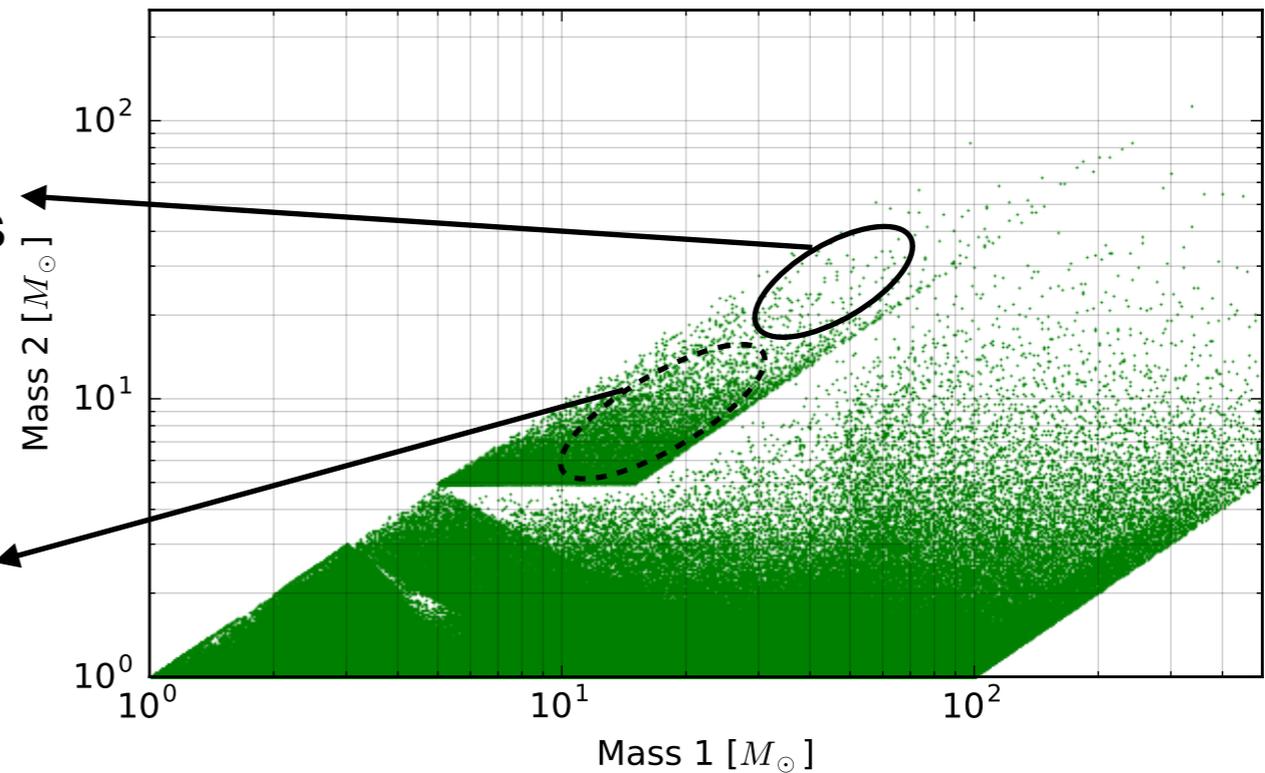
Solution:

Template prior
(Dent, T., and Veitch, J., (2014))

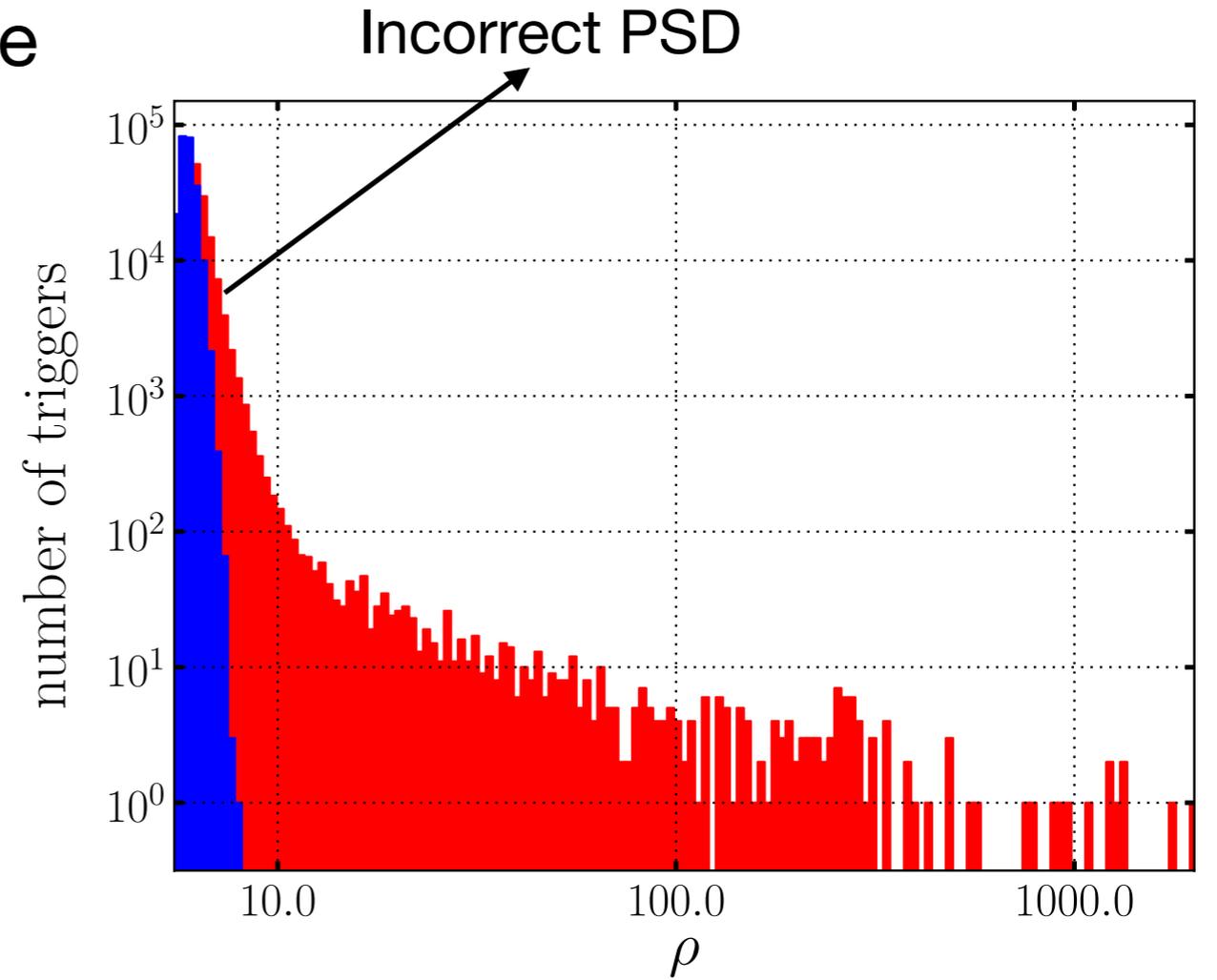
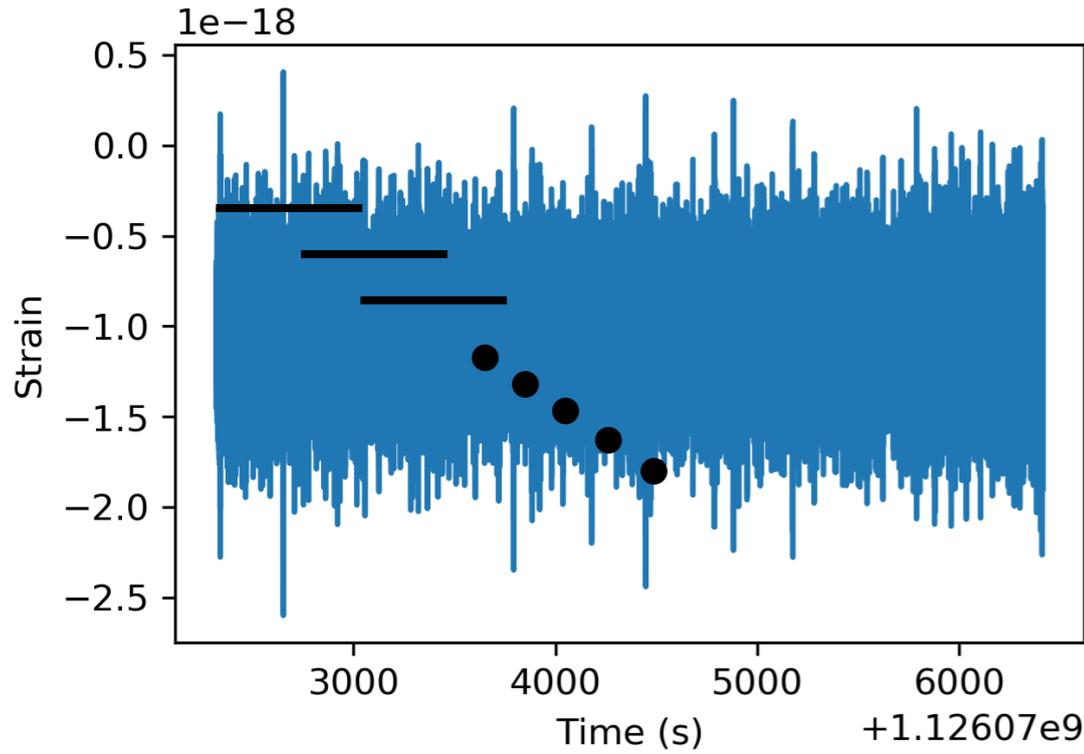
Multiple banks (Roulet et. al., (2019), Olsen et. al., (2022))

Most
detections

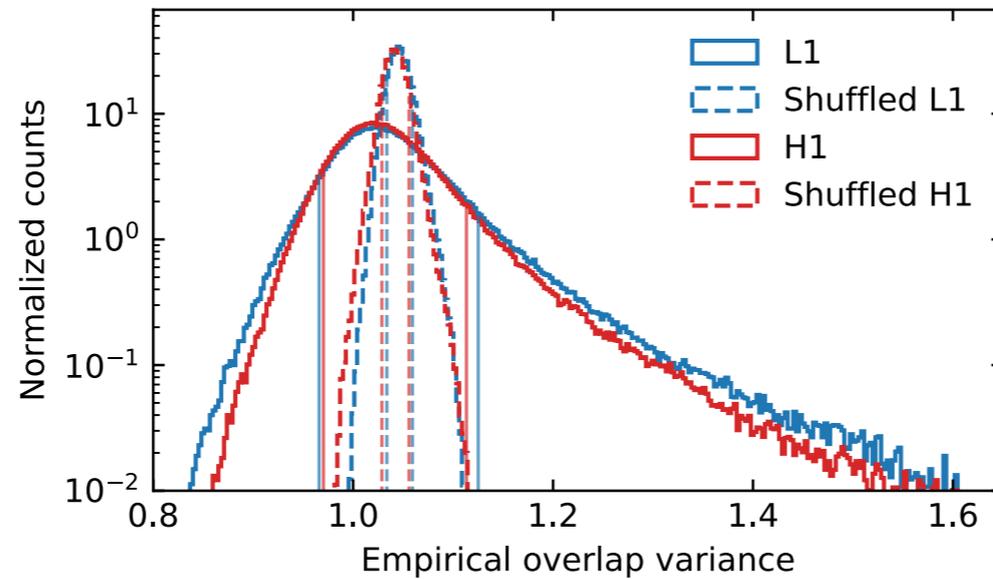
??



Systematic: non-stationary noise



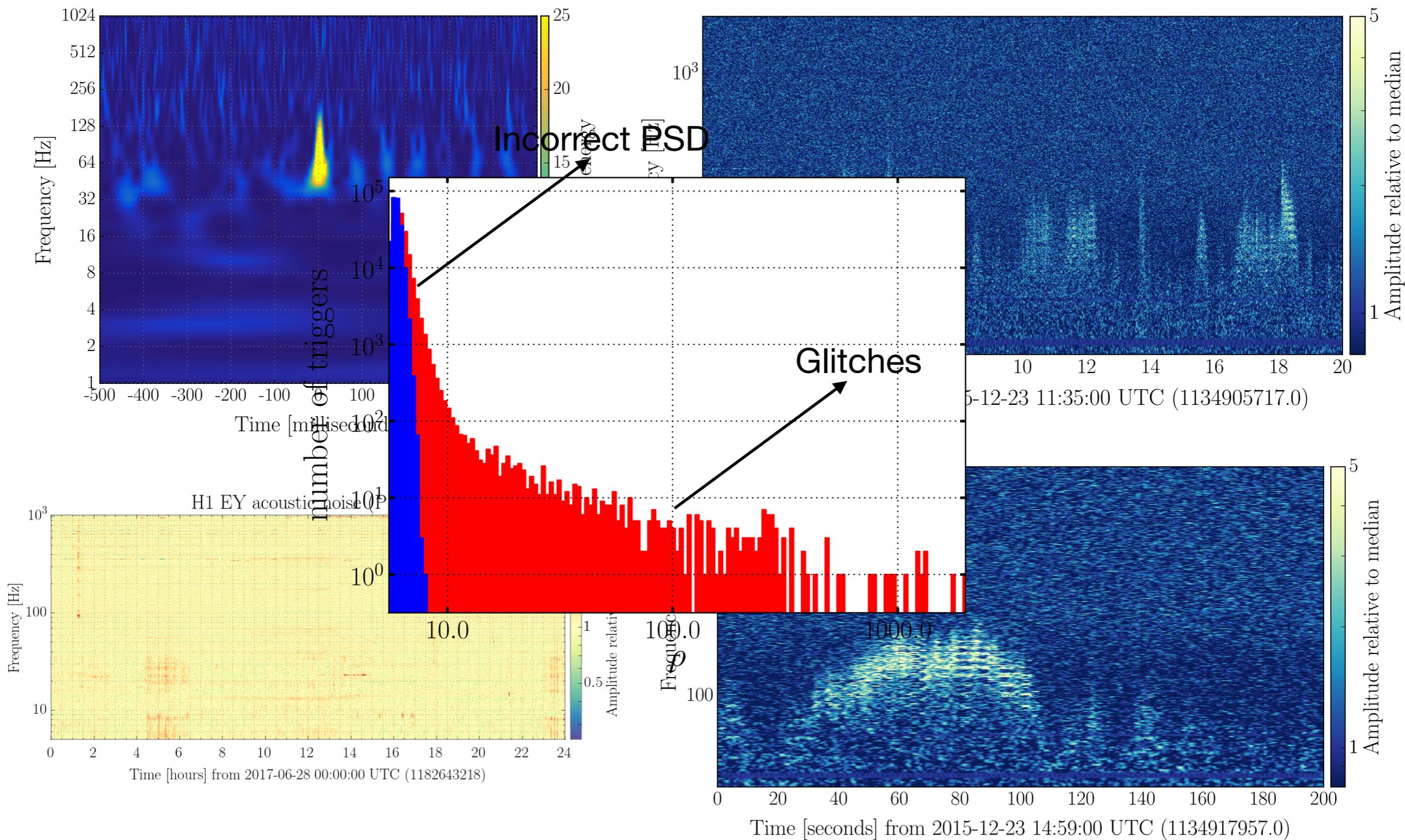
$$\frac{Z}{\sigma_Z} = \frac{\sum_f \frac{h^*(f)d(f)}{\sigma^2(f)}}{\sqrt{\sum_f \frac{|h(f)|^2}{\sigma^2(f)}}$$



$$\langle d(t + \tau)d(t) \rangle = C(\tau) \longrightarrow \langle d(t + \tau/2)d(t - \tau/2) \rangle = C(\tau, t) \approx C(\tau)f(t)$$

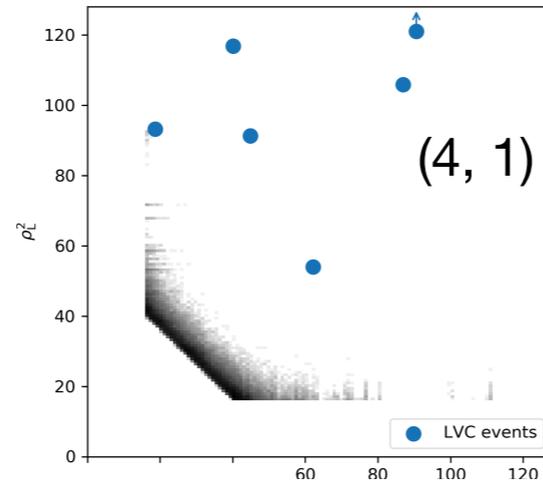
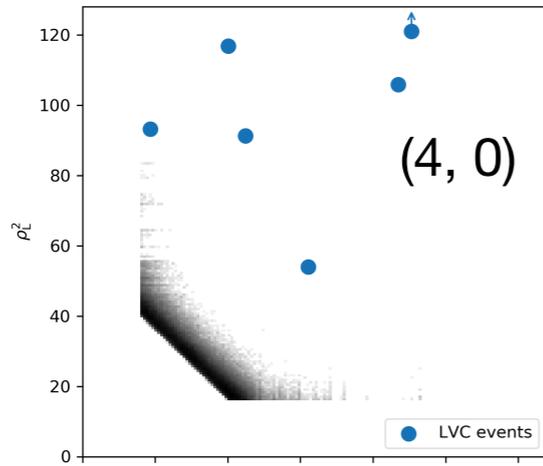
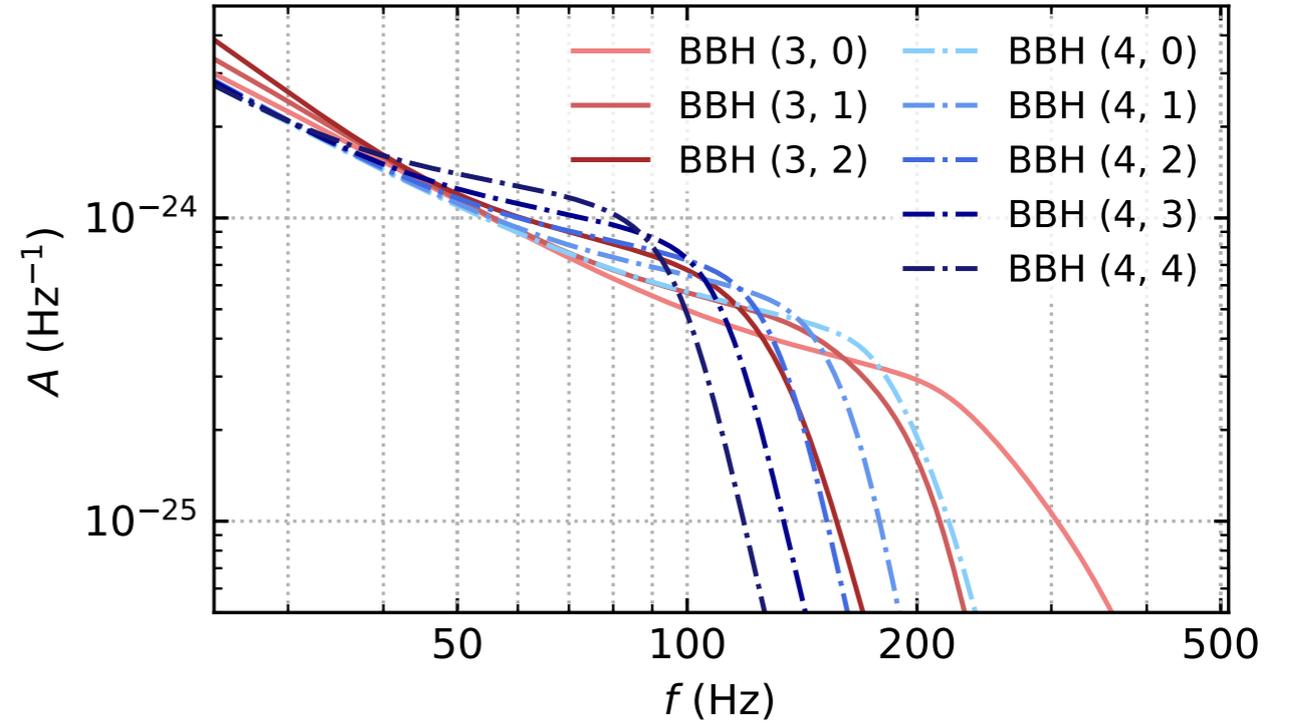
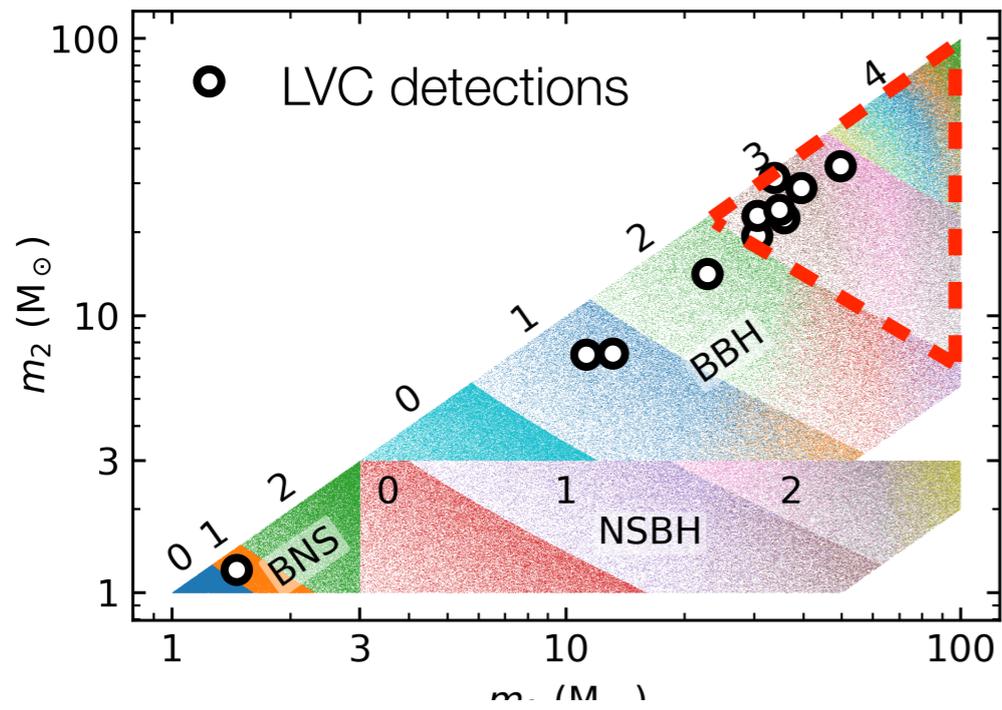
Glitches

$$z(h) = \sum_f \frac{d(f)h^*(f)}{\sigma^2(f)}$$

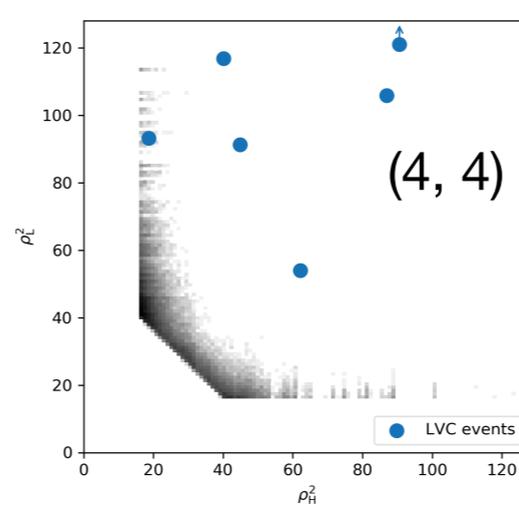
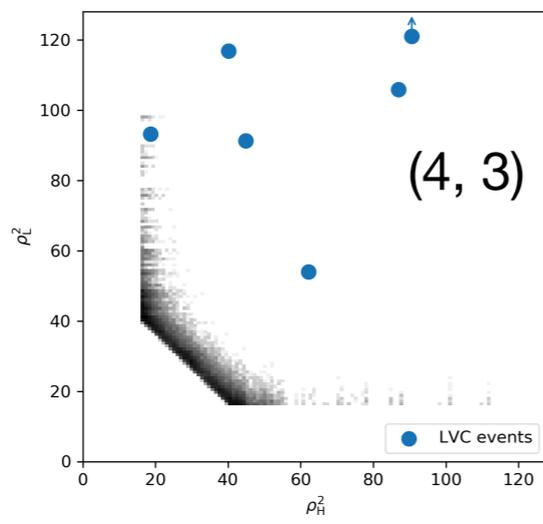
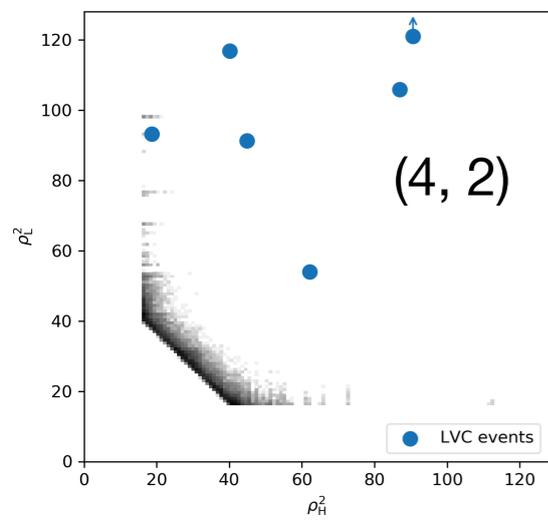


Glitches

$$z(h) = \sum_f \frac{d(f)h^*(f)}{\sigma^2(f)} \longrightarrow p(z)$$



Vary with detector and parameters



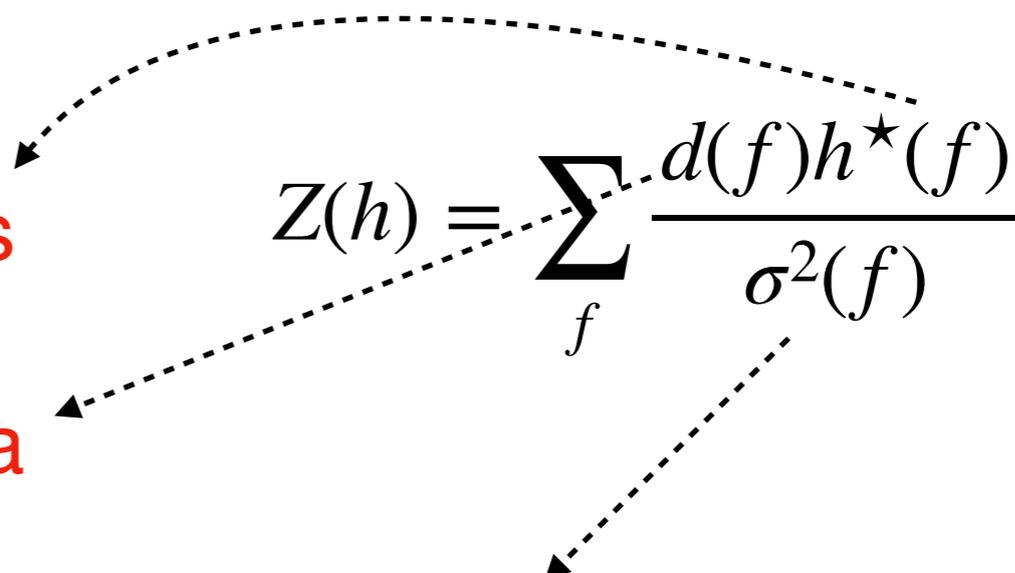
Candidate lists (O3)

Name	Inst.	MBTA			GstLAL			PyCBC			PyCBC-BBH		
		FAR (yr ⁻¹)	SNR	p_{astro}	FAR (yr ⁻¹)	SNR	p_{astro}	FAR (yr ⁻¹)	SNR	p_{astro}	FAR (yr ⁻¹)	SNR	p_{astro}
GW190403.051519	HL	--	--	--	--	--	--	--	--	7.7	8.0	0.61	
GW190408.181802	HLV	8.7×10^{-5}	14.4	1.00	$< 1.0 \times 10^{-5}$	14.7	1.00	2.5×10^{-4}	13.1	1.00	$< 1.2 \times 10^{-4}$	13.7	1.00
GW190412	HLV	$< 1.0 \times 10^{-5}$	18.2	1.00	$< 1.0 \times 10^{-5}$	19.0	1.00	$< 1.1 \times 10^{-4}$	17.4	1.00	$< 1.2 \times 10^{-4}$	17.9	1.00
GW190413.052954	HL	--	--	--	--	--	--	<i>170</i>	<i>8.5</i>	<i>0.13</i>	0.82	8.5	0.93
GW190413.134308	HLV	0.34	10.3	0.99	<i>39</i>	<i>10.1</i>	<i>0.04</i>	<i>21</i>	<i>9.3</i>	<i>0.48</i>	0.18	8.9	0.99
GW190421.213856	HL	1.2	9.7	0.99	0.0028	10.5	1.00	5.9	10.1	0.75	0.014	10.1	1.00
GW190425	LV	--	--	--	0.034 [†]	12.9	0.78	--	--	--	--	--	--
GW190426.190642	HL	--	--	--	--	--	--	--	--	--	4.1	9.6	0.75
GW190503.185404	HLV	0.013	12.8	1.00	$< 1.0 \times 10^{-5}$	12.0	1.00	0.038	12.2	1.00	0.0026	12.2	1.00
GW190512.180714	HLV	0.038	11.7	0.99	$< 1.0 \times 10^{-5}$	12.2	1.00	1.1×10^{-4}	12.4	1.00	$< 1.1 \times 10^{-4}$	12.4	1.00
GW190513.205428	HLV	0.11	13.0	0.99	1.3×10^{-5}	12.3	1.00	<i>19</i>	<i>11.6</i>	<i>0.49</i>	0.044	11.8	1.00
GW190514.065416	HL	--	--	--	<i>450</i>	<i>8.3</i>	<i>0.00</i>	--	--	--	2.8	8.4	0.76
GW190517.055101	HLV	0.11	11.3	1.00	0.0045	10.8	1.00	0.0095	10.4	1.00	3.5×10^{-4}	10.3	1.00
GW190519.153544	HLV	7.0×10^{-5}	13.7	1.00	$< 1.0 \times 10^{-5}$	12.4	1.00	$< 1.0 \times 10^{-4}$	13.2	1.00	$< 1.1 \times 10^{-4}$	13.2	1.00
GW190521	HLV	0.042	13.0	0.96	0.20	13.3	0.79	0.44	13.7	0.96	0.0013	13.6	1.00
GW190521.074359	HL	$< 1.0 \times 10^{-5}$	22.2	1.00	$< 1.0 \times 10^{-5}$	24.4	1.00	$< 1.8 \times 10^{-5}$	24.0	1.00	$< 2.3 \times 10^{-5}$	24.0	1.00
GW190527.092055	HL	--	--	--	0.23	8.7	0.85	--	--	--	<i>19</i>	<i>8.4</i>	<i>0.33</i>
GW190602.175927	HLV	3.0×10^{-4}	12.6	1.00	$< 1.0 \times 10^{-5}$	12.3	1.00	0.29	11.9	0.98	0.013	11.9	1.00
GW190620.030421	LV	--	--	--	0.011 [†]	10.9	0.99	--	--	--	--	--	--
GW190630.185205	LV	--	--	--	$< 1.0 \times 10^{-5}$	15.2	1.00	--	--	--	0.24	15.1	1.00
GW190701.203306	HLV	35	11.3	0.87	0.0057	11.7	0.99	0.064	11.9	0.99	0.56	11.7	1.00
GW190706.222641	HLV	0.0015	11.9	1.00	5.0×10^{-5}	12.5	1.00	3.7×10^{-4}	11.7	1.00	0.34	12.6	1.00
GW190707.093326	HL	0.032	12.6	1.00	$< 1.0 \times 10^{-5}$	13.2	1.00	$< 1.0 \times 10^{-5}$	13.0	1.00	$< 1.9 \times 10^{-5}$	13.0	1.00
GW190708.232457	LV	--	--	--	3.1×10^{-4} [†]	13.1	1.00	--	--	--	--	--	--
GW190719.215514	HL	--	--	--	--	--	--	--	--	--	0.63	8.0	0.92
GW190720.000836	HLV	0.094	11.6	1.00	$< 1.0 \times 10^{-5}$	11.5	1.00	1.4×10^{-4}	10.6	1.00	$< 7.8 \times 10^{-5}$	11.4	1.00
GW190725.174728*	HLV	3.1	9.8	0.59	--	--	--	0.46	9.1	0.96	2.9	8.8	0.82
GW190727.060333	HLV	0.023	12.0	1.00	$< 1.0 \times 10^{-5}$	12.1	1.00	0.0056	11.4	1.00	2.0×10^{-4}	11.1	1.00
GW190728.064510	HLV	7.5×10^{-4}	13.1	1.00	$< 1.0 \times 10^{-5}$	13.4	1.00	$< 8.2 \times 10^{-5}$	13.0	1.00	$< 7.8 \times 10^{-5}$	13.0	1.00
GW190731.140936	HL	6.1	9.1	0.80	0.33	8.5	0.78	--	--	--	1.9	7.8	0.83
GW190803.022701	HLV	77	9.0	0.96	0.073	9.1	0.94	<i>81</i>	<i>8.7</i>	<i>0.17</i>	0.39	8.7	0.97
GW190805.211137	HL	--	--	--	--	--	--	--	--	--	0.63	8.3	0.95
GW190814	LV	$< 2.0 \times 10^{-4}$	20.4	1.00	$< 1.0 \times 10^{-5}$	22.2	1.00	0.17	19.5	1.00	--	--	--
GW190828.063405	HLV	$< 1.0 \times 10^{-5}$	15.2	1.00	$< 1.0 \times 10^{-5}$	16.3	1.00	$< 8.5 \times 10^{-5}$	13.9	1.00	$< 7.0 \times 10^{-5}$	15.9	1.00
GW190828.065509	HLV	0.16	10.8	0.96	3.5×10^{-5}	11.1	1.00	2.8×10^{-4}	10.5	1.00	1.1×10^{-4}	10.5	1.00
GW190910.112807	LV	--	--	--	0.0029 [†]	13.4	1.00	--	--	--	--	--	--
GW190915.235702	HLV	0.0055	12.7	1.00	$< 1.0 \times 10^{-5}$	13.0	1.00	6.8×10^{-4}	13.0	1.00	$< 7.0 \times 10^{-5}$	13.1	1.00
GW190916.200658*	HLV	6.9×10^3	8.2	0.66	<i>12</i>	<i>8.2</i>	<i>0.09</i>	--	--	--	4.7	7.9	0.64
GW190917.114630	HLV	--	--	--	0.66	9.5	0.77	--	--	--	--	--	--
GW190924.021846	HLV	0.0049	11.9	0.99	$< 1.0 \times 10^{-5}$	13.0	1.00	$< 8.2 \times 10^{-5}$	12.4	1.00	8.3×10^{-5}	12.5	1.00
GW190925.232845*	HV	<i>100</i>	<i>9.4</i>	<i>0.35</i>	--	--	--	<i>73</i>	<i>9.0</i>	<i>0.02</i>	0.0072	9.9	0.99
GW190926.050336*	HLV	--	--	--	1.1	9.0	0.54	--	--	--	<i>87</i>	<i>7.8</i>	<i>0.09</i>
GW190929.012149	HLV	2.9	10.3	0.64	0.16	10.1	0.87	<i>120</i>	<i>9.4</i>	<i>0.14</i>	<i>14</i>	<i>8.5</i>	<i>0.41</i>
GW190930.133541	HL	0.34	10.0	0.87	0.43	10.1	0.76	0.018	9.8	1.00	0.012	10.0	1.00

Name	Inst.	cWB			GstLAL			MBTA			PyCBC-broad			PyCBC-BBH		
		FAR (yr ⁻¹)	SNR	p_{astro}	FAR (yr ⁻¹)	SNR	p_{astro}	FAR (yr ⁻¹)	SNR	p_{astro}	FAR (yr ⁻¹)	SNR	p_{astro}	FAR (yr ⁻¹)	SNR	p_{astro}
GW191103.012549	HL	--	--	--	--	--	--	27	9.0	0.13	4.8	9.3	0.77	0.46	9.3	0.94
GW191105.143521	HLV	--	--	--	<i>24</i>	<i>10.0</i>	<i>0.07</i>	0.14	10.7	> 0.99	0.012	9.8	> 0.99	0.036	9.8	> 0.99
GW191109.010717	HL	< 0.0011	15.6	> 0.99	0.0010	15.8	> 0.99	1.8×10^{-4}	15.2	> 0.99	0.096	13.2	> 0.99	0.047	14.4	> 0.99
GW191113.071753	HLV	--	--	--	--	--	--	26	9.2	0.68	1.1×10^4	8.3	< 0.01	1.2×10^3	8.5	< 0.01
GW191126.115259	HL	--	--	--	<i>80</i>	<i>8.7</i>	<i>0.02</i>	<i>59</i>	<i>8.5</i>	<i>0.30</i>	<i>22</i>	<i>8.5</i>	<i>0.39</i>	<i>3.2</i>	<i>8.5</i>	<i>0.70</i>
GW191127.050227	HLV	--	--	--	<i>0.25</i>	<i>10.3</i>	<i>0.49</i>	1.2	9.8	0.73	<i>20</i>	<i>9.5</i>	<i>0.47</i>	4.1	8.7	0.74
GW191129.134029	HL	--	--	--	$< 1.0 \times 10^{-5}$	13.3	> 0.99	0.013	12.7	> 0.99	$< 2.6 \times 10^{-5}$	12.9	> 0.99	$< 2.4 \times 10^{-5}$	12.9	> 0.99
GW191204.110529	HL	--	--	--	<i>21</i>	<i>9.0</i>	<i>0.07</i>	1.3×10^4	8.1	< 0.01	980	8.9	< 0.01	3.3	8.9	0.74
GW191204.171526	HL	$< 8.7 \times 10^{-4}$	17.1	> 0.99	$< 1.0 \times 10^{-5}$	15.6	> 0.99	$< 1.0 \times 10^{-5}$	17.1	> 0.99	$< 1.4 \times 10^{-5}$	16.9	> 0.99	$< 1.2 \times 10^{-5}$	16.9	> 0.99
GW191215.223052	HLV	0.12	9.8	0.95	$< 1.0 \times 10^{-5}$	10.9	> 0.99	0.22	10.8	> 0.99	0.0016	10.3	> 0.99	0.28	10.2	> 0.99
GW191216.213338	HV	--	--	--	$< 1.0 \times 10^{-5}$	18.6	> 0.99	9.3×10^{-4}	17.9	> 0.99	0.0019	18.3	> 0.99	7.6×10^{-4}	18.3	> 0.99
GW191219.163120	HLV	--	--	--	--	--	--	--	--	--	4.0	8.9	0.82	--	--	--
GW191222.033537	HL	$< 8.9 \times 10^{-4}$	11.1	> 0.99	$< 1.0 \times 10^{-5}$	12.0	> 0.99	0.0099	10.8	> 0.99	0.0021	11.5	> 0.99	9.8×10^{-5}	11.5	> 0.99
GW191230.180458	HLV	0.050	10.3	0.95	0.13	10.3	0.87	8.1	9.8	0.40	52	9.6	0.29	0.42	9.9	0.96
GW200112.155838	LV	--	--	--	$< 1.0 \times 10^{-5}$ [†]	17.6	> 0.99	--	--	--	--	--	--	--	--	--
GW200115.042309	HLV	--	--	--	$< 1.0 \times 10^{-5}$	11.5	> 0.99	0.0055	11.2	> 0.99	$< 1.2 \times 10^{-4}$	10.8	> 0.99	--	--	--
GW200128.022011	HL	1.3	8.8	0.63	0.022	10.1	0.97	3.3	9.4	0.98	0.63	9.8	0.95	0.0043	9.9	> 0.99
GW200129.065458	HLV	--	--	--	$< 1.0 \times 10^{-5}$	26.5	> 0.99	--	--	--	$< 2.3 \times 10^{-5}$	16.3	> 0.99	$< 1.7 \times 10^{-5}$	16.2	> 0.99
GW200202.154313	HLV	--	--	--	$< 1.0 \times 10^{-5}$	11.3	> 0.99	--	--	--	--	--	--	0.025	10.8	> 0.99
GW200208.130117	HLV	--	--	--	0.0096	10.7	0.99	0.46	10.4	> 0.99	0.18	9.6	0.98	3.1×10^{-4}	10.8	> 0.99
GW200208.222617	HLV	--	--	--	<i>160</i>	<i>8.2</i>	<i>< 0.01</i>	<i>420</i>	<i>8.9</i>	<i>0.02</i>	--	--	--	4.8	7.9	0.70
GW200209.085452	HLV	--	--	--	0.046	10.0	0.95	12	9.7	0.97	550	9.2	0.04	1.2	9.2	0.89
GW200210.092254	HLV	--	--	--	<i>1.2</i>	<i>9.5</i>	<i>0.42</i>	--	--	--	17	8.9	0.53	7.7	8.9	0.54
GW200216.220804	HLV	--	--	--	0.35	9.4	0.77	2.4×10^3	8.8	0.02						

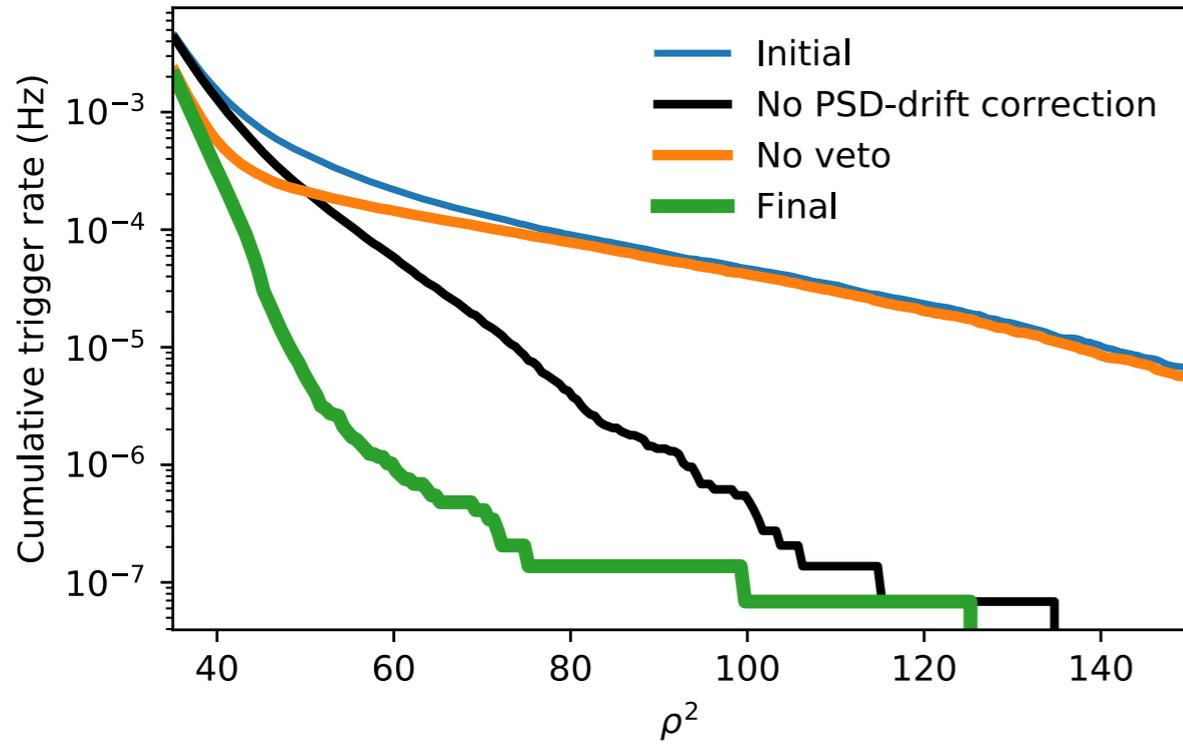
Our search

- Construct template banks
- Whiten and clean the data
- Generate triggers and correct for non-stationary noise
- Collect coincident triggers and veto remaining glitches
- Coherently analyze remaining coincident triggers

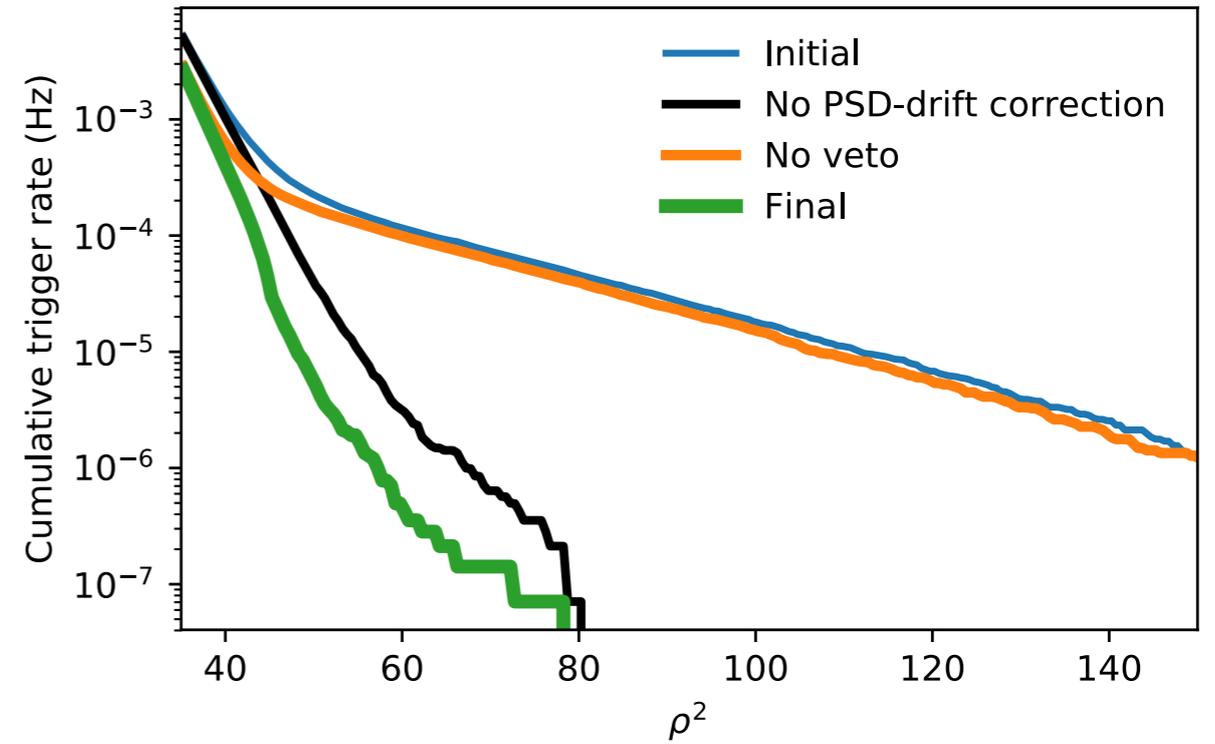
$$Z(h) = \sum_f \frac{d(f)h^*(f)}{\sigma^2(f)}$$


Impact of improved noise modeling and vetoes

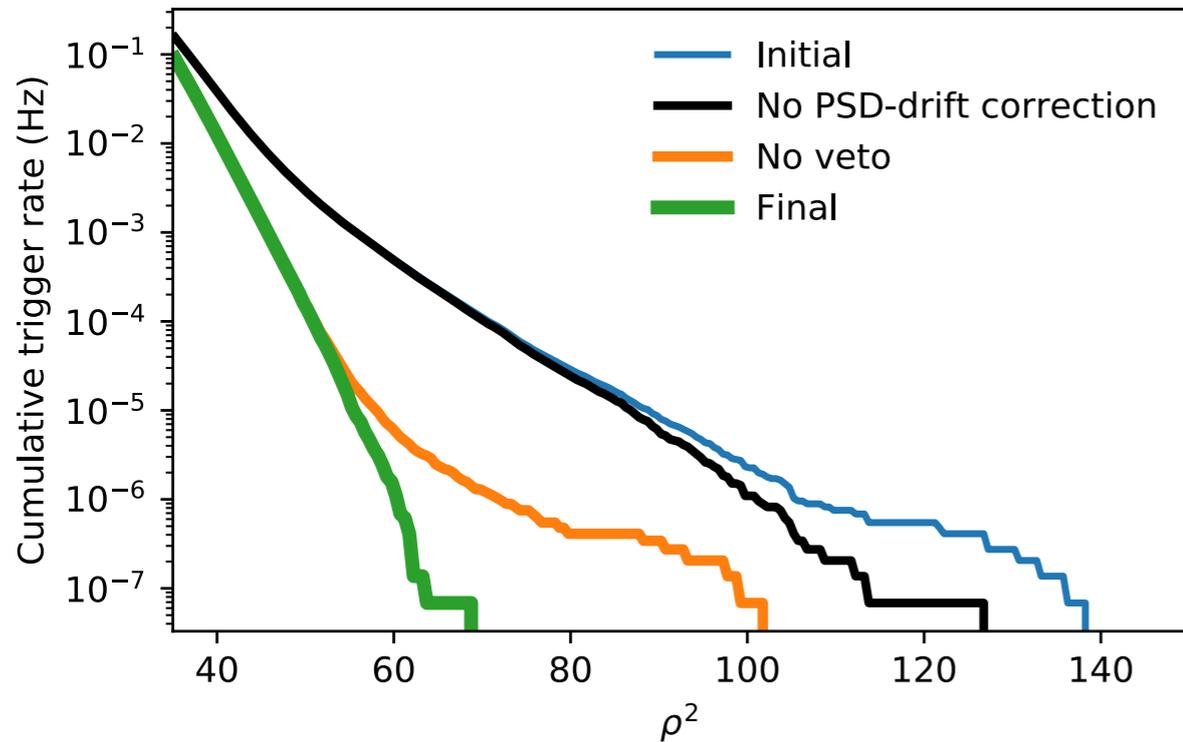
Hanford, bank BBH (3, 0) triggers in O2



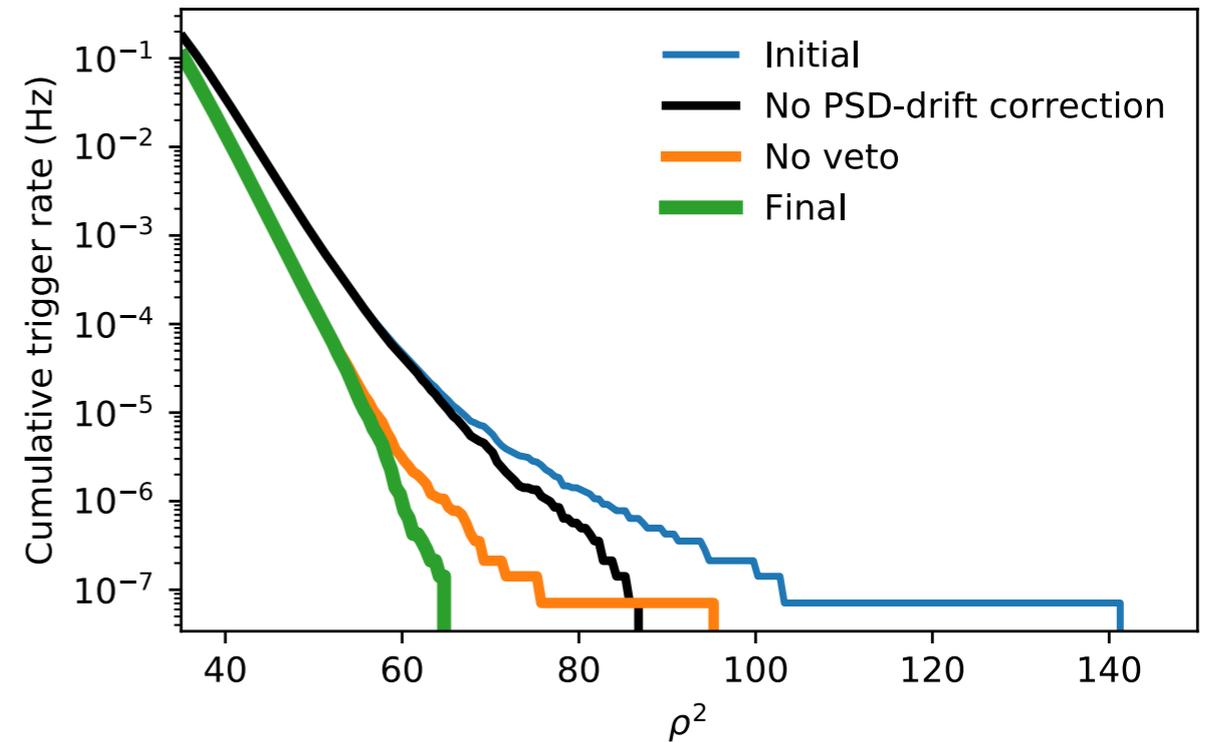
Livingston, bank BBH (3, 0) triggers in O2



Hanford, bank BBH (0, 0) triggers in O2



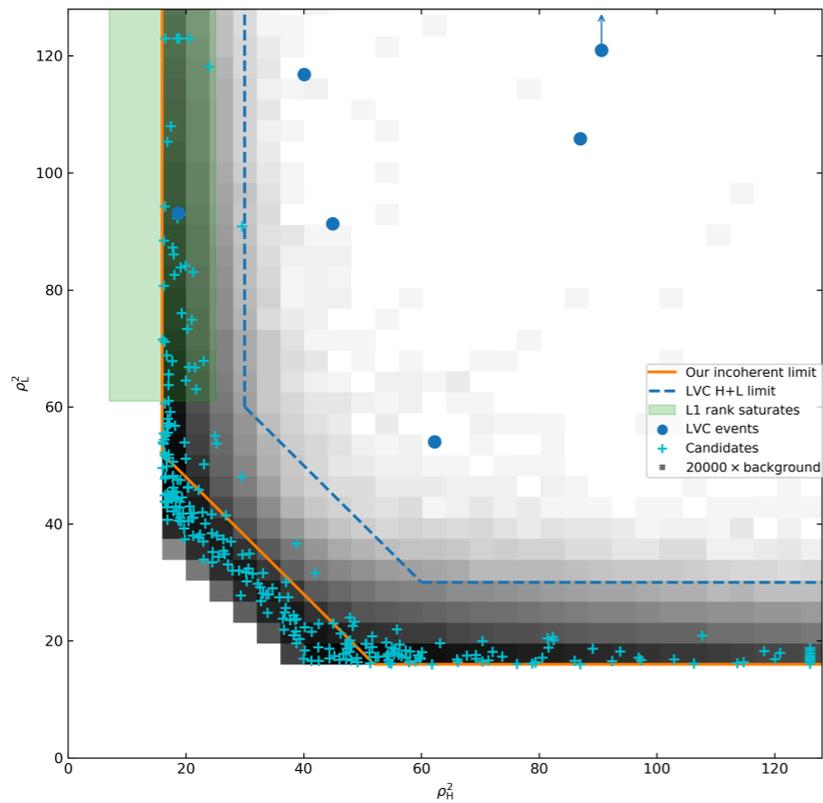
Livingston, bank BBH (0, 0) triggers in O2



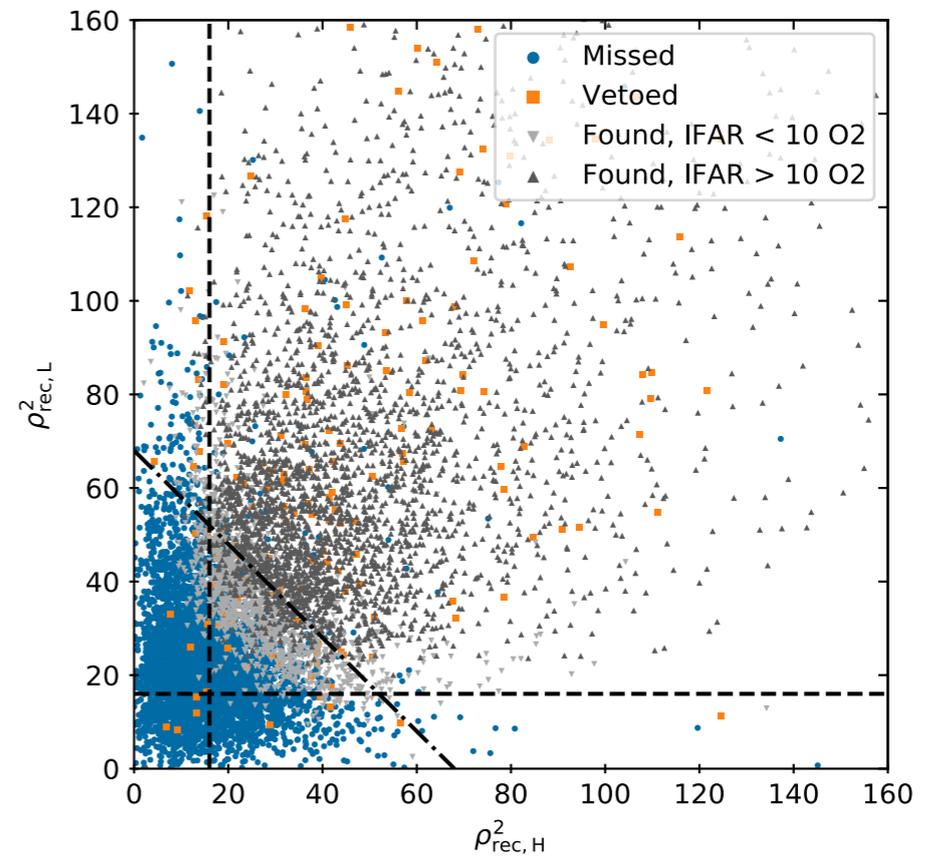
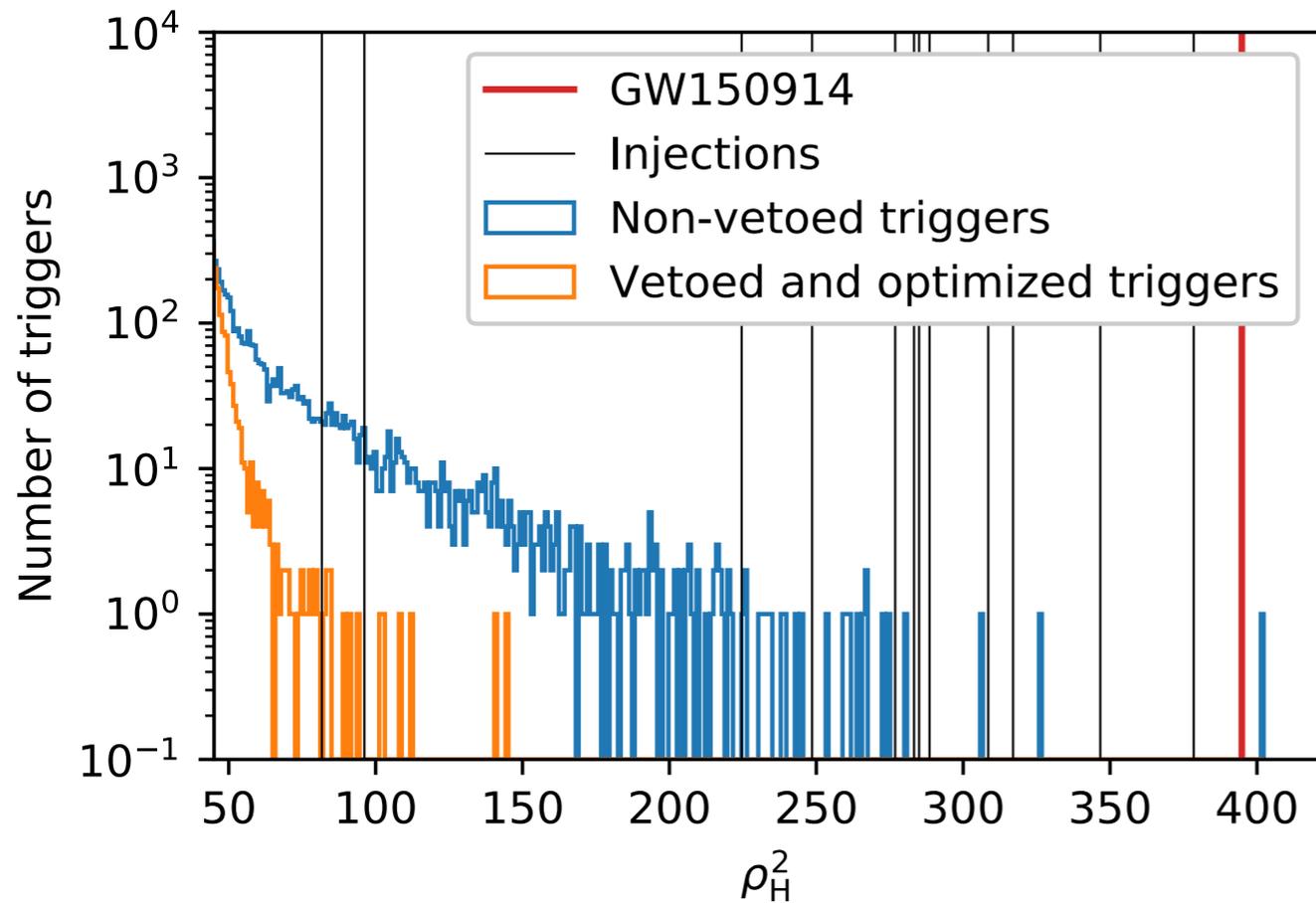
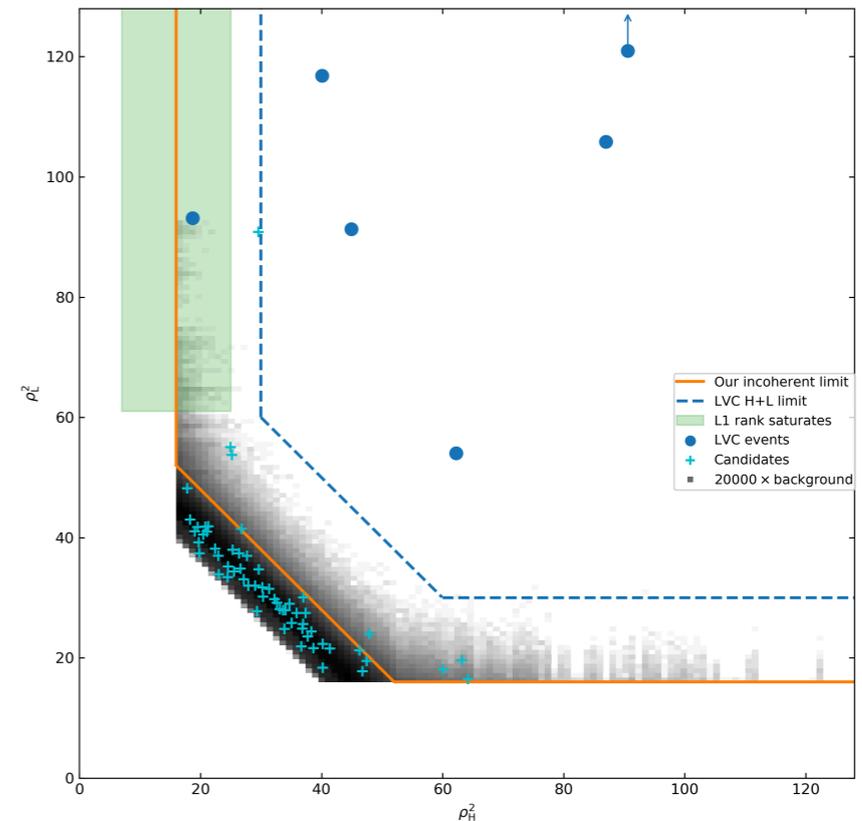
Effect of mitigating glitches

$$z(h) = \sum_f \frac{d(f)h^*(f)}{\sigma^2(f)}$$

Background
No vetoes

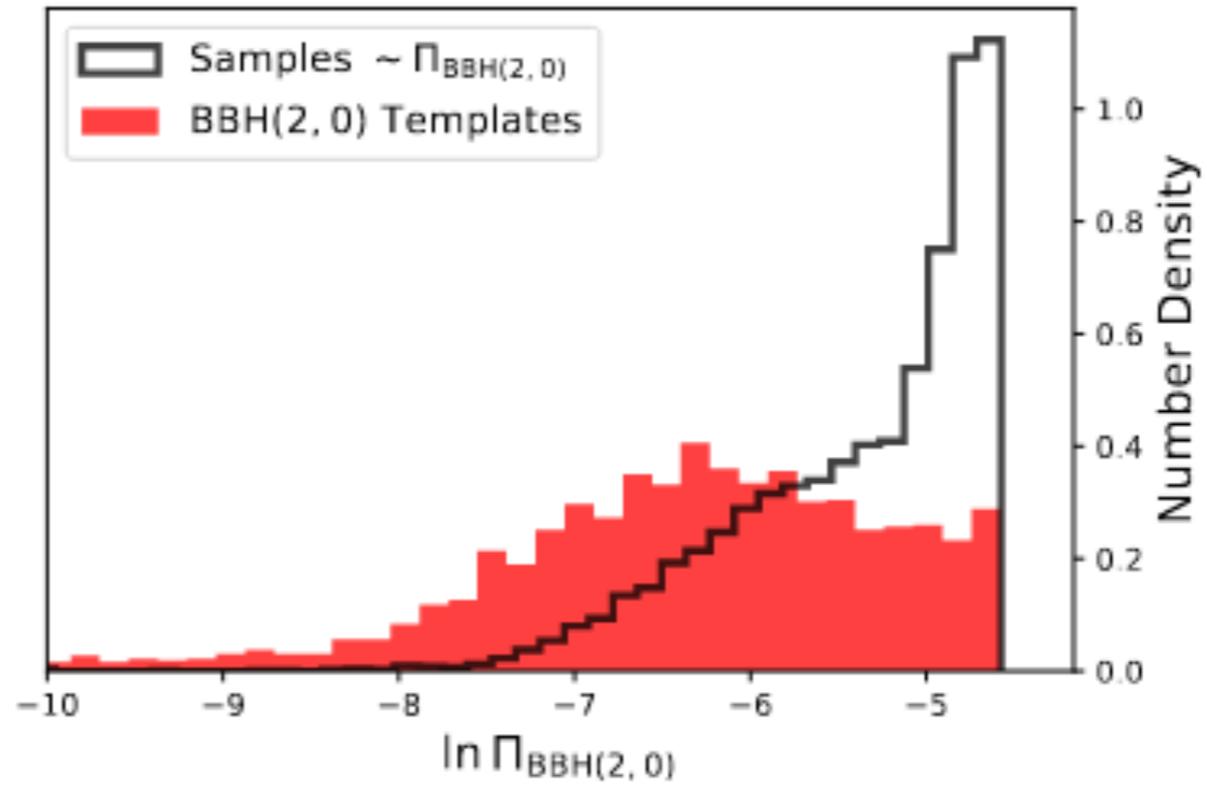


With
vetoes

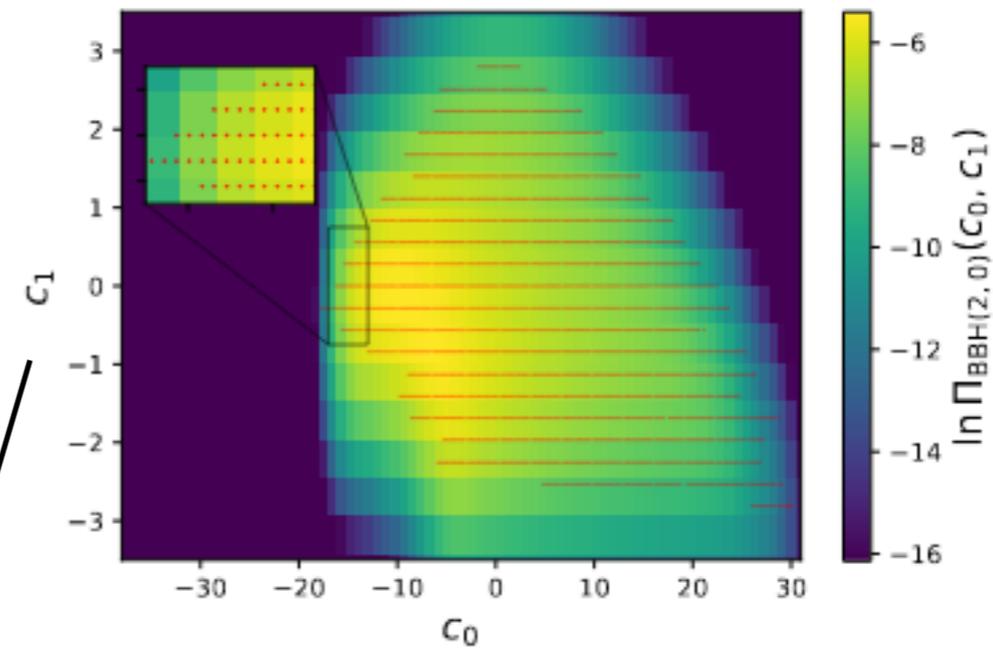


Injections

Template Prior



uniform c_α grid spacing over-represents regions of low prior relative to its coverage of the regions where physical templates are most likely. The template prior therefore decreases the impact of the look-elsewhere effect on the most physically probable regions of geometric grid space by raising the effective detection bar for templates in regions of grid space where the ratio of grid coordinate volume to physical parameter space volume is too large.



Π uniform in component masses and effective spin

See also Nitz et. al., (2017)

Geometric coordinates

Olsen, S., TV, et. al. (2022)

Multi-detector coherence

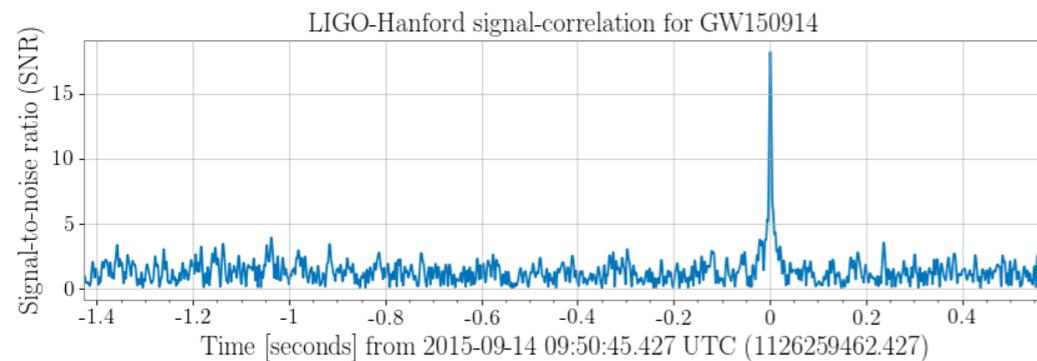
$$p(d|\mathcal{I}) = \int d\Pi(\mathcal{E}_m, \mathcal{E}_d) L(\mathcal{I}, \mathcal{E}_m, \mathcal{E}_d).$$

Prior

Likelihood

Extrinsic parameters:

- Inclination
- Orbital phase
- RA
- Dec
- Polarization angle
- Distance
- Merger time



Credit: (GWpy)

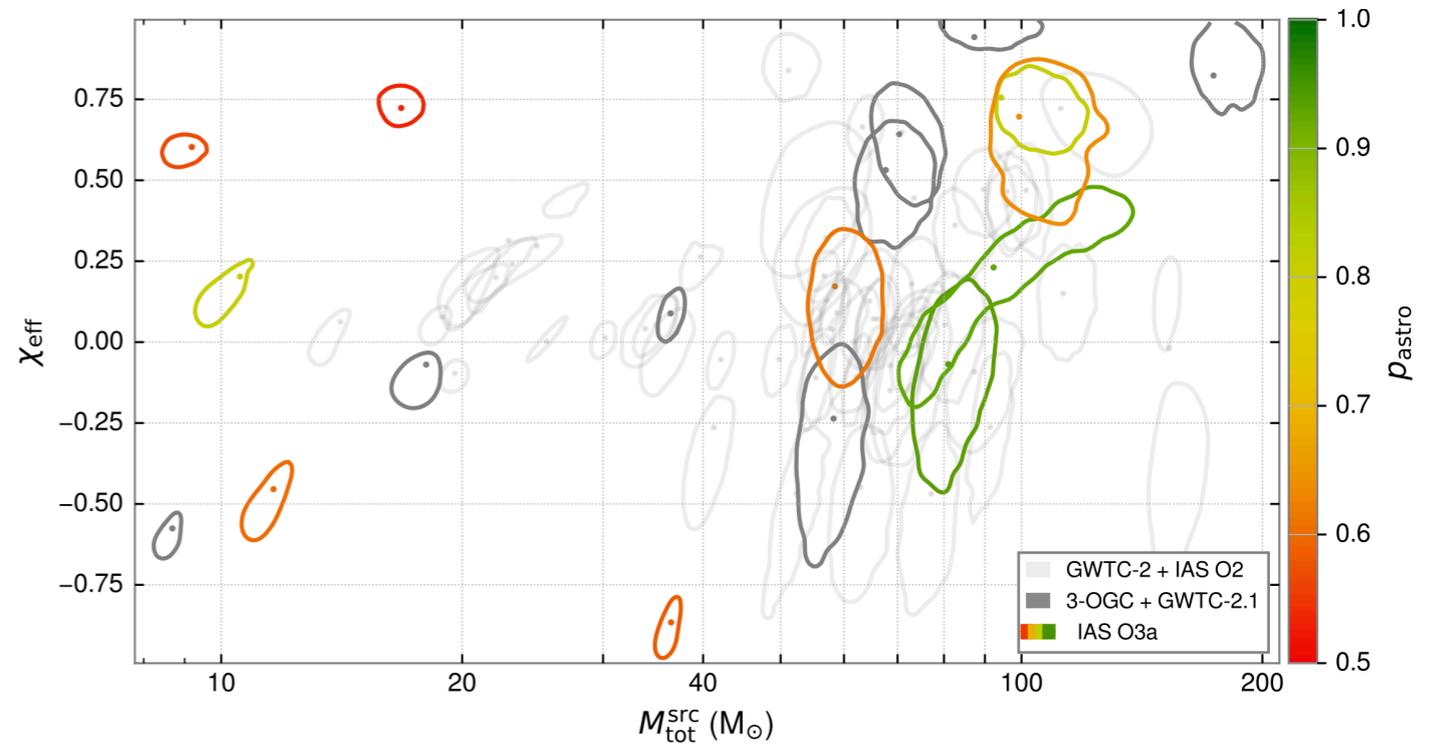
$$\frac{p(d|\mathcal{I})}{p(d|\text{noise})} = \frac{4\pi\Pi_0 D_0^3}{T} \prod_{k \in \text{detectors}} \int d\tau_k \exp\left(\frac{1}{2} |Z_k(\tau_k)|^2\right) \int \frac{d\alpha}{2\pi} \frac{d\delta \cos \delta}{2} [\delta_{k,0} + (1 - \delta_{k,0})\delta_D(\Delta\tau_k - \Delta\tau_k(\alpha, \delta))] \times \int \frac{d\mu}{2} \frac{d\psi}{2\pi} \exp\left\{-\frac{1}{2} \left(\|\mathbf{Z}\|^2 - |\hat{\mathbf{z}}_0(\mu, \alpha, \delta, \psi) \cdot \mathbf{Z}|^2\right)\right\} \|\mathbf{z}_0\|^3 g(|\hat{\mathbf{z}}_0 \cdot \mathbf{Z}|). \quad (\text{D82})$$

Candidate lists for O3a

Event Name	Bank	ρ_H^2	ρ_L^2	p_{astro}	IFAR (yr)		
					IAS ^a	GWTC-2.1	3-OGC
GW190403.051519	(4, 0)	23.1	29.7	—	Veto	0.13	—
GW190408.181802	(3, 0)	95.4	109.2	1.00	> 1000	> 1000	> 1000
GW190412.053044	(2, 0)	76.2	245.5	1.00	> 1000	> 1000	> 1000
GW190413.052954	(4, 0)	26.7	50.5	0.83	4.2	1.2	1.4
GW190413.134308	(4, 2)	30.1	62.3	1.00	> 1000	2.9	6.4
GW190421.213856	(4, 1)	68.0	42.0	1.00	> 1000	71.4	> 1000
GW190426.190642	(5, 5)	24.1	42.7	0.33	0.19	0.24	—
GW190503.185404	(3, 1)	83.2	57.7	1.00	> 1000	> 1000	> 1000
GW190512.180714	(2, 0)	39.4	119.4	1.00	> 1000	> 1000	> 1000
GW190513.205428	(3, 0)	78.0	66.0	1.00	> 1000	> 1000	> 1000
GW190514.065416	(4, 1)	38.9	31.7	0.98	290	0.36	0.19
GW190517.055101	(3, 1)	48.7	58.5	1.00	> 1000	9.1	66.1
GW190519.153544	(4, 2)	81.6	128.7	1.00	> 1000	> 1000	> 1000
GW190521.030229	(5, 5)	65.0	129.8	—	Veto	769	805
GW190521.074359	(3, 1)	142.3	431.3	1.00	> 1000	> 1000	> 1000
GW190527.092055	(3, 1)	27.4	46.9	0.92	10.8	4.3	0.37
GW190602.175927	(4, 3)	41.9	111.6	1.00	> 1000	> 1000	391
GW190701.203306	(2, 2)	25.1	53.8	0.23	0.084	1.8	0.13
GW190706.222641	(4, 3)	91.3	79.2	1.00	> 1000	2.9	> 1000
GW190707.093326	(1, 0)	63.7	97.5	1.00	> 1000	> 1000	> 1000
GW190719.215514	(3, 1)	37.0	33.2	0.90	8.5	1.6	0.25
GW190720.000836	(1, 0)	44.7	62.3	1.00	> 1000	10.6	559
GW190725.174728	(1, 0)	31.3	59.1	0.96	34.2	2.2	0.41
GW190727.060333	(4, 0)	76.0	61.3	1.00	> 1000	> 1000	> 1000
GW190728.064510	(1, 0)	58.4	110.1	1.00	> 1000	> 1000	> 1000
GW190731.140936	(3, 1)	28.9	39.6	0.76	2.1	0.53	0.43
GW190803.022701	(3, 1)	30.6	43.7	0.94	15.7	2.6	2.4
GW190805.211137	(4, 1)	18.8	54.8	0.81	3.3	1.6	—
GW190828.063405	(3, 1)	112.6	142.3	1.00	> 1000	> 1000	> 1000
GW190828.065509	(2, 1)	54.5	53.6	1.00	> 1000	> 1000	> 1000
GW190909.114149 ^b	(3, 1)	31.3	32.4	0.52	0.45	0.010	—
GW190915.235702	(3, 0)	92.4	71.1	1.00	> 1000	> 1000	> 1000
GW190916.200658	(4, 2)	27.1	36.5	0.93	20.7	< 0.001	0.22
GW190917.114629	(0, 0)	26.8	40.6	0.35	0.17	1.5	—
GW190924.021846	(1, 0)	31.9	94.9	—	Veto	> 1000	> 1000
GW190926.050336	(3, 1)	45.3	31.4	0.96	25.3	0.91	0.27
GW190929.012149	(5, 2)	40.2	51.2	1.00	> 1000	6.2	3.1
GW190930.133541	(1, 0)	41.1	55.6	1.00	> 1000	55.6	295

Name	Bank	$m_1(M_\odot)$	$m_2(M_\odot)$	χ_{eff}	z	$\ln \mathcal{L}_{\text{max}}$	ρ_H^2	ρ_L^2	IFAR (yr) ^a	p_{astro}
GW190707.083226	(4, 2)	52^{+17}_{-12}	32^{+12}_{-11}	$-0.2^{+0.5}_{-0.6}$	$0.6^{+0.4}_{-0.3}$	43.9	37.0	31.5	23.2	0.94
GW190711.030756	(3, 1)	80^{+50}_{-40}	18^{+11}_{-7}	$0.2^{+0.3}_{-0.7}$	$0.41^{+0.24}_{-0.16}$	49.5	19.8	60.7	11.2	0.93
GW190818.232544	(4, 3)	67^{+23}_{-19}	38^{+17}_{-15}	$0.7^{+0.2}_{-0.3}$	$1.0^{+0.6}_{-0.4}$	40.5*	33.0	32.0	3.4	0.81
GW190704.104834	(0, 0)	7^{+6}_{-2}	$3.2^{+1.2}_{-1.1}$	$0.20^{+0.27}_{-0.14}$	$0.10^{+0.03}_{-0.03}$	48.7*	47.0	32.1	2.8	0.81
GW190906.054335	(3, 1)	37^{+12}_{-8}	24^{+8}_{-8}	$0.1^{+0.4}_{-0.5}$	$0.9^{+0.4}_{-0.3}$	34.1*	23.6	38.1	0.73	0.61
GW190821.124821	(1, 0)	$7.6^{+3.9}_{-1.7}$	$4.0^{+1.0}_{-1.1}$	$-0.45^{+0.33}_{-0.17}$	$0.17^{+0.06}_{-0.06}$	48.5*	28.1	49.4	0.71	0.60
GW190814.192009	(5, 4)	68^{+28}_{-19}	48^{+21}_{-18}	$0.5^{+0.4}_{-0.6}$	$1.5^{+0.8}_{-0.7}$	25.2	29.9	33.4	0.65	0.64
GW190910.012619	(1, 1)	34^{+3}_{-3}	$2.9^{+0.3}_{-0.2}$	$-0.87^{+0.19}_{-0.11}$	$0.16^{+0.04}_{-0.04}$	40.2*	35.7	32.1	0.65	0.58
GW190920.113516	(0, 0)	$6.0^{+3.3}_{-1.5}$	$3.2^{+0.9}_{-1.0}$	$0.60^{+0.26}_{-0.07}$	$0.13^{+0.05}_{-0.05}$	40.7	26.4	48.0	0.56	0.57
GW190718.160159	(1, 0)	$10.0^{+4.5}_{-1.8}$	$6.8^{+1.4}_{-2.1}$	$0.73^{+0.10}_{-0.17}$	$0.28^{+0.10}_{-0.09}$	41.1*	23.5	47.6	0.48	0.53

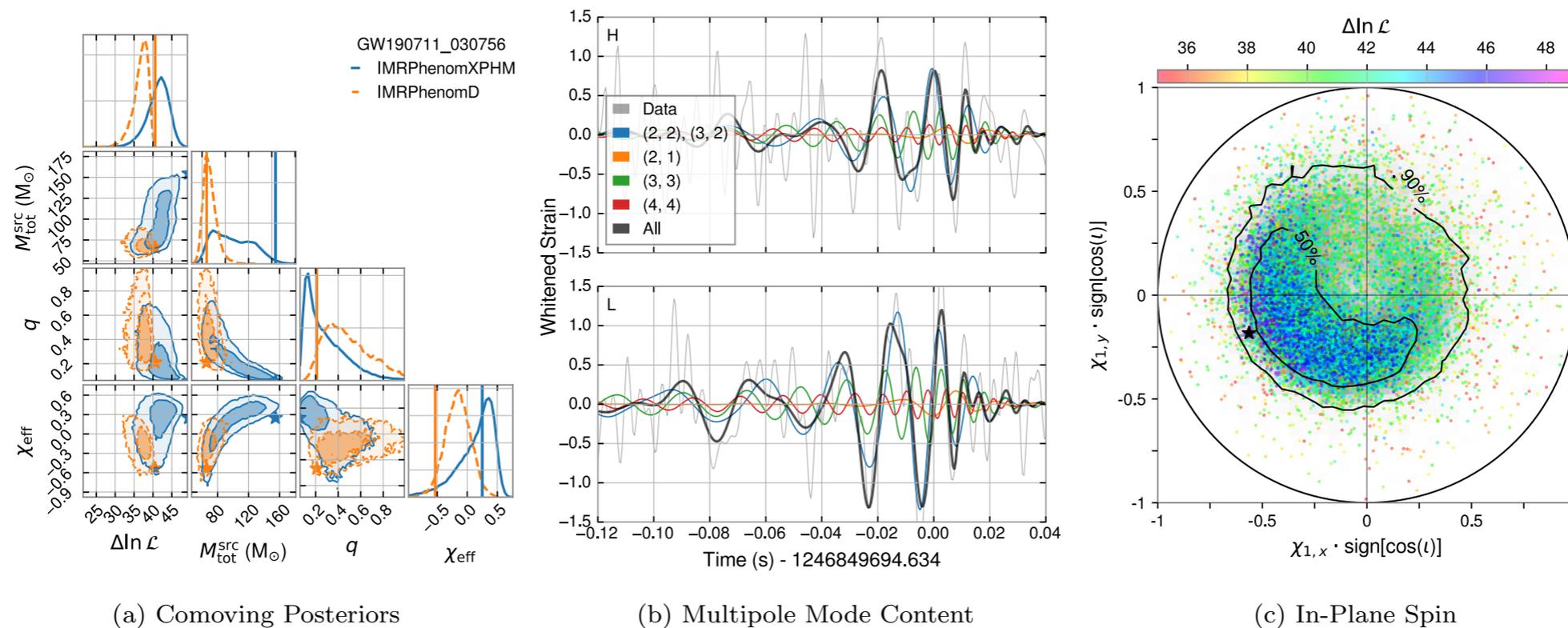
New candidates



Results on known candidates

Interesting new events

- Signals with higher harmonics and orbital precession

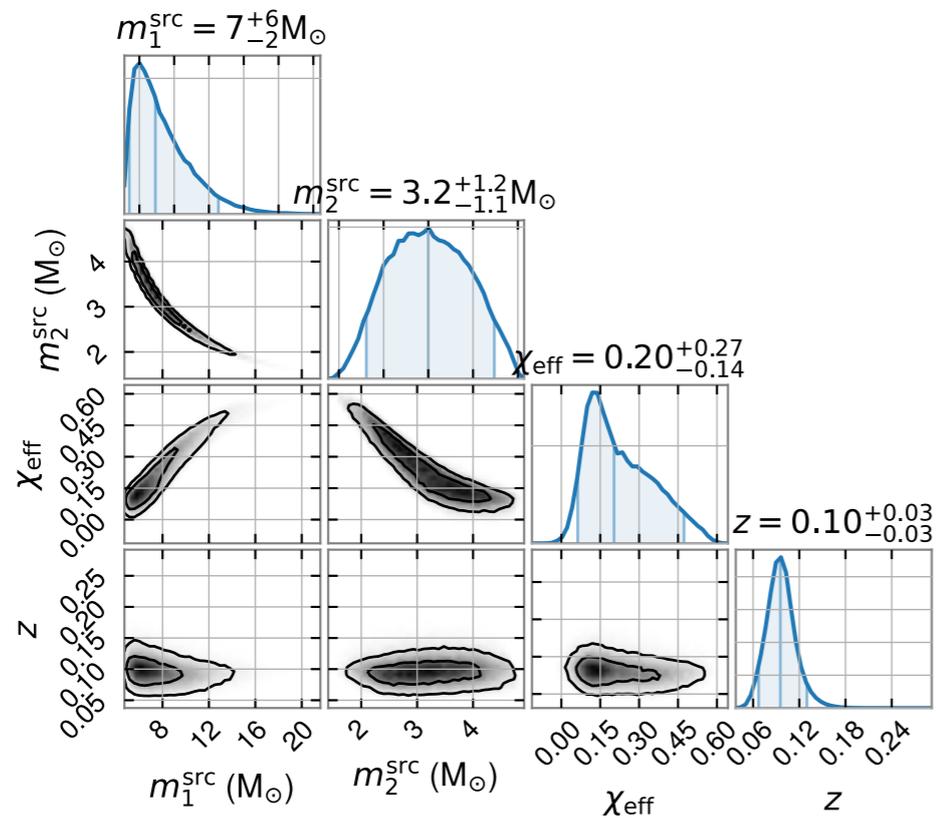


GW190711_030756 ($p_{\text{astro}} = 0.93$)

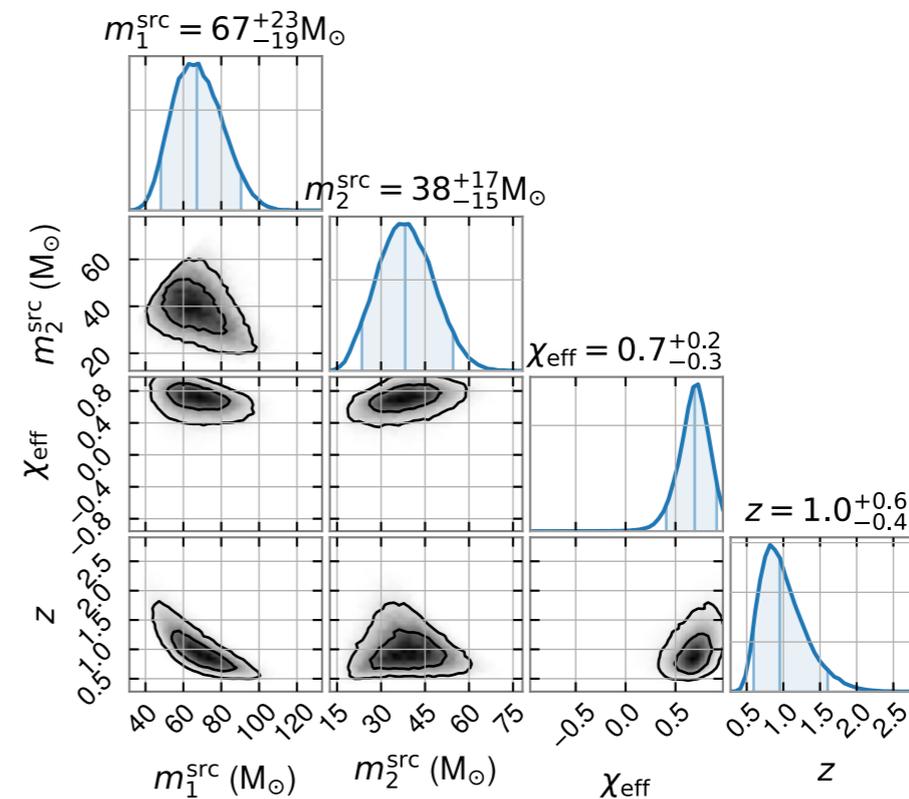
Interesting new events

- Events with BH masses in the lower and upper mass gaps

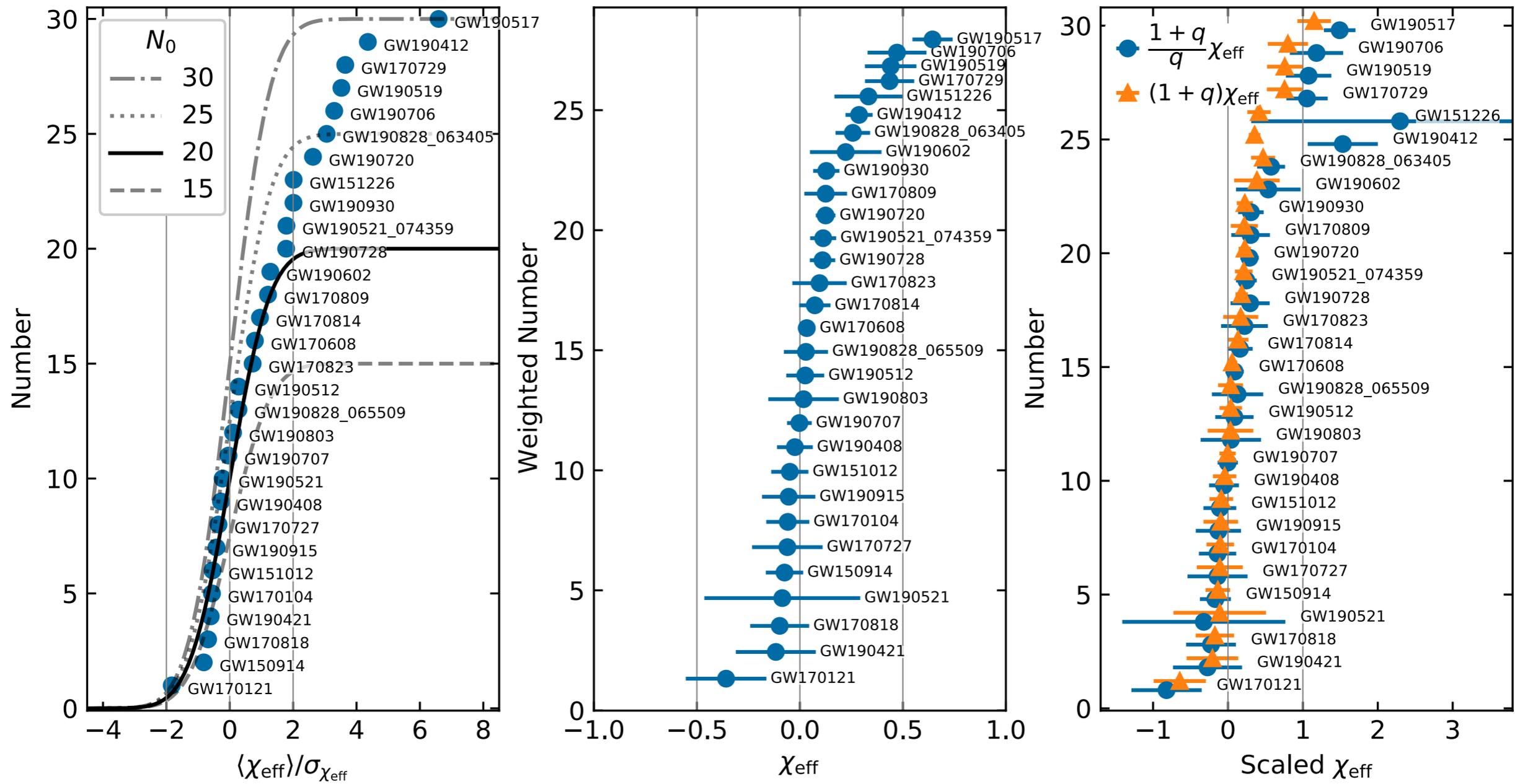
GW190704_104834: IFAR = 2.8 (y)



GW190818_232544: IFAR = 3.4 (y)



Population analysis



Puzzle in the distribution of the effective spin

$$\chi_{\text{eff}} := \frac{\chi_1 + q \chi_2}{1 + q} \cdot \hat{L},$$

Roulet et. al., (2021)

Population analysis

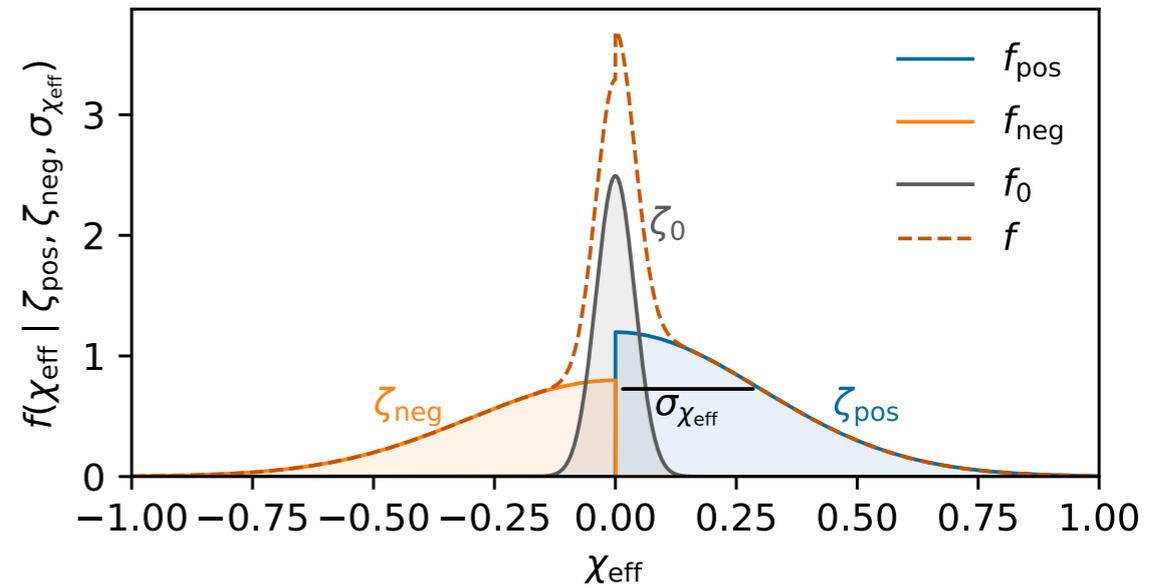
$$P(\{d_i\} | \lambda) \propto e^{-N_a(\lambda)} \prod_{i=1}^{N_{\text{trig}}} \left(\frac{dN_a(\lambda)}{dN_a(\lambda_0)} \Big|_{d_i} p_{\text{astro},i}(\lambda_0) + 1 - p_{\text{astro},i}(\lambda_0) \right).$$

(Unknown) True model for BBH population

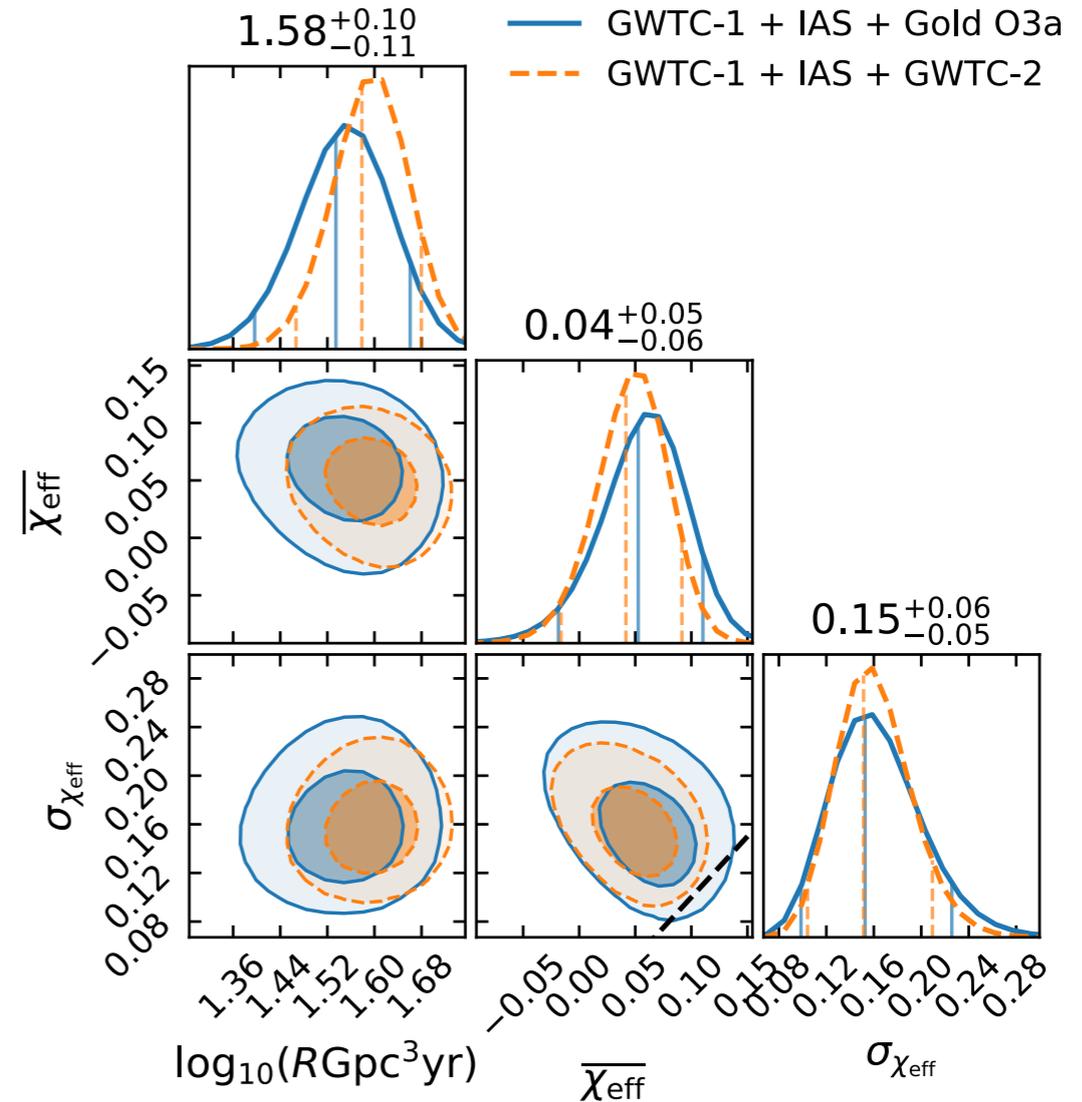
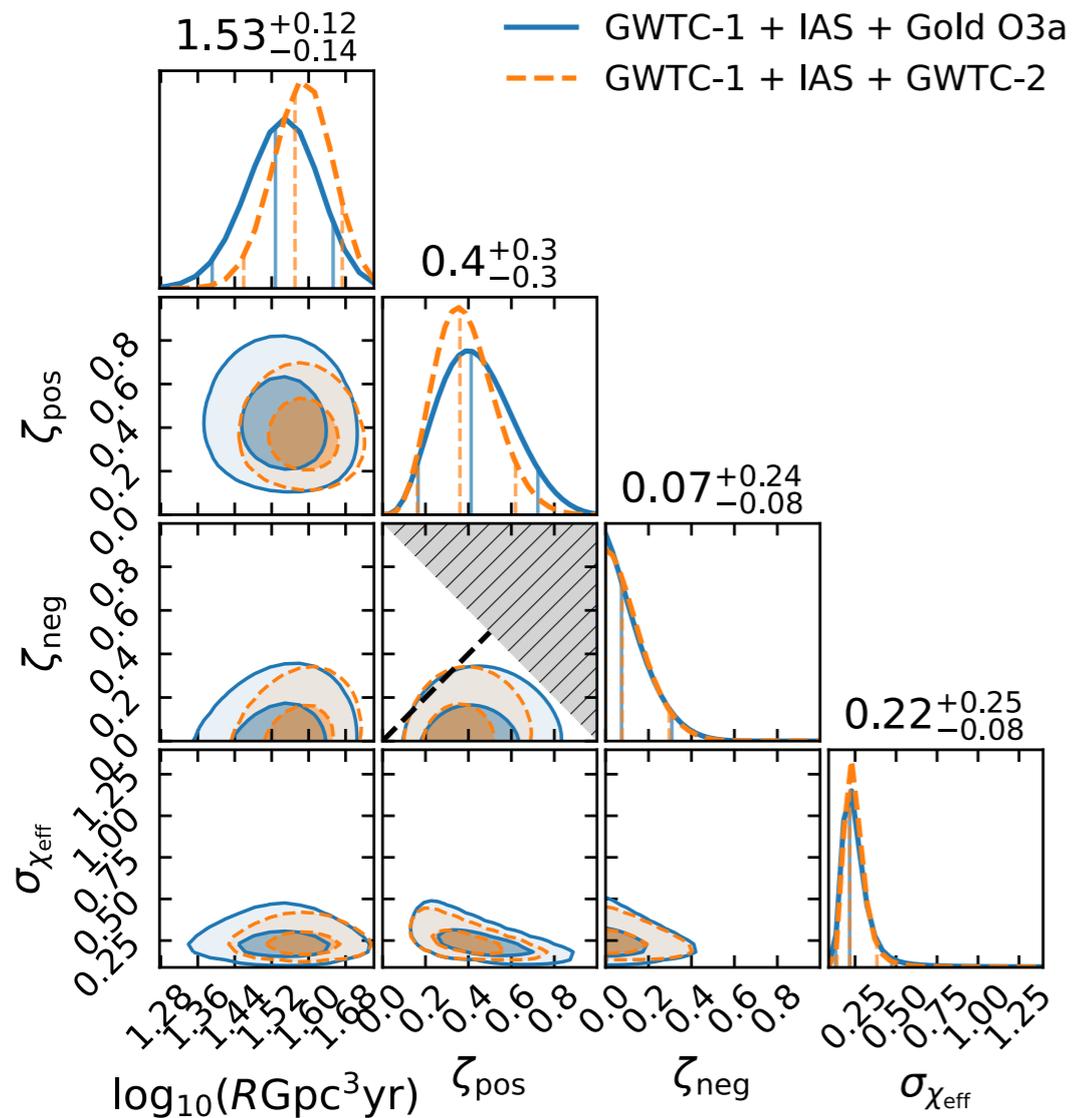
Fiducial model for BBH population

$$f(\chi_{\text{eff}}, m_{1s}, q, D_L) = f_{\chi_{\text{eff}}}(\chi_{\text{eff}}) f_{m_{1s}}(m_{1s}) f_q(q) f_{D_L}(D_L).$$

$$f_{\chi_{\text{eff}}}(\chi_{\text{eff}} | \zeta_{\text{pos}}, \zeta_{\text{neg}}, \sigma_{\chi_{\text{eff}}}) = \zeta_0 \mathcal{N}(\chi_{\text{eff}}; \sigma_0 = 0.04) + \zeta_{\text{neg}} \mathcal{N}_{<0}(\chi_{\text{eff}}; \sigma_{\chi_{\text{eff}}}) + \zeta_{\text{pos}} \mathcal{N}_{>0}(\chi_{\text{eff}}; \sigma_{\chi_{\text{eff}}}).$$



Population analysis



$$f_{\chi_{\text{eff}}}(\chi_{\text{eff}} | \zeta_{\text{pos}}, \zeta_{\text{neg}}, \sigma_{\chi_{\text{eff}}}) = \zeta_0 \mathcal{N}(\chi_{\text{eff}}; \sigma_0 = 0.04) + \zeta_{\text{neg}} \mathcal{N}_{<0}(\chi_{\text{eff}}; \sigma_{\chi_{\text{eff}}}) + \zeta_{\text{pos}} \mathcal{N}_{>0}(\chi_{\text{eff}}; \sigma_{\chi_{\text{eff}}}) .$$

$$f(\chi_{\text{eff}}) \sim \mathcal{N}(\bar{\chi}_{\text{eff}}, \sigma_{\chi_{\text{eff}}})$$

Roulet et. al., (2021)

Also: Galaudage, S., et. al. (2021)

Symmetric χ_{eff}

Positive χ_{eff}

Positive/Negative mixture χ_{eff}

Gaussian χ_{eff}

$\Delta \max \ln L$ $\Delta \ln Z$

0 0

$2.1^{+0.5}_{-0.4}$ $1.6^{+0.5}_{-0.3}$

$2.1^{+0.5}_{-0.4}$ $1.4^{+0.4}_{-0.2}$

$0.2^{+0.7}_{-0.6}$ $-0.2^{+0.6}_{-0.8}$

Interesting future directions

- The improvements to the LIGO detector have largely left the low-frequency noise ($< \sim 60-70\text{Hz}$) unaffected between the O2-O3 runs. What could be the reason?
- If this is mitigated and we approach “design” sensitivity, we should generically see more cycles in BBH waveforms. Are current waveform models good enough to analyze these signals?
- Can we include more physics (orbital precession + higher harmonics) into the signal models for searches?

Summary

- The availability of the LVK data gives everyone in the community the opportunity to try out and propose new ideas and methods.
- We developed a new end-to-end analysis pipeline that incorporates several innovations
- We achieved an improvement in sensitive volume as quantified by the false alarm rates of known events and injections
- We have new candidate events in all the observing runs, above the thresholds for detection as defined by the LVC
- We also find interesting new results when we analyze the astrophysical population - the distribution of effective spins suggests a preference for positively aligned vs negatively aligned spins in the population