

From 0 to 50 in 12 years – Development of binary black hole searches in LIGO/Virgo

Commentary on B. Zackay,
'Detecting GW on the open LIGO-Virgo data'

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LIGO Document G2101205-v2



Topics to mention (if time)

- 1) Recent history : How the (statistical) context has changed since ~2009
- 2) Present / near future : Continuing and harder statistical challenges
- 3) A remark on data geometry and GW event ranking

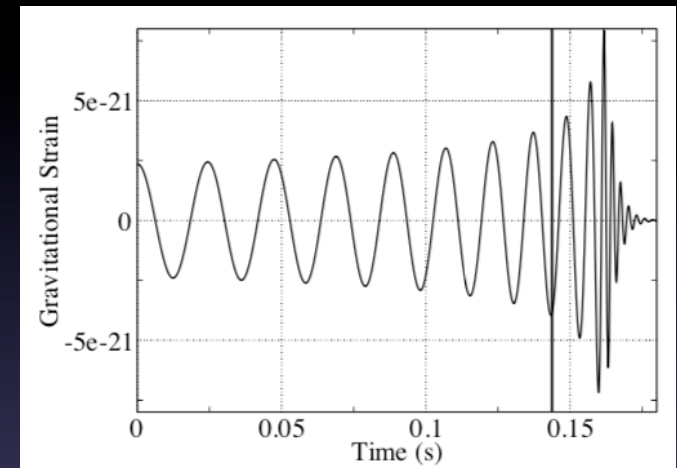
The statistical framework

- 'The matched filter is the optimal detection statistic'

... how many ways can that statement be wrong or misleading?

Problems with the signal (historical)

- Initial LIGO (2005-2007) : *no general BBH signal model available*
- EOB / NR
- ***EOBNRv1***
A. Buonanno et al., Phys. Rev. D 76, 104049 (2007)
- IMRPhenom ...
- SEOBNRvN ...



LSC/Virgo, Phys. Rev. D 83, 122005 (2011)

Now have quantified model accuracy & effects of (omitting) spin precession, higher multipole emission, ...

How to search with 'the wrong signals' ?

1. Signal model uncertain in some parameter regions
2. Signal templates do not include all physical parameters / effects (spins, HM, ...)

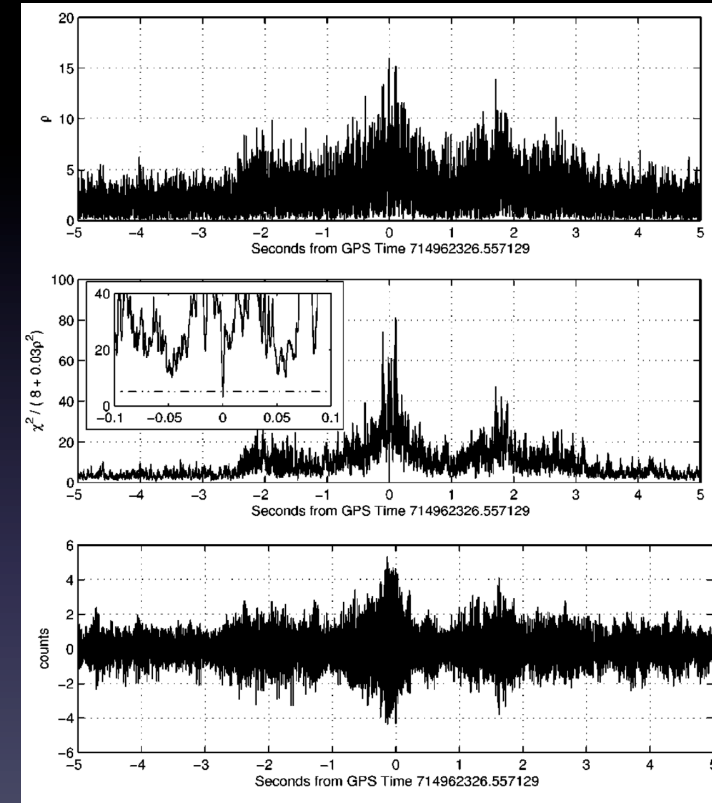
Not aware of formal work on matched filter search with uncertain / incomplete template set

- e.g. expanded search space including parameterised uncertainties ?
- necessarily pick up extra noise \Rightarrow suboptimal

Problems with the data

LIGO-Virgo data are *in general* non-stationary & non-Gaussian

- Infinite number of ways this could manifest itself
- To make progress need *approximately correct* description of detector noise



LSC, PRD 69, 122001 (2004)

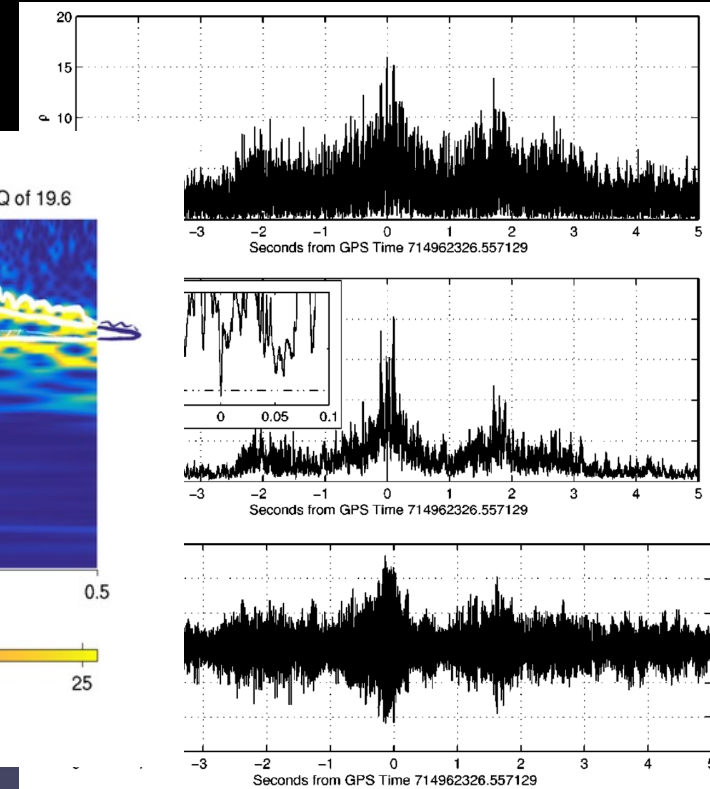
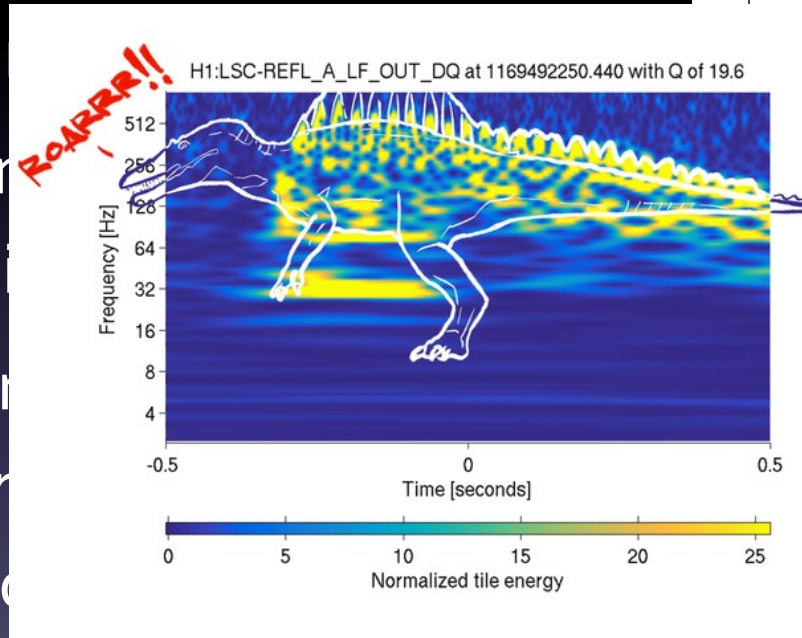
- NB - detector behaviour is different in every run !

Problems with the data

LIGO-Virgo data are *in general*

non-stationary

- Infinite number of non-stationary events could manifest themselves
- To make progress, we need a *primarily* consistent detector noise



- NB - detector behaviour is different in every run !

'Usual' assumptions about noise

- Consider as 'quasi-stationary' (approx. stationary over short time periods)
+ isolated 'bad' times inconsistent with Gaussian distribution (high amplitude 'glitches')
- Measure longer time drifts in noise spectrum
(Zackay et al. demonstrate *shorter-time* variations & effect on optimal statistic)
- Use generic excess power & signal consistency checks to identify and remove / penalize 'glitch' times
(IAS search uses 'inpainting' excision & several novel checks)

Initial LIGO search philosophy

How to impose signal consistency if we don't know the signal accurately?

- Heuristic : 'Eyes wide open'
 - Deliberately loose cuts/penalties for events 'somewhat mismatching' templates
 - should also allow for extra physics (spin, HM ...)?
 - ss a consequence *significantly higher* noise background
- IAS search using opposite heuristic – 'reject anything with significant mismatch to templates'
 - requires confidence in model accuracy
 - what fraction of loud signals are really rejected?

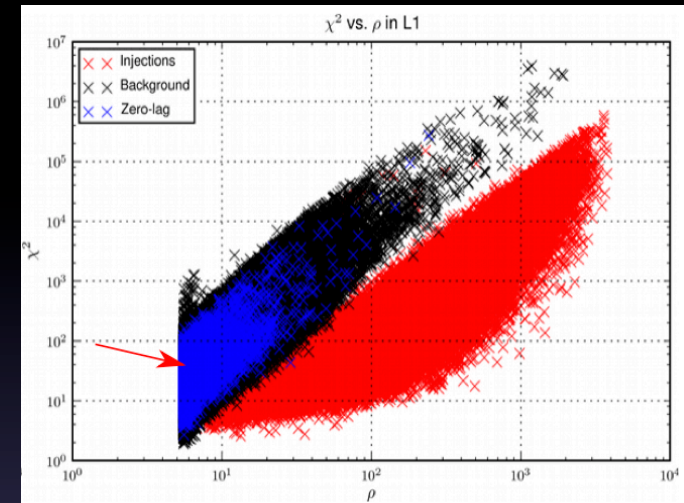


image: LSC-V / K. Cannon

How do the assumptions fare in O₃?

New player : Long-duration, band-limited excess noise

- Caused by scattered light re-entering ifo arms

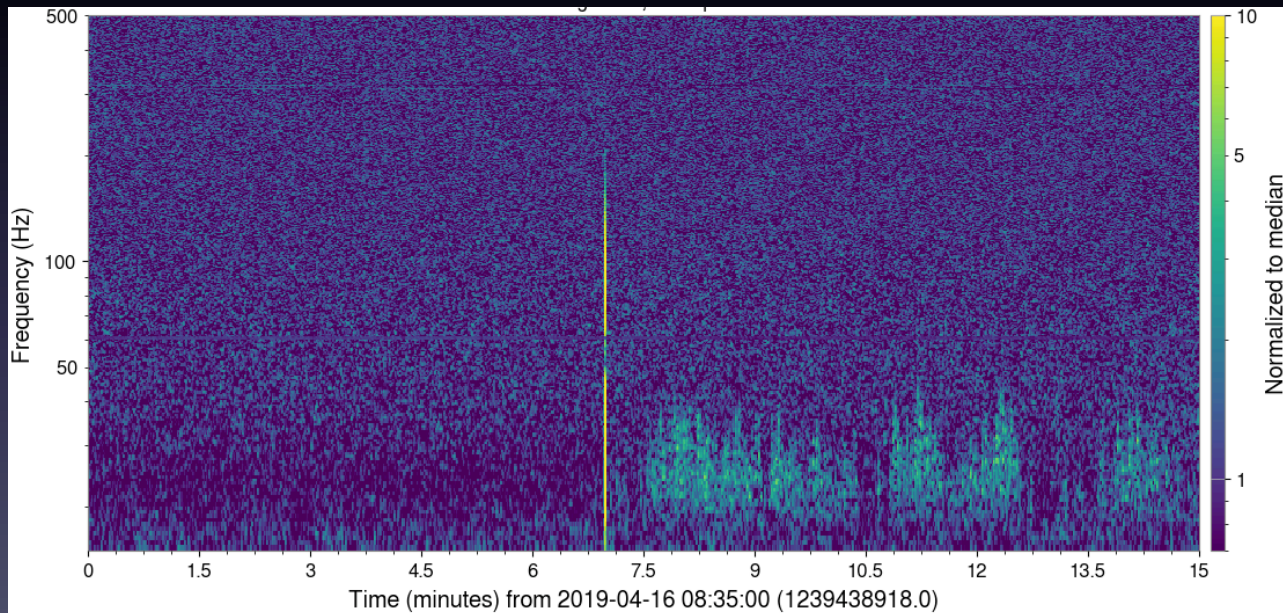


image: LDVW (LIGO open data)

- Does not fit within previous paradigm ...

Astrophysics in a time of scattering

- Many O3a events detected *despite* scattering noise

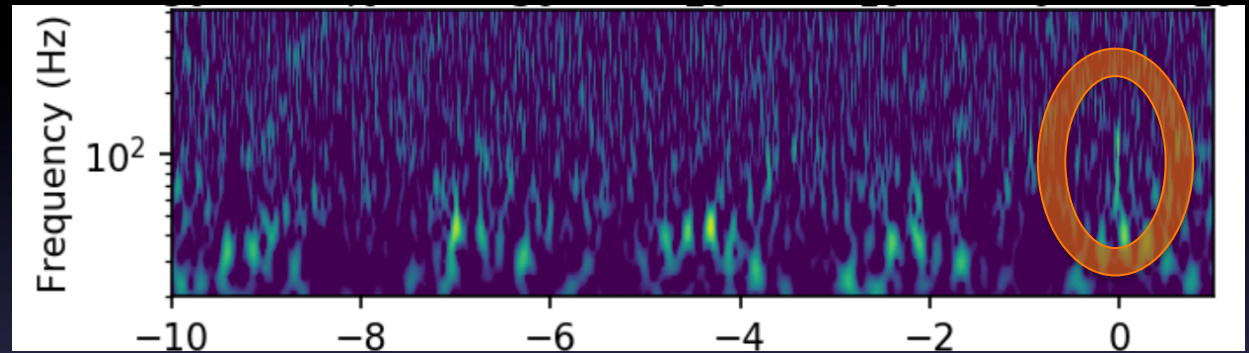


image: PyCBC (LIGO open data)

- require 'mitigation' to extract event properties
 - detailed modelling & subtraction of extended excess noise

Name	Mitigation
GW190413.134308	L1 glitch subtraction, glitch-only model
GW190424.180648	L1 glitch subtraction, glitch-only model
GW190425	L1 glitch subtraction, glitch-only model
GW190503.185404	L1 glitch subtraction, glitch-only model
GW190513.205428	L1 glitch subtraction, glitch-only model
GW190514.065416	L1 glitch subtraction, glitch-only model
GW190701.203306	L1 glitch subtraction, glitch+signal model
GW190727.060333	L1 f_{\min} : 50 Hz
GW190814	L1 f_{\min} : 30 Hz; H1 non-observing data used
GW190924.021846	L1 glitch subtraction, glitch-only model

Problems with the signal population

- Before (late) 2015 : ~zero idea of the actual BBH population
- Rate 'predictions' cover many orders of magnitude & ~no idea about masses, spins, ...

- No overall search *figure of merit* (... for what source population ?)

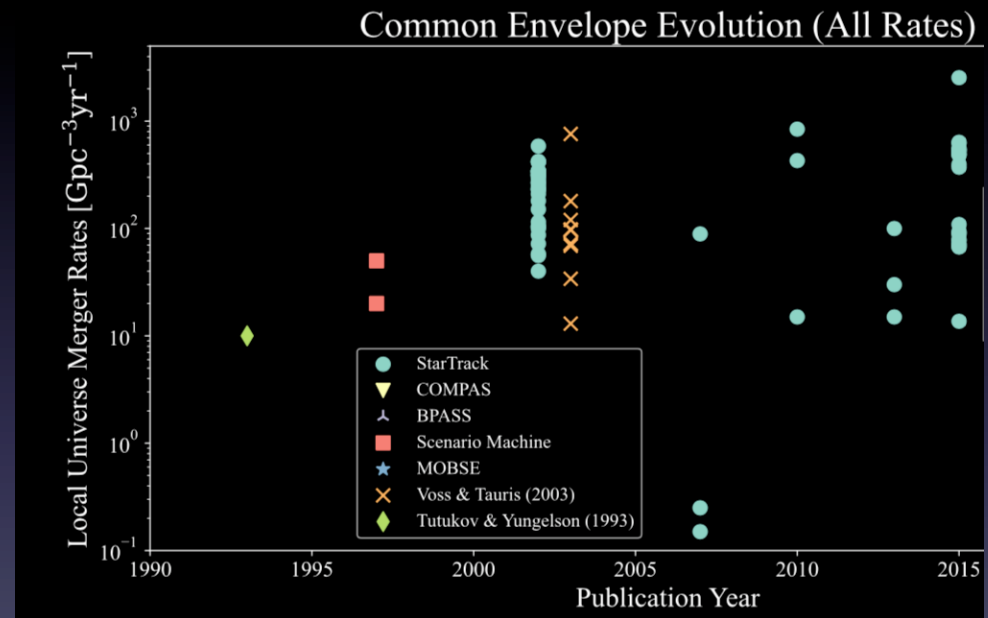


image: C. Rodriguez, CMU

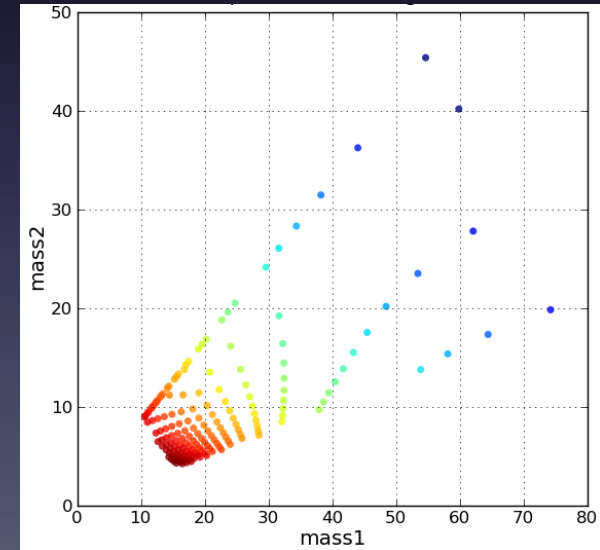
- Initial LIGO optimized *de facto* weighting each template equally (goal: rate upper limit over a wide range of binaries)

A (historical) alternative approach

- 2011 : 'Focused highmass' proposal (unpublished) motivated by Belczynski et al. 2010 ApJL

Our results indicate that BH–BH binaries are 25–times more likely sources than NS–NS systems and that we are on the cusp of GW detection.

- Most likely first detection scenario : near equal mass BBH ($q = m_2/m_1 > 1/4$)
- Much smaller template bank
- Full use of signal consistency
- **Factor of 2.5 – 3 improved sensitive volume over 'standard' LIGO search**

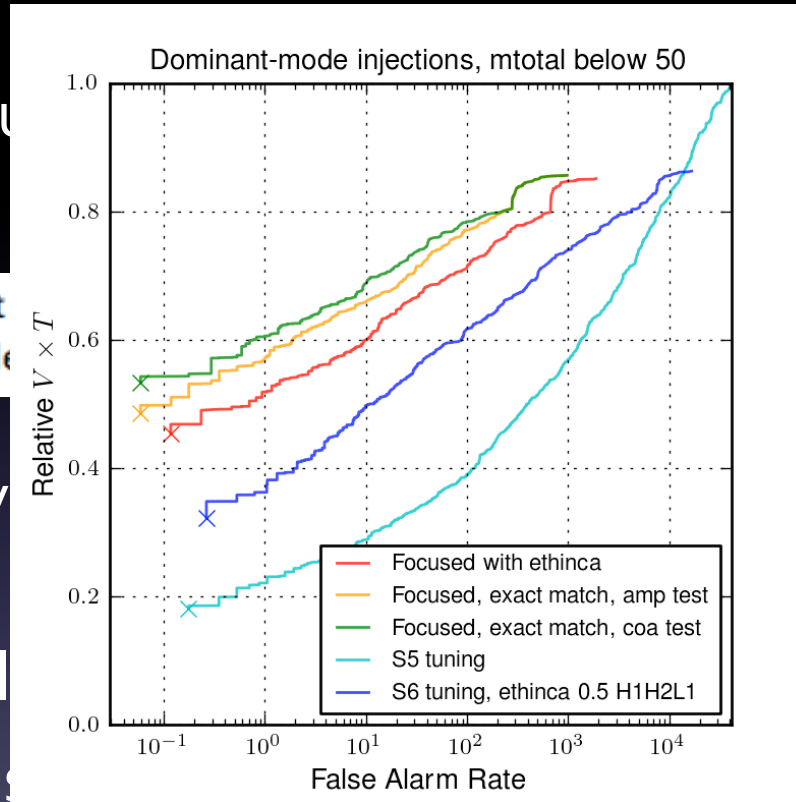


A (historical) alternative approach

- 2011 : 'Focused' search (published)
- motivated by LIGO results

Our results indicate that we are on the cusp of GW detection

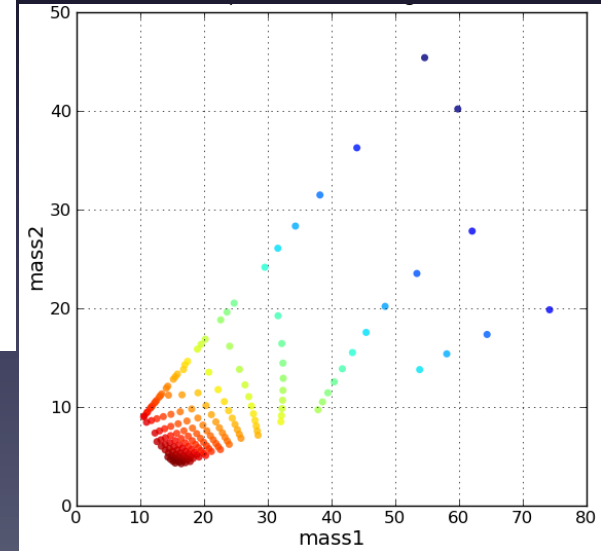
- Most likely to find NS-NS systems and that we are on the cusp of GW detection
- Near equal to standard LIGO search
- Much smaller volume
- Full use of sensitive volume
- Factor of 2.5 – 3 improved sensitive volume over 'standard' LIGO search



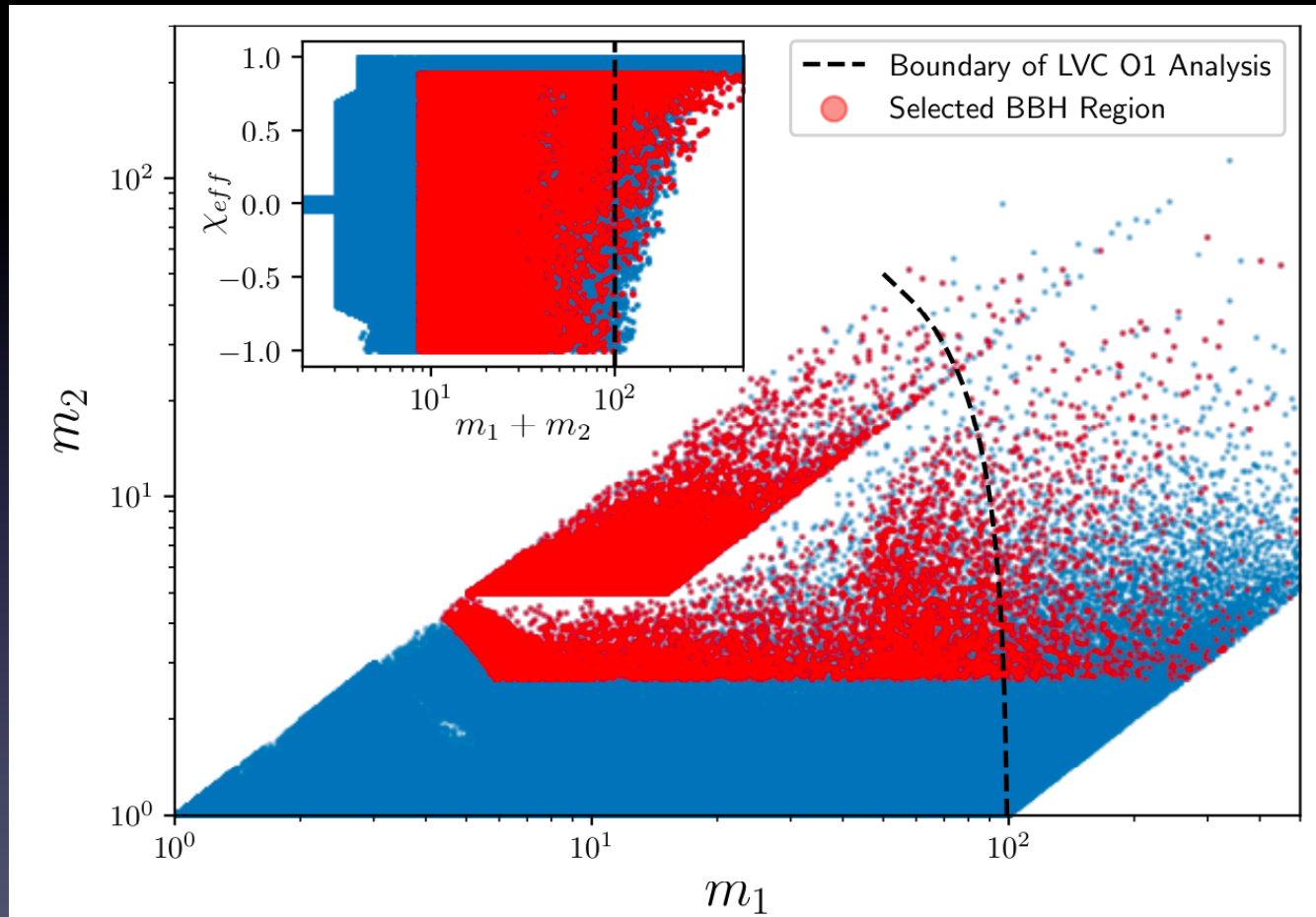
(published)

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than NS-NS systems and that we

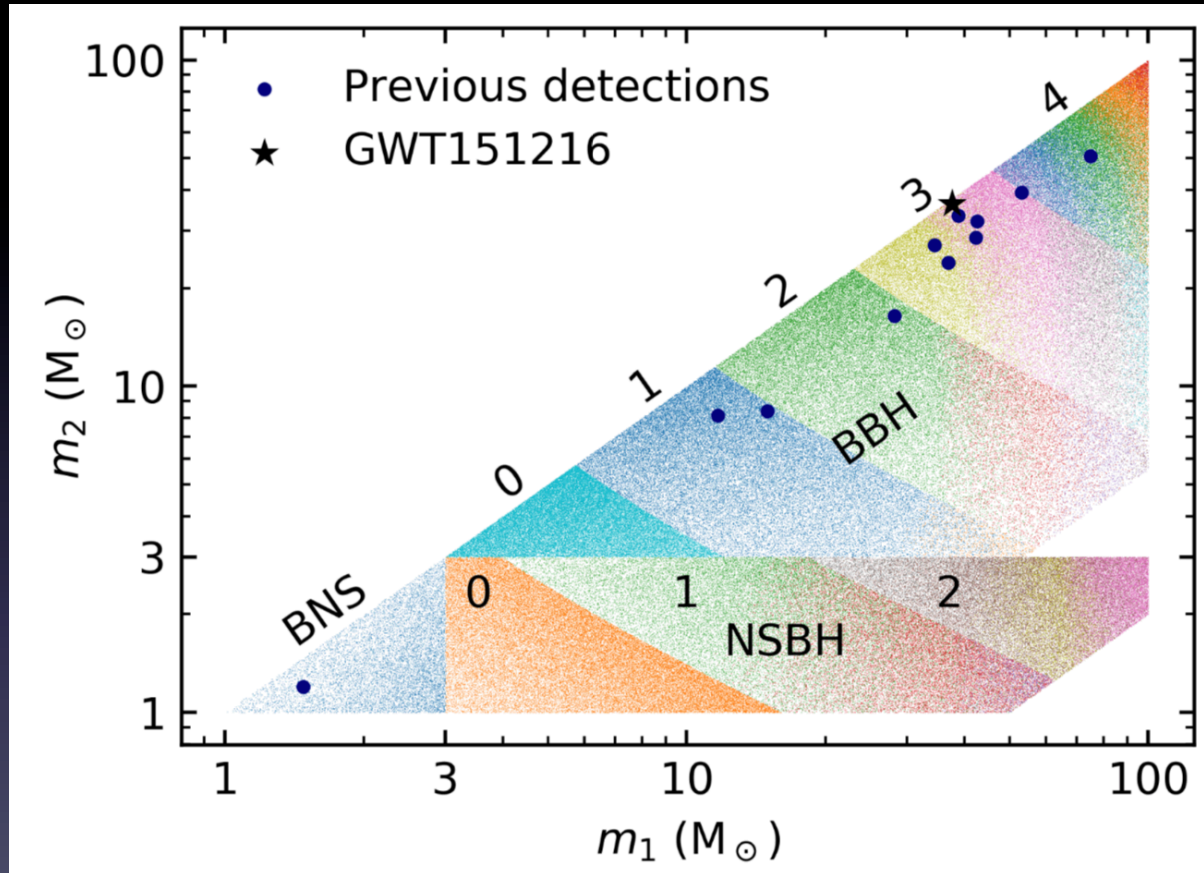


First appearance of 'focused BBH'



Reanalysis of O1 open data : '1-OGC', Nitz et al. ApJ **872** 195 (2019)

GW events vs. IAS template bank



- Odds for signal vs. noise \propto
density of signal events / density of noise events

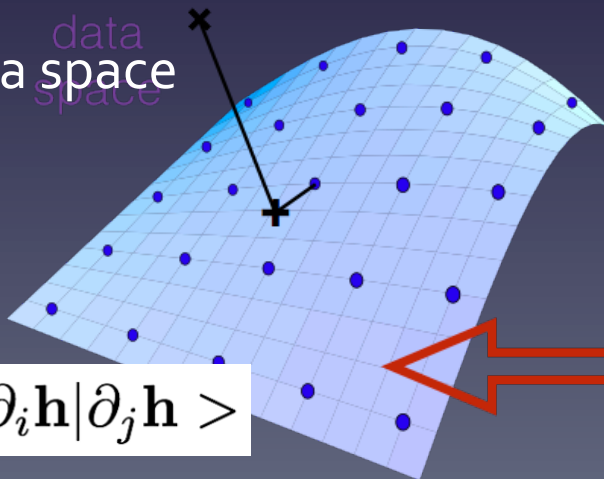
(from B. Zackay's talk)

Optimal search for a known population

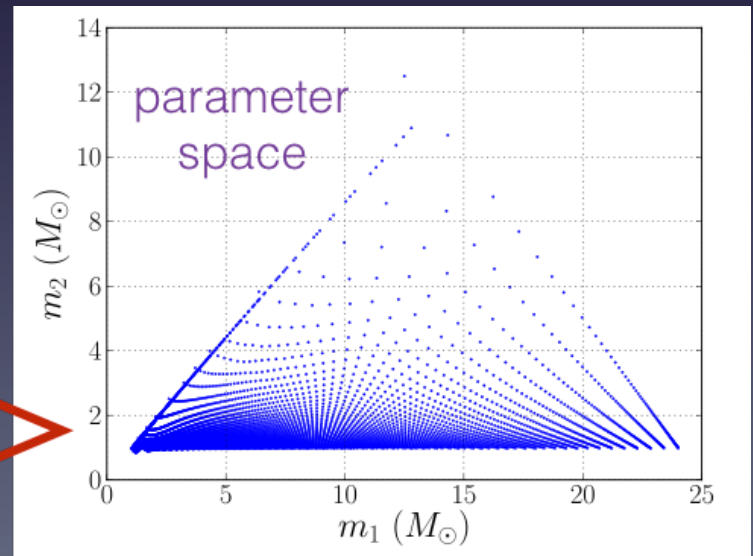
GW detection is a *composite hypothesis* problem

- Formally: marginalize \mathcal{L} over signal parameters (time, phase, masses, spins ...)
- Approximate \mathcal{L} as quadratic near peak : get Fisher matrix
- Also get *metric over binary parameter space*

data
data space
space



$$g_{ij} = \langle \partial_i \mathbf{h} | \partial_j \mathbf{h} \rangle$$



Fisher metric \Rightarrow noise event density

Marginalize over masses & spins \Rightarrow

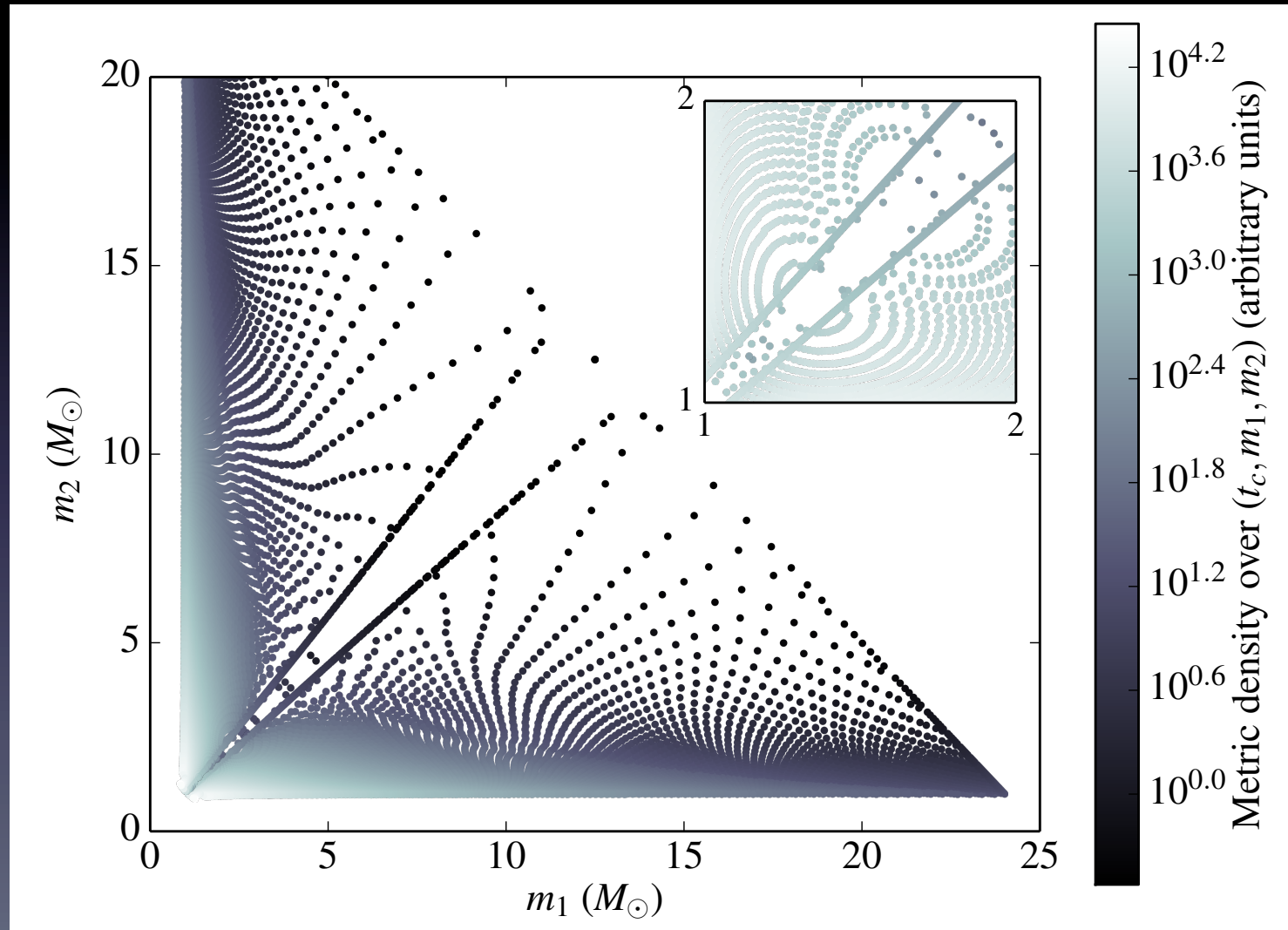
get factors

$$\exp\left(-\frac{\rho_{\max}^2}{2}\right)$$

$$\sqrt{\frac{(2\pi)^{n+1}}{\det \gamma(\vec{\theta}^{\text{ML}})}}$$

- Matched filter likelihood effectively corrected by $1/(\text{Fisher metric density})$ – cf. $dV \propto \sqrt{\det g_{ij}}$
- Similar to *template placement metric* (describing density of independent templates)
- Interpretation:
likelihood ratio $\propto 1/(\text{density of noise events})$

'Information' density is not source population density!



Takeaways

- Always necessary to understand data (noise !)
 - New noise description for every observing run ?
 - No 'silver bullet' (... deep learning ?!)
- Matched filter search is easier to optimize if you know your signal model is accurate
 - still not solved for precessing spins, HM, eccentricity ...
- Matched filter search is easier to optimize if you know what your signal *population* is
 - so far not applied for precessing spins, HM, eccentricity ...
- What does 'optimize' mean anyway ?
 - do we always want to maximize total signal count ??