

# Gravitational wave data analysis



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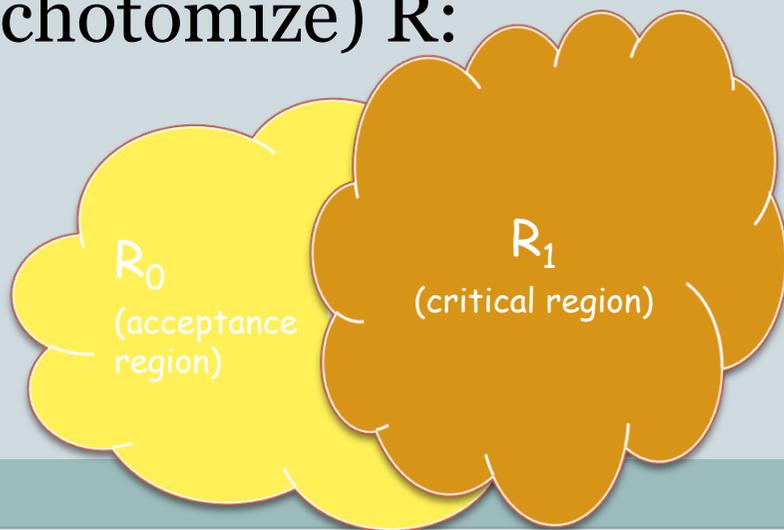
# This presentation



- Aim: make it possible for you to read and understand the observational papers of LIGO and Virgo.
  - Very brief basics of signal detection
  - Searches for compact binary inspiral signals

# Hypothesis testing

- Our data  $\{y_i\} \in \mathbb{R}$
- Consider a detection problem:  $H_0$  signal absent,  $H_1$  signal present
- Question: has hypothesis  $H_0$  or  $H_1$  produced our data ?
- Deciding means finding a way to partition (dichotomize)  $\mathbb{R}$ :

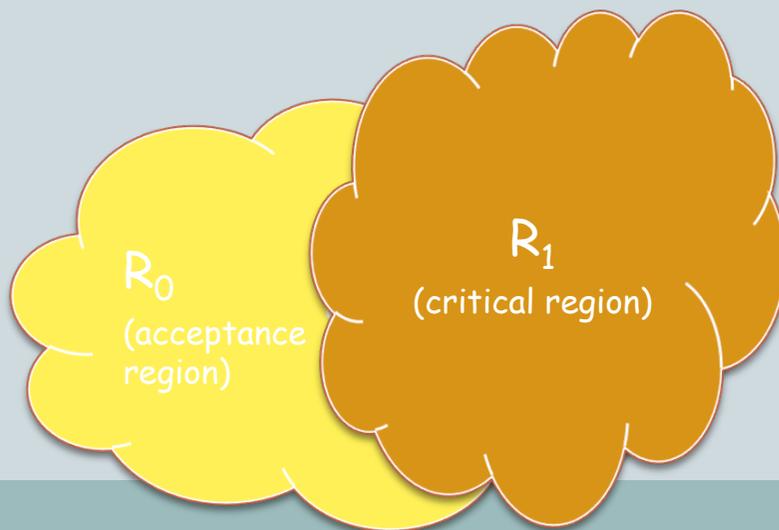


if  $\{y_i\} \in R_0 \rightarrow D_0$   
if  $\{y_i\} \in R_1 \rightarrow D_1$

# Hypothesis testing- types of errors



- Type I error: decide  $D_1$ , when  $H_0$  holds
  - False alarm probability,  $P_{fa}=P(D_1|H_0)$ , size of the test
- Type II errors:decide  $D_0$ , when  $H_1$  holds
  - False dismissal probability,  $P_{fd}=P(D_0|H_1)$ ,  $1-P_{fd}$  is the power of the test



if  $\{y_i\} \in R_0 \rightarrow D_0$   
if  $\{y_i\} \in R_1 \rightarrow D_1$

# Neymann-Pearson criterium



- The decision should be such that at fixed  $P_{fa}$  the  $P_{fd}$  is the smallest.
- It can be demonstrated that the corresponding partition is any level surface of a function of the data called the likelihood:

$$\Lambda(\mathbf{y}) = \frac{p_1(\mathbf{y})}{p_0(\mathbf{y})}$$

prob given  $H_1$   
prob given  $H_0$

# Neymann-Pearson criterium



- The specific level surface that one takes depends on convenience and defines the detection statistic.
- The partition is a threshold on the detection statistic that determines the  $P_{fa}$  and the  $P_{fd}$ .

$$R_1 \supset y \mid \Lambda(y) > \Lambda^*$$

$$P(\Lambda(y) > \Lambda^* \mid H_0) = P_{fa}$$

# A very simple example

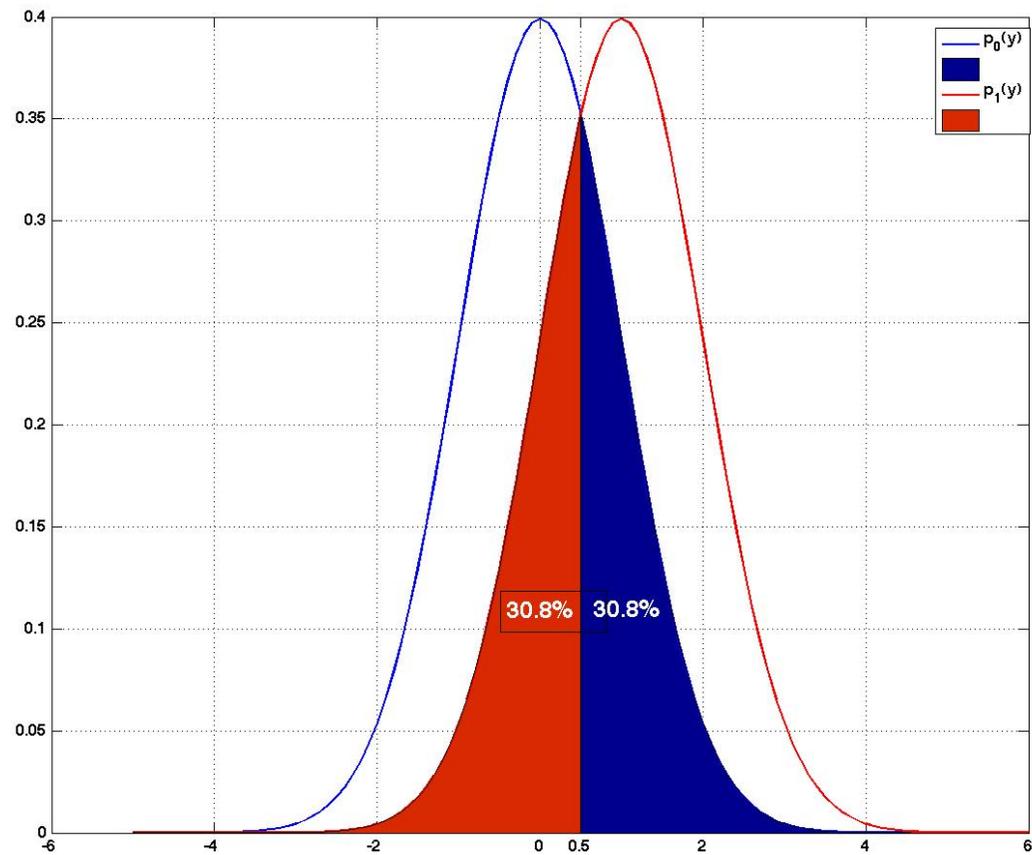


- Consider a single measurement
- $y = n + s$ 
  - $n$  is Gaussian noise, zero mean and unit variance
  - $s = 1$  is a constant, our signal

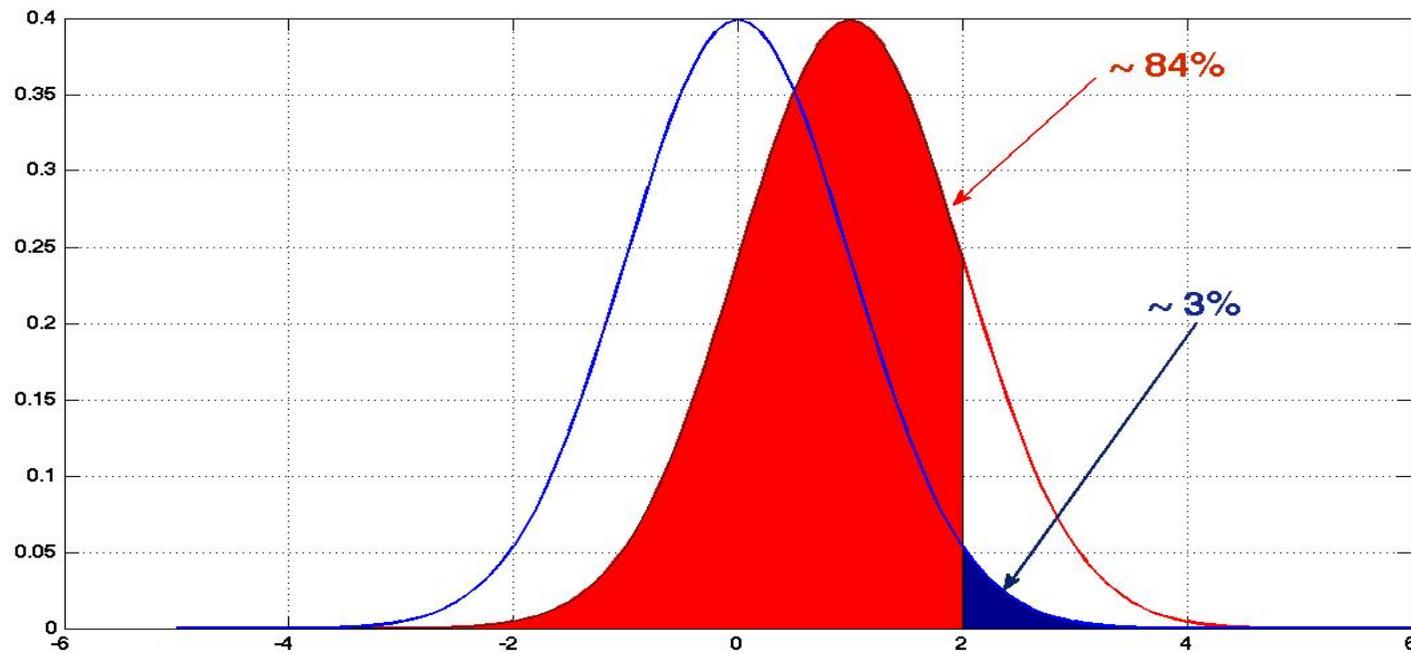
$$p_0 = \frac{1}{\sqrt{2\pi}} e^{-y^2/2} \quad p_1 = \frac{1}{\sqrt{2\pi}} e^{-(y-1)^2/2} \quad \text{and} \quad \Lambda(y) = \frac{p_1(y)}{p_0(y)}$$

- Neymann-Pearson Criterium:
  - $\Lambda(y) = e^{(y-1/2)}$  → Rule is: threshold on  $y$

# If we set $y^* = 0.5$

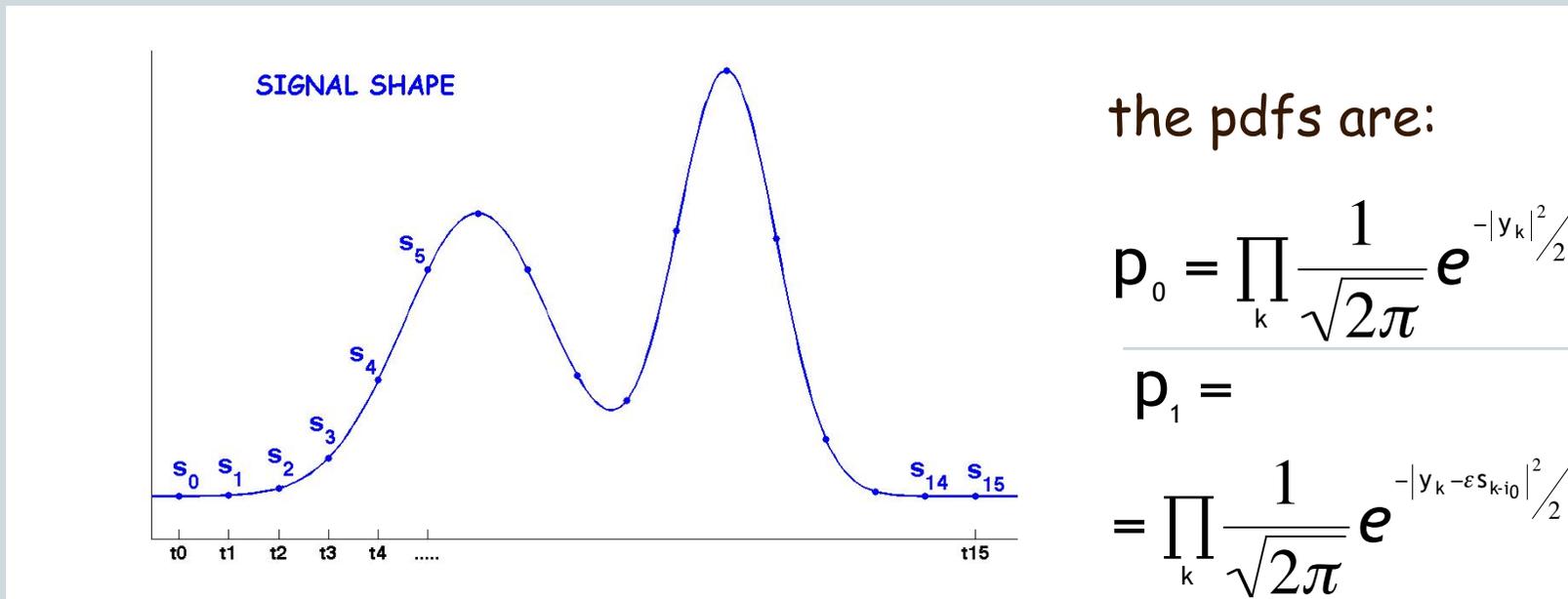


If we set  $y^* = 2$



# Another example, a set of measurements

- Consider some measurements
- $y_i(i_0) = n_i + \epsilon s_{i-i_0}$ 
  - $n$  is Gaussian noise, zero mean and unit variance
  - $s_{i-i_0}$  is a signal of known shape, arriving at time  $i_0$



a convenient level surface of the likelihood is  $\rho(\{y\}, i_0) = \sum_k \epsilon s_{k-i_0} y_k$

# Likelihood and matched filtering

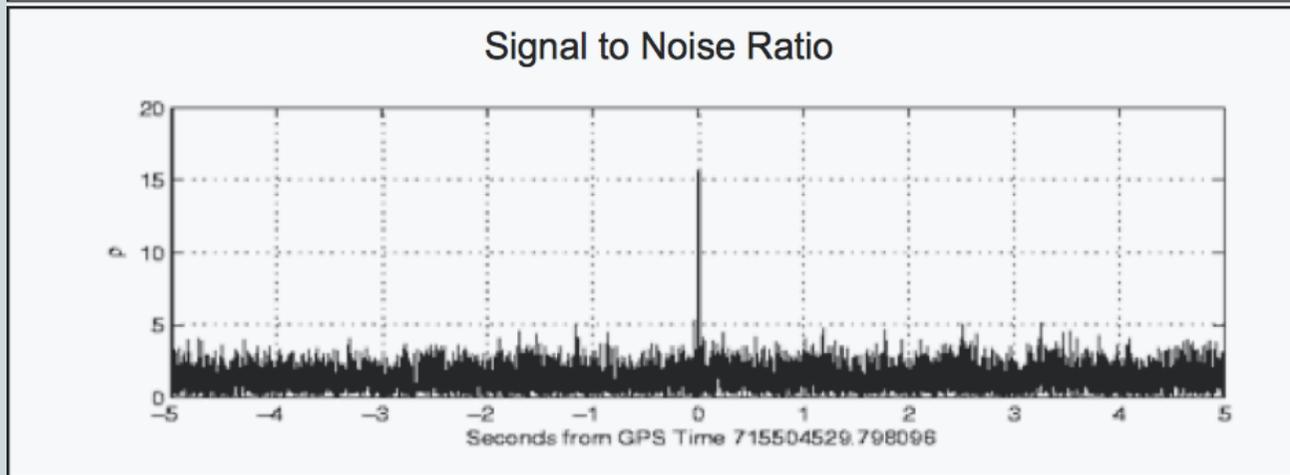
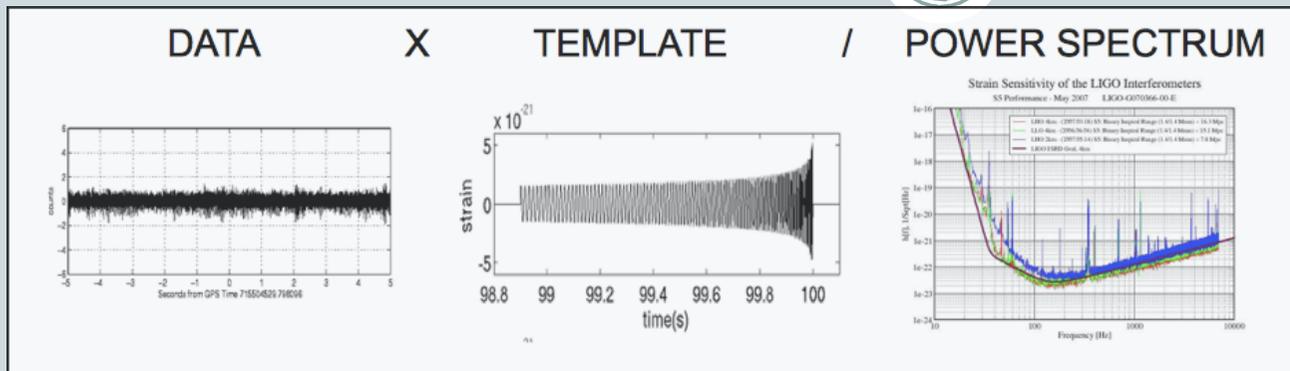


- so our detection statistic is  $D(\{\mathbf{y}\}, \mathbf{i}_0) = \sum_k \varepsilon \mathbf{s}_{k-i_0} \mathbf{y}_k$
- in the continuum  $\sum_k \varepsilon \mathbf{s}_{k-i_0} \mathbf{y}_k \Rightarrow \int_{-\infty}^{+\infty} dt s(t - \tau_0) \mathbf{y}(t)$
- and in the Fourier domain:

$$\rho(\tau_0) = \int d\omega \frac{S^*(\omega, \tau_0) Y(\omega)}{N(\omega)}$$

the standard expression for matched filtering, N being the noise spectrum

# Matched filter



FT of waveform with parameters  $a$ , at  $t_0$

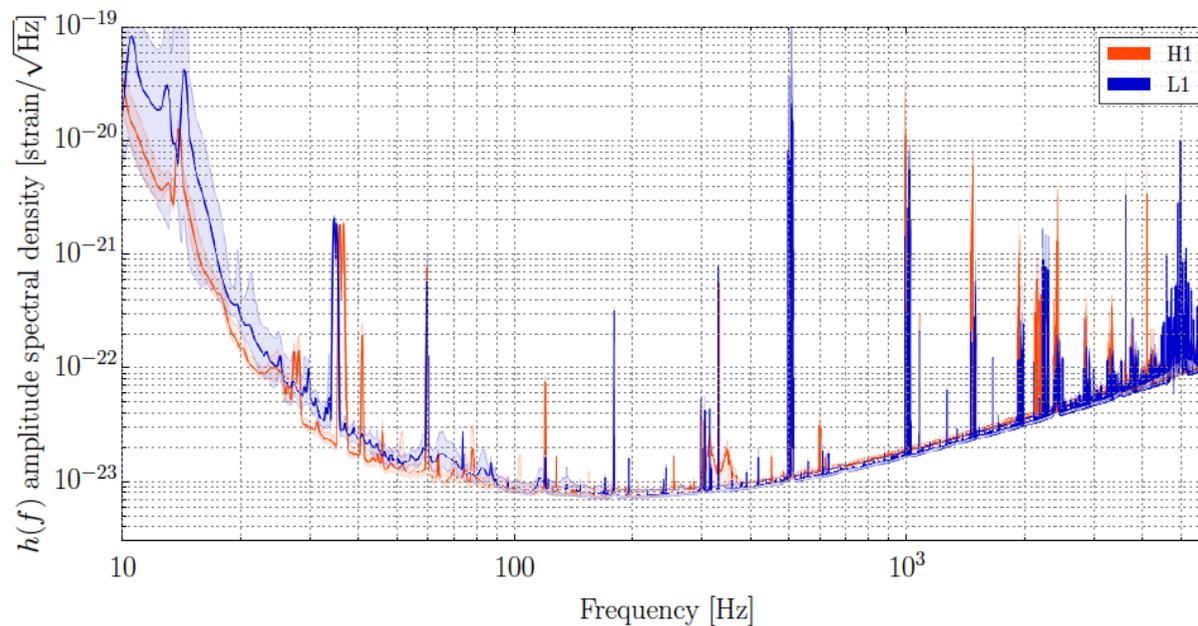
FT of data

$$\rho(t_0, a) = \int d\omega \frac{H^*(\omega, a, t_0) X(\omega)}{S_h(\omega)}$$

noise  $\rightarrow$   $S_h(\omega)$

# What signals may be detectable ?

time scales of ms to s, compact objects, high accelerations:

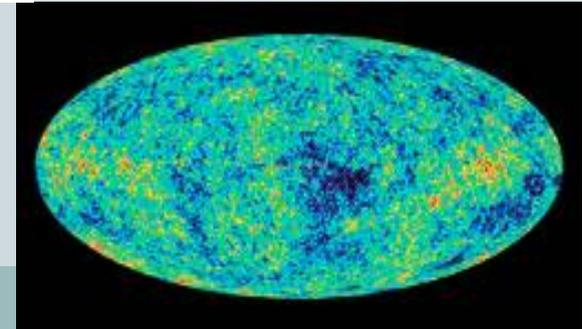
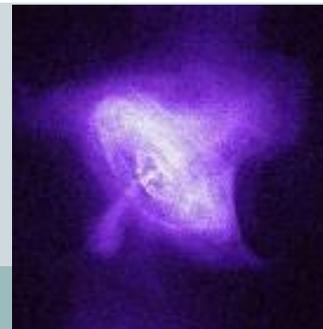
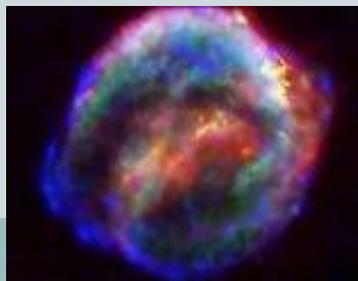


from inspiraling compact objects

bursts, typically arising from catastrophic events

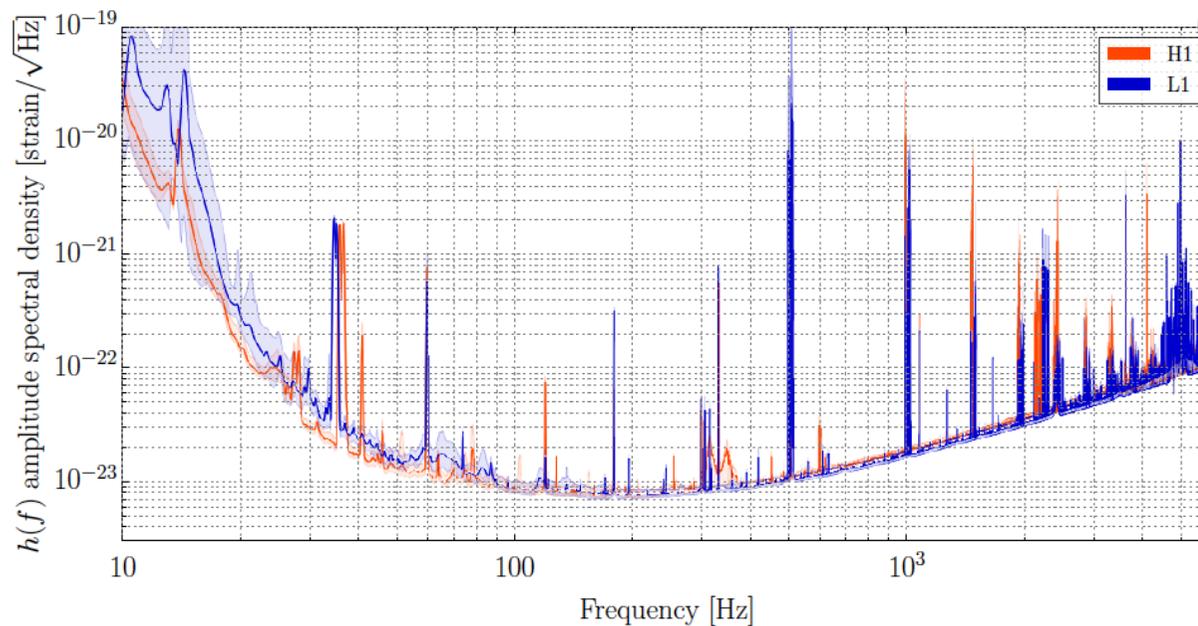
continuous quasi-periodic waves

stochastic background of gravitational radiation



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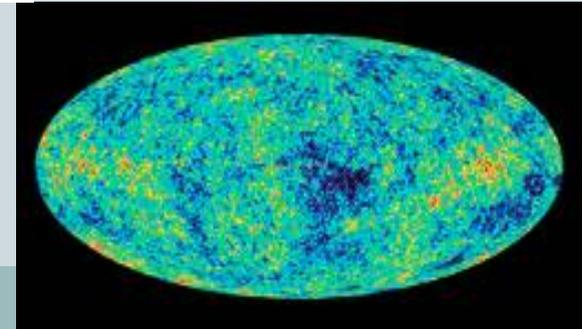
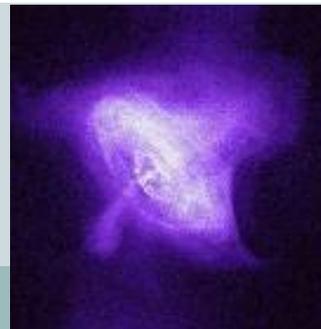
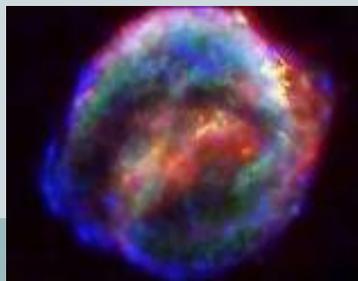


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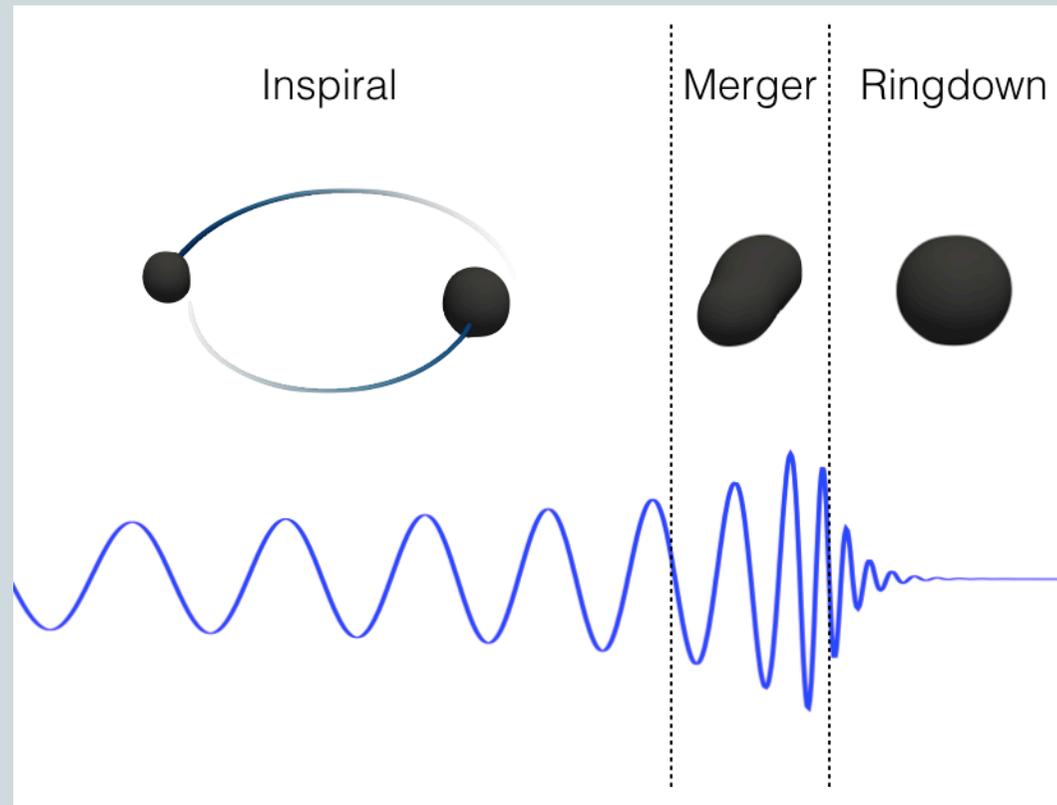
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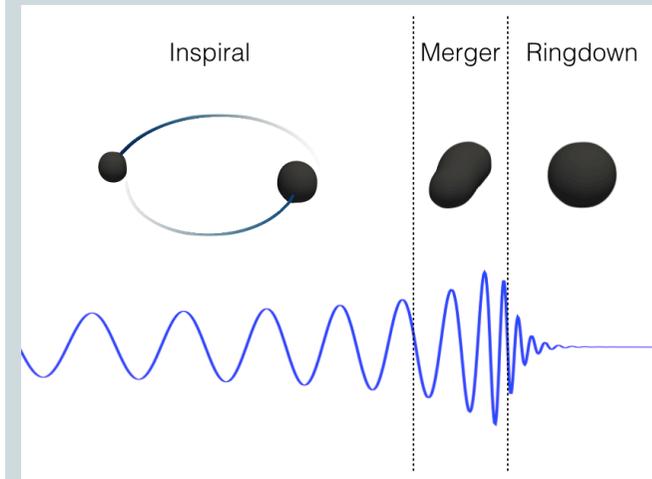
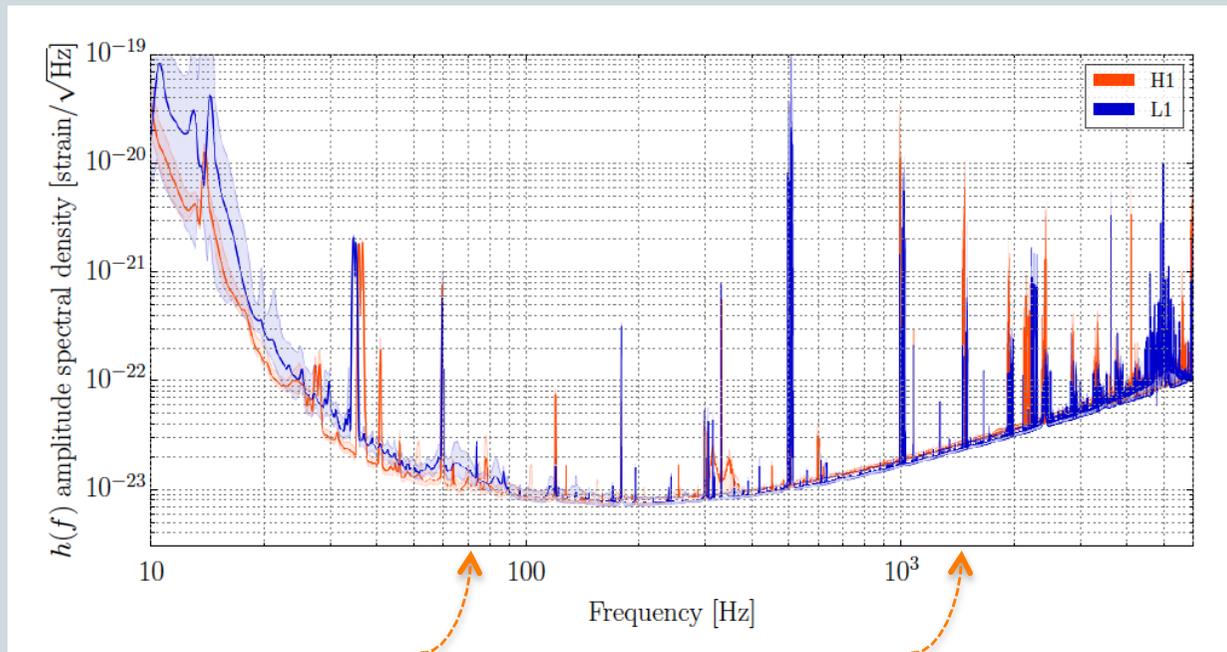
# Compact binary coalescences as sources of GW



- Final evolution of compact binary systems involving neutron stars and/or black holes, driven by gravitational radiation



# What mass ranges do we see ?

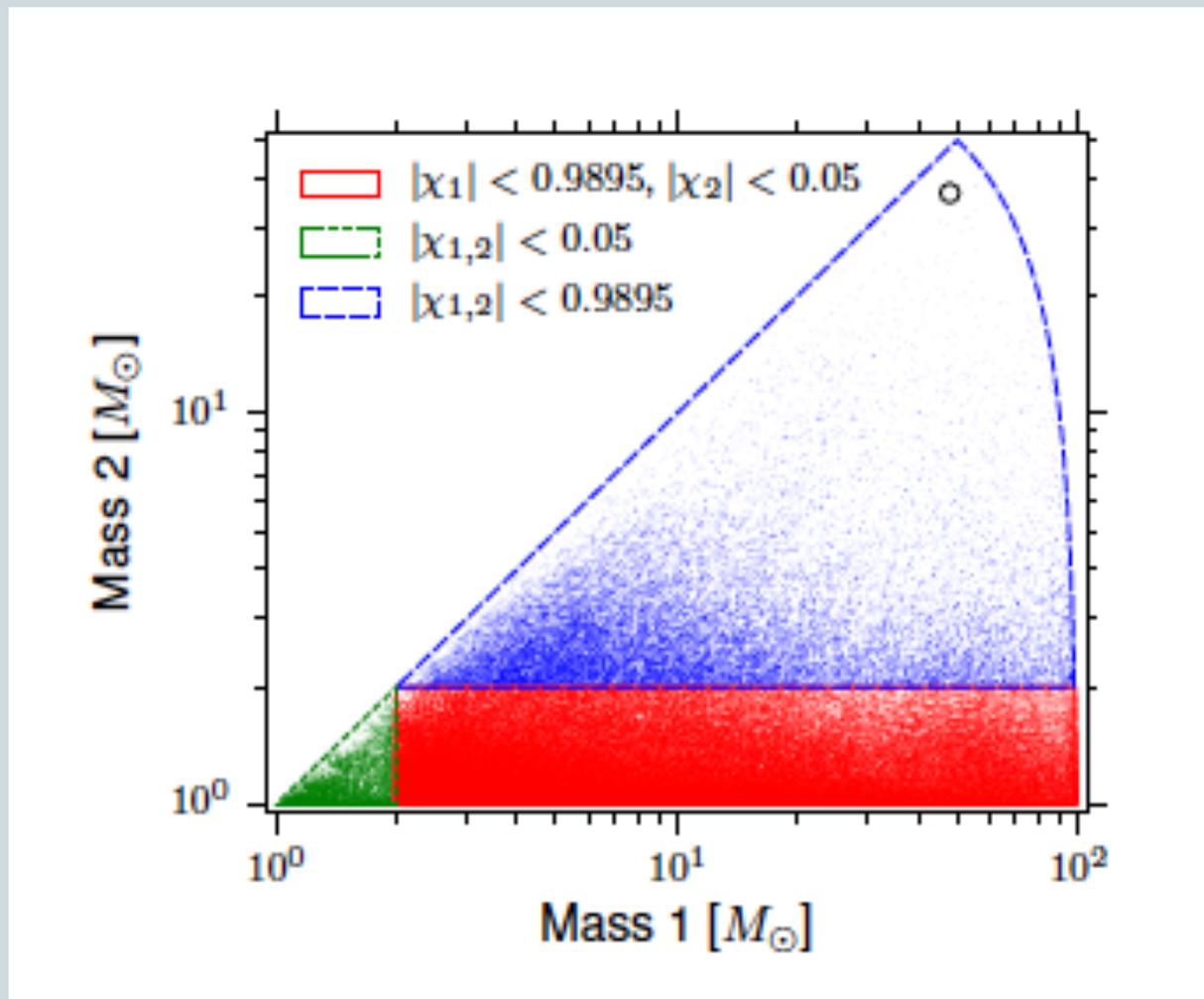


merger freq for  
 $M_{\text{tot}} \approx 60 M_{\odot}$

merger freq for  
 $M_{\text{tot}} \approx 2.4 M_{\odot}$

**4300 Hz  $M_{\odot}/M_{\text{tot}}$**

# Mass range (O1 search)



# How do we search for signals ?

## Matched filter

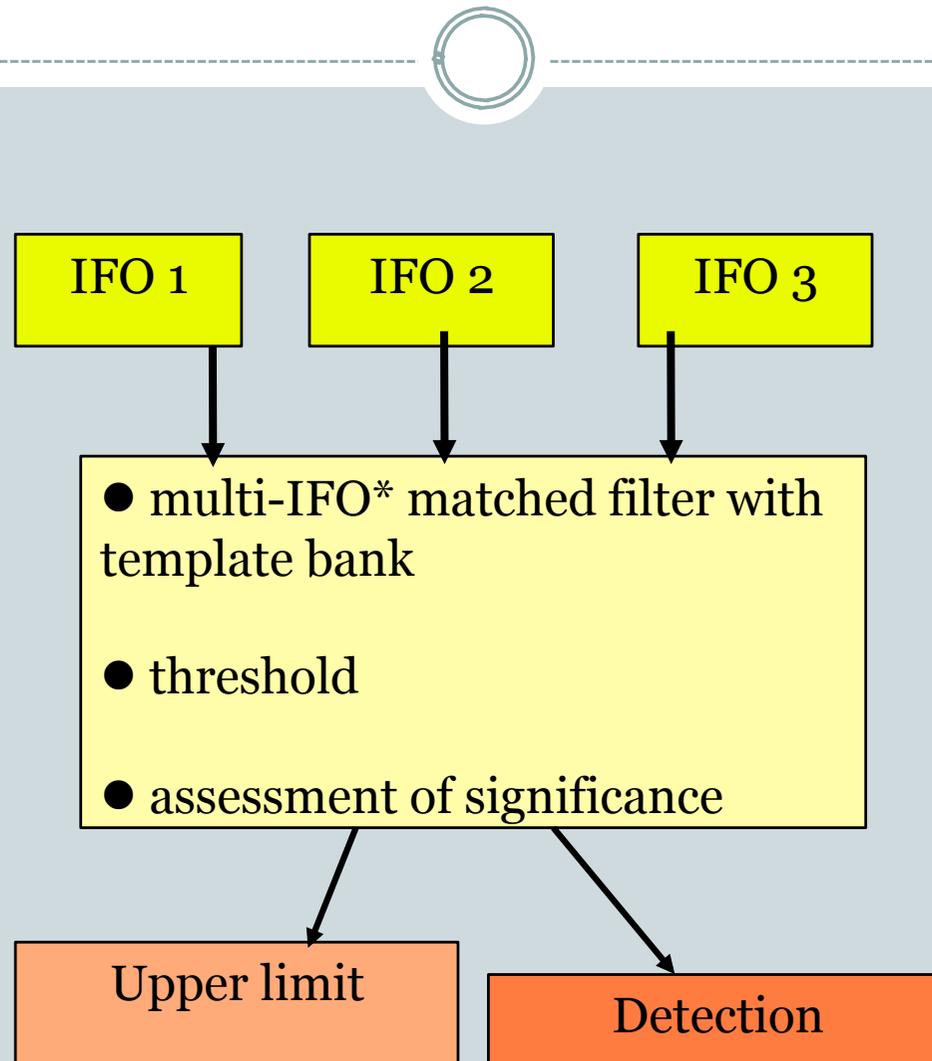
- At best you know what you're looking for; then you use a matched filter:

The diagram illustrates the matched filter equation with labels for its components:

- A box labeled "FT of waveform with parameters a, at t<sub>0</sub>" has an arrow pointing to the  $H^*(\omega, a, t_0)$  term in the numerator.
- A box labeled "FT of data" has an arrow pointing to the  $X(\omega)$  term in the numerator.
- A box labeled "noise" has an arrow pointing to the  $S_h(\omega)$  term in the denominator.

$$\rho(t_0, a) = \int d\omega \frac{H^*(\omega, a, t_0) X(\omega)}{S_h(\omega)}$$

# Idealized pipeline schematics



\*IFO : interferometer

# but it's more complicated:



- the matched filter is optimal detection statistic for Gaussian stationary noise but our data are neither Gaussian nor stationary:
  - Weed-out spurious noise:
    - ✦ Data quality flags
    - ✦ Coincidence schemes
    - ✦ Signal-based noise rejection techniques
  - Ad-hoc inspection of interesting candidates:
    - ✦ Correlations with environmental channel
    - ✦ Examine overall status of detectors
  - But the problem remains of assessing the significance
    - ✦ Problem of background/noise estimation

# The problem with large spurious noise events



**Matched filter:** is designed to give a large response when the signal waveform matches the template, but it also gives a large response when the instrumental noise has a large glitch. Even if the glitch shape looks nothing like a waveform, it can still drive the filter to give a large response.

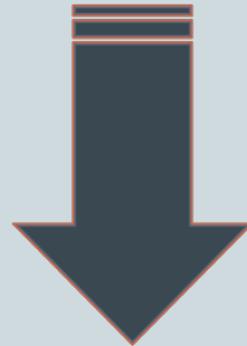
**Noisy data:** the noise of *GW* detectors presents sporadic prominent non-Gaussian glitches. This is a problem.

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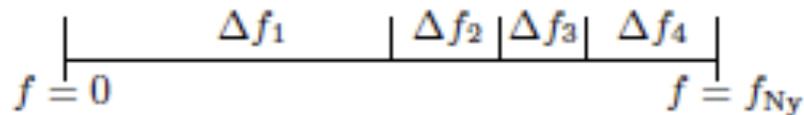
## MITIGATION SCHEMES

# A counter-measure: signal-based veto, the $\chi^2$ test



If it looks like a duck, quacks like a duck, swims like a duck, then it is a duck.

# Does it *really* look like a duck ? the $\chi^2$ test



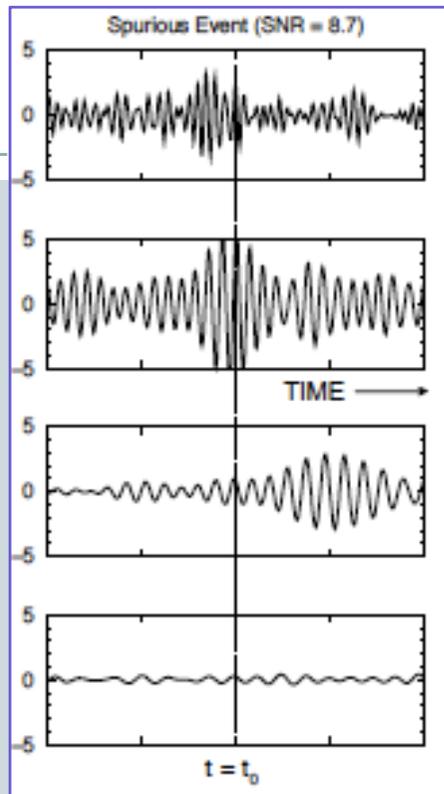
**Main idea:** consider  $p$  frequency sub-bands each contributing the same to the matched filter SNR,  $z$ , if a signal is present. Compute the matched filter detection statistic  $z_j$  for each of the sub-bands and verify that this is the case:

$$\chi^2 = \frac{p}{2p-2} \sum_{j=1}^p \left( z_j - \frac{z}{p} \right)^2 \quad E[\chi^2] = p-1 \quad \text{var}[\chi^2] = 2(p-1)$$

if the hypothesis is correct the residuals are random Gaussian variables and their square sum a chi square variable.

notation note :  $\rho \leftrightarrow z$  here

# An example



$z_4(t)$

$z_3(t)$

$z_2(t)$

$z_1(t)$

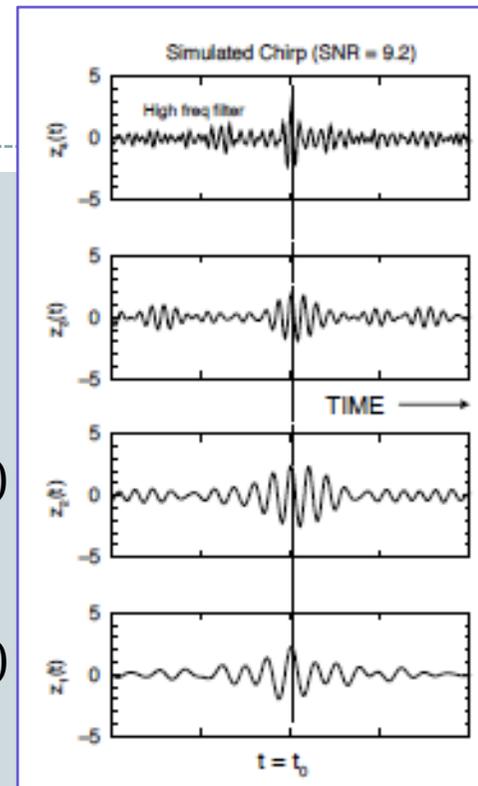


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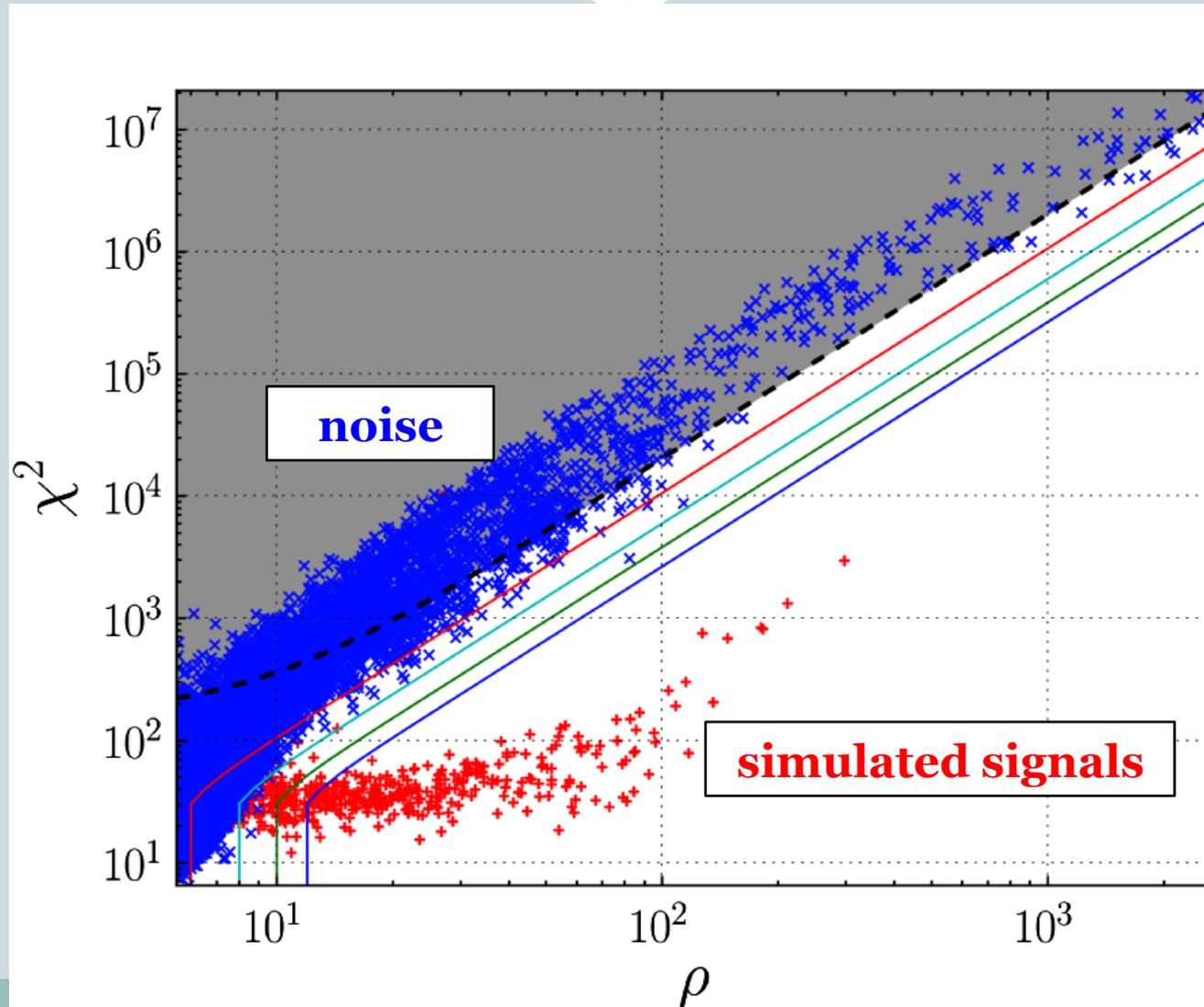


$$\begin{aligned}
 z_1 &= 0.23 \\
 z_2 &= 0.84 \\
 z_3 &= 5.57 \\
 z_4 &= 2.33 \\
 z &= z_1 + z_2 + z_3 + z_4 = 8.97 \\
 \chi^2 &= 4 \sum_{j=1}^4 (z_j - z/4)^2 = 68.4 \\
 P_{\chi^2 \geq 68.4} &= 1 - \frac{\gamma(3/2, 34.2)}{\Gamma(3/2)} = 9.4 \times 10^{-15}.
 \end{aligned}$$

$$\begin{aligned}
 z_1 &= 2.25 \\
 z_2 &= 2.44 \\
 z_3 &= 1.87 \\
 z_4 &= 2.64 \\
 z &= z_1 + z_2 + z_3 + z_4 = 9.2 \\
 \chi^2 &= 4 \sum_{j=1}^4 (z_j - z/4)^2 = 1.296 \\
 P_{\chi^2 \geq 1.296} &= 1 - \frac{\gamma(3/2, 0.648)}{\Gamma(3/2)} = 73\%.
 \end{aligned}$$

notation note :  $\rho \leftrightarrow z$  here

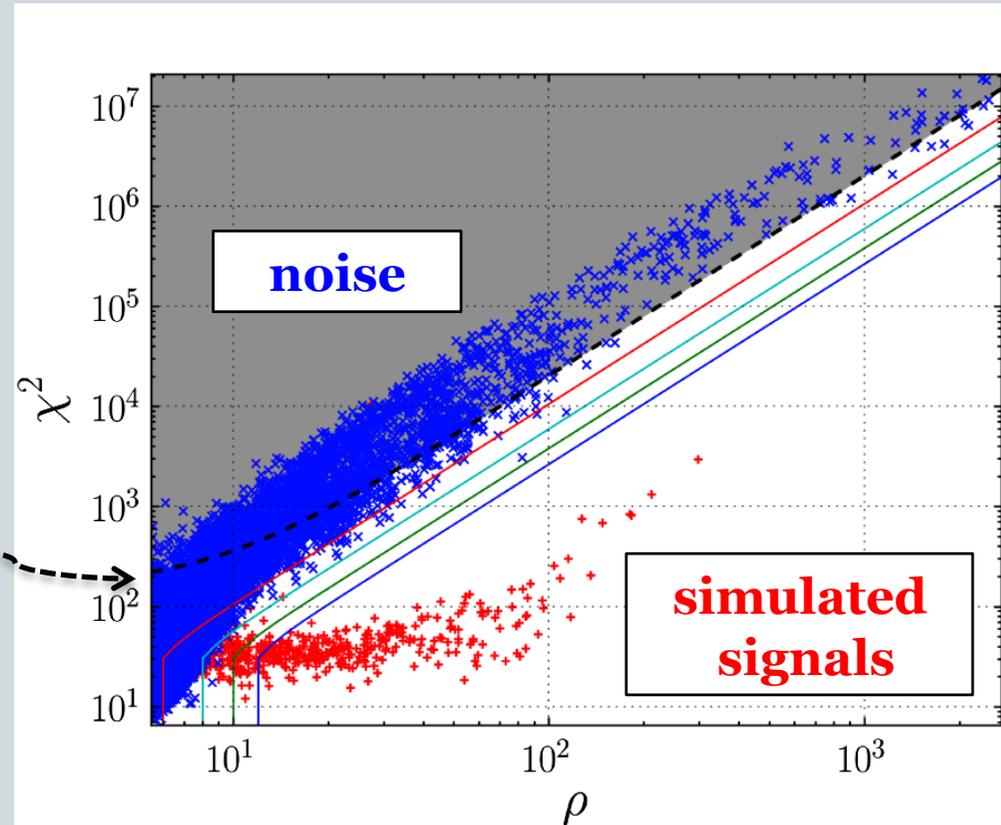
# Cannot just threshold on $\chi^2$



# Use of $\chi^2$ : a veto



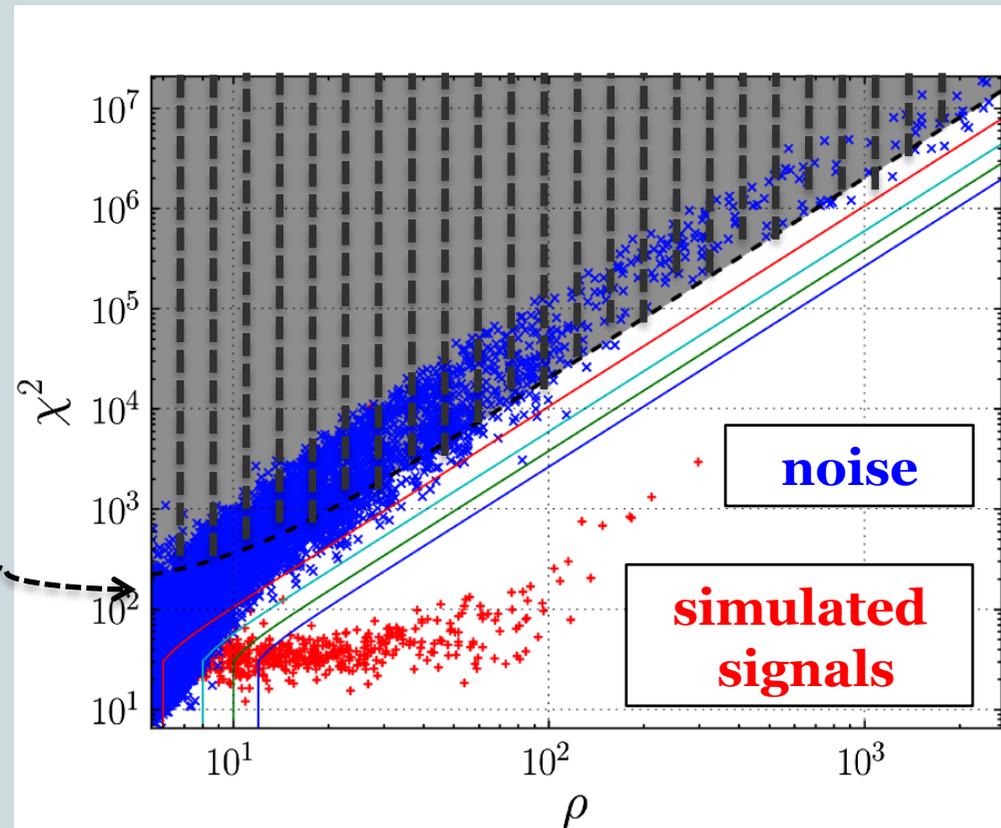
- in previous example number of freq bins  $p=4$  usually  $p=16$
- $n_{\text{dof}} = 2p-2$
- veto all triggers with  $\chi^2 > 10 (p + 0.2 \rho^2)$



# Use of $\chi^2$ : a veto



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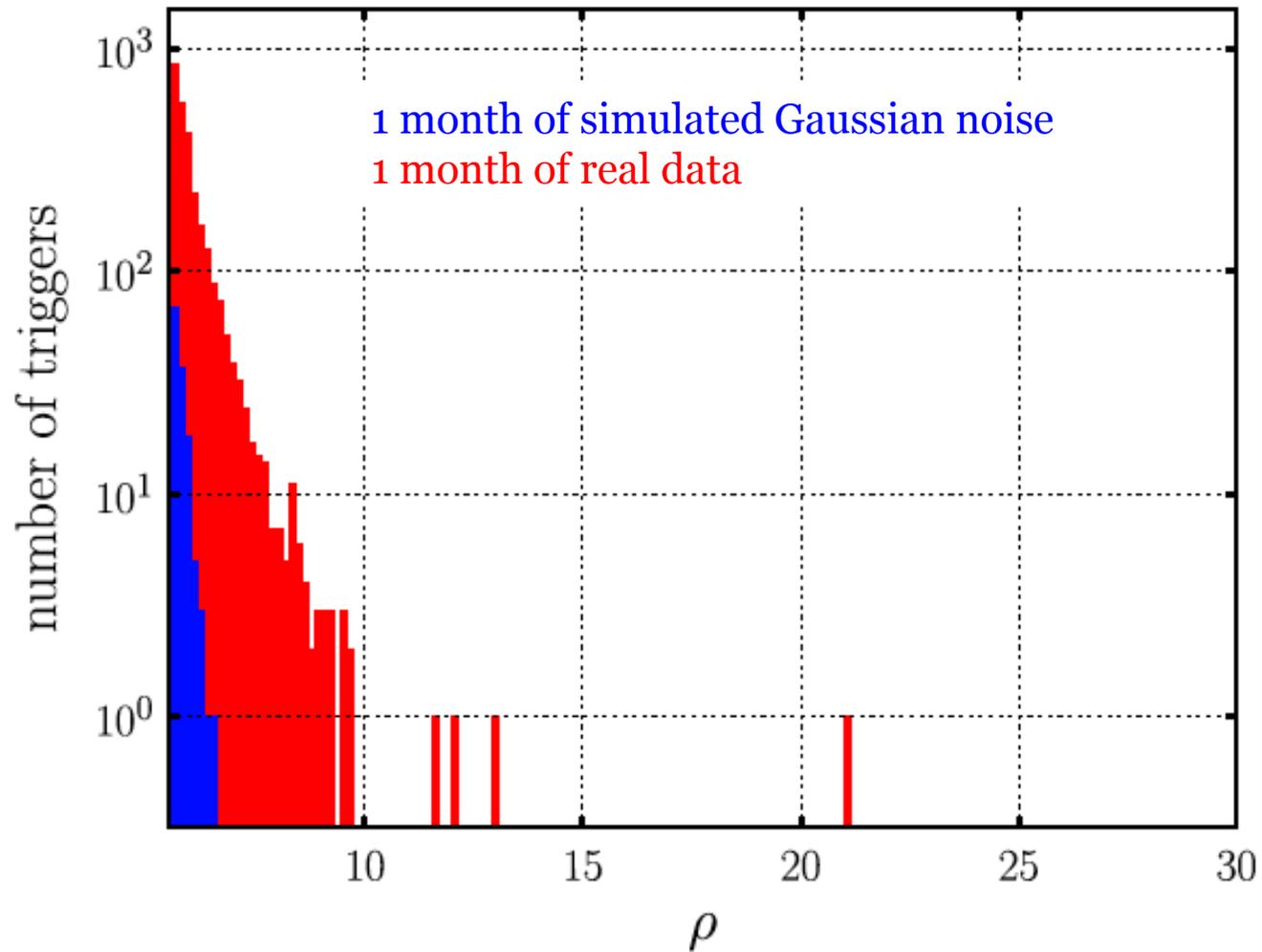
# Coincidence requirement



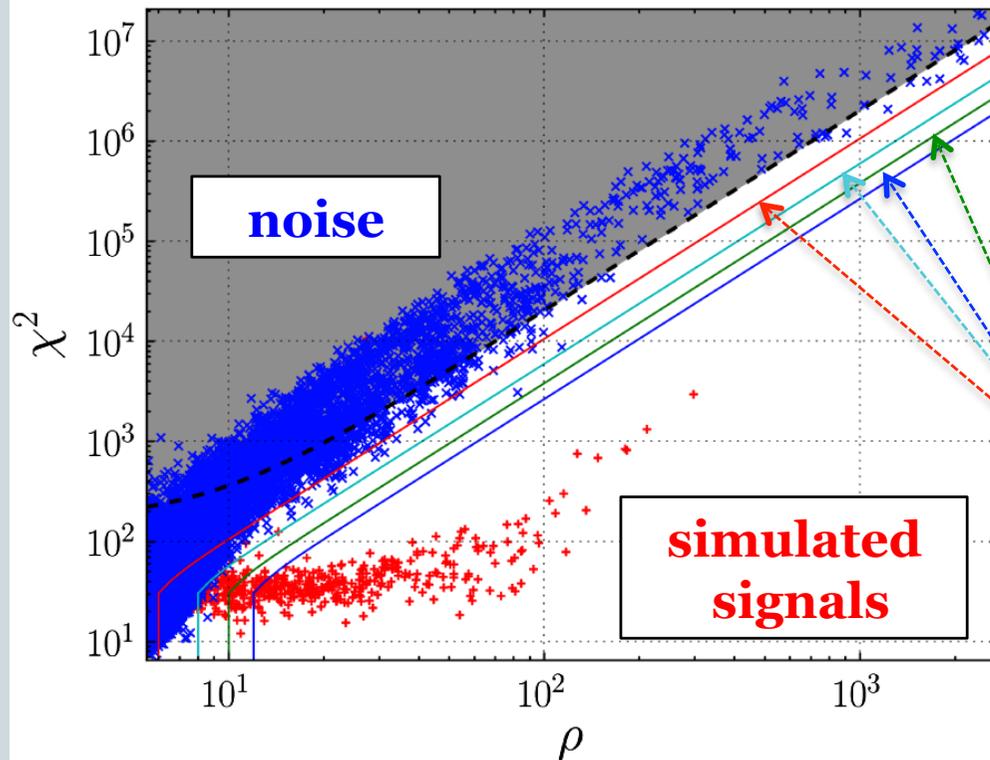
- After the  $\chi^2$  veto we also veto triggers in one detector that do not have a consistent counterpart trigger in the other detector
  - Consistency in waveform parameters
  - Close enough in time



# But still many spurious triggers



# Go back to single-detector triggers and construct a new detection statistic

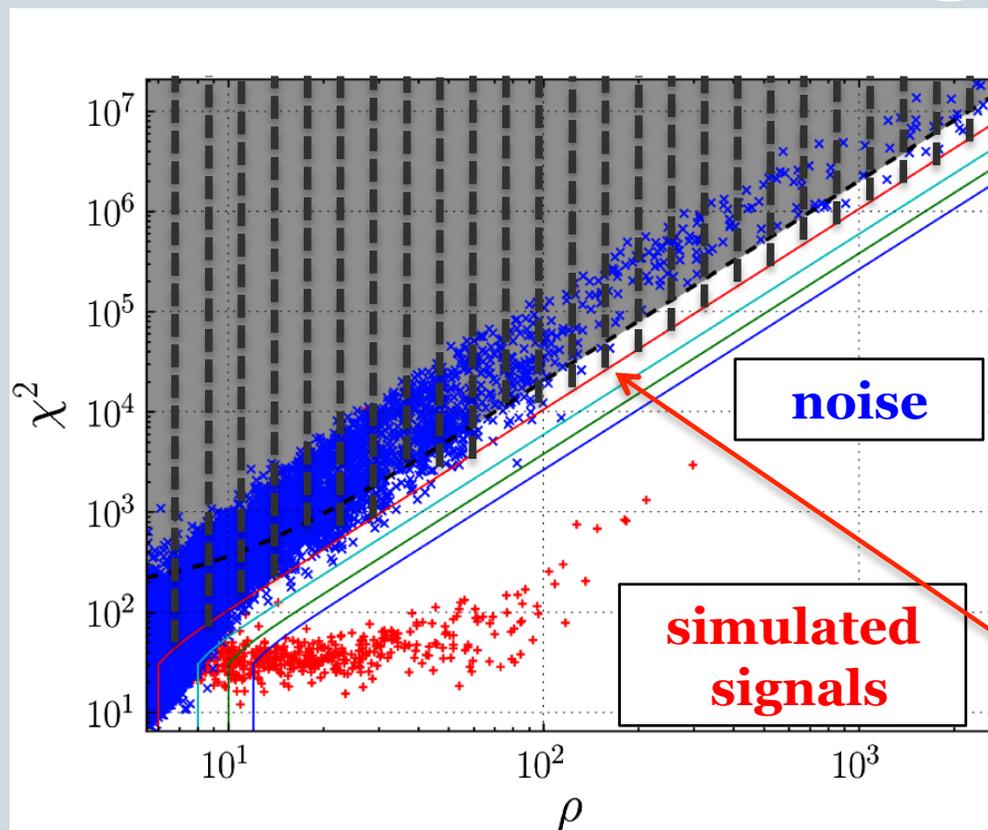


$$\rho_{\text{new}}^2 = \begin{cases} \rho^2 & \text{if } \chi_r^2 \leq 1 \\ \frac{\rho^2}{\left\{ \frac{1}{2} [1 + (\chi_r^2)^3] \right\}^{1/3}} & \text{if } \chi_r^2 > 1 \end{cases}$$

with  $\chi_r^2 = \chi^2 / n_{\text{dof}}$

This statistic matches the constant false alarm contours and discriminates well between signals and noise.

# A new detection statistic



$$\rho_{\text{new}}^2 = \begin{cases} \rho^2 & \text{if } \chi_r^2 \leq 1 \\ \frac{\rho^2}{\left\{ \frac{1}{2} [1 + (\chi_r^2)^3] \right\}^{1/3}} & \text{if } \chi_r^2 > 1 \end{cases}$$

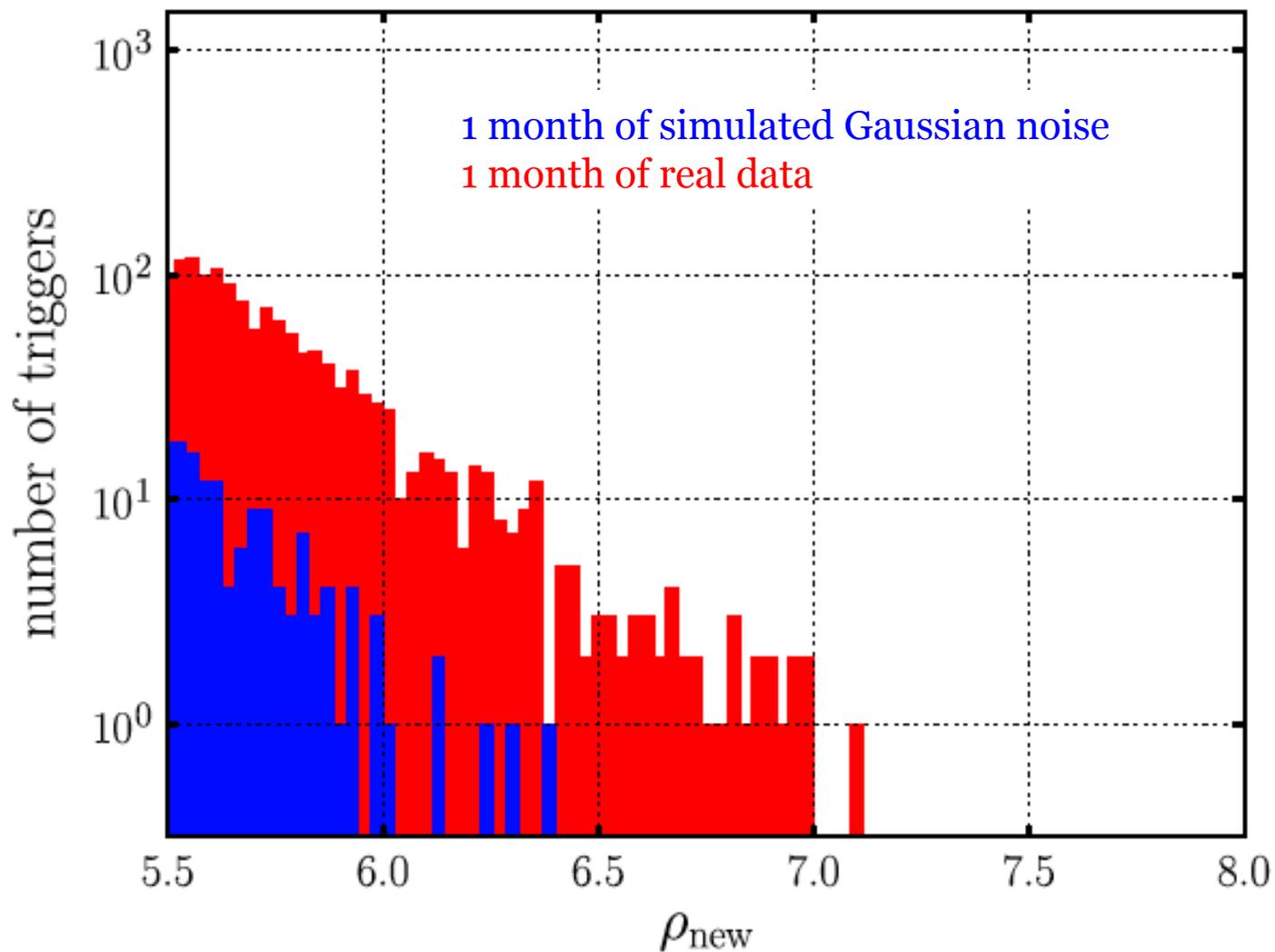
with  $\chi_r^2 = \chi^2 / n_{\text{dof}}$

This statistic matches the false alarm contours and discriminates well between signals and noise.

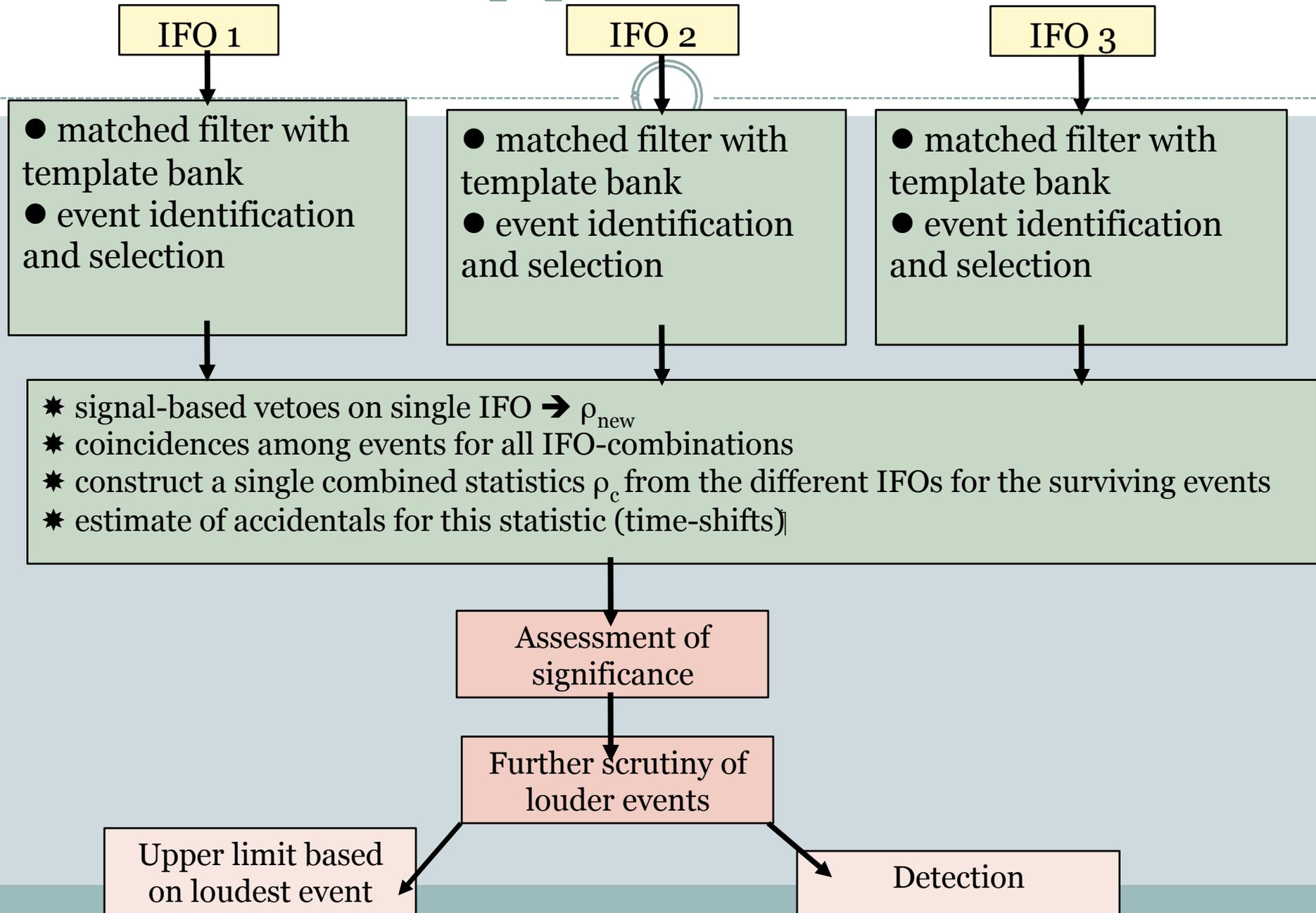
Triggers with  $\rho_{\text{new}}^2 < 5.5$  are vetoed  
Then again coincidences

The final combined detection statistic:  $\rho_c^2 = \sum \rho_{\text{new},i}^2$

# Remarkable improvement



# Actual pipeline schematics



# Assessment of significance

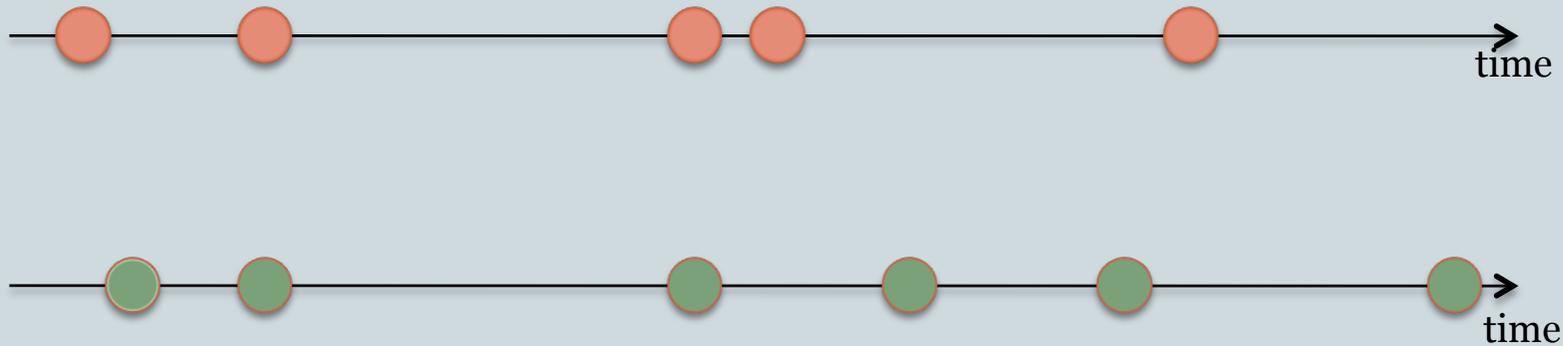


- The analysis produces a list of coincident triggers, each with an associated combined SNR,  $\rho_c$ .
- These triggers need to be compared with those that one would obtain by chance, i.e. the accidentals, the background. We do this by comparing the distributions.
- How do we estimate the background? We repeat the analysis on off-source data (by time-shifting the data streams).
- If an on-source coincidence trigger is significantly above the estimated background, then it is a candidate event that warrants further inspection.

# Time-slides



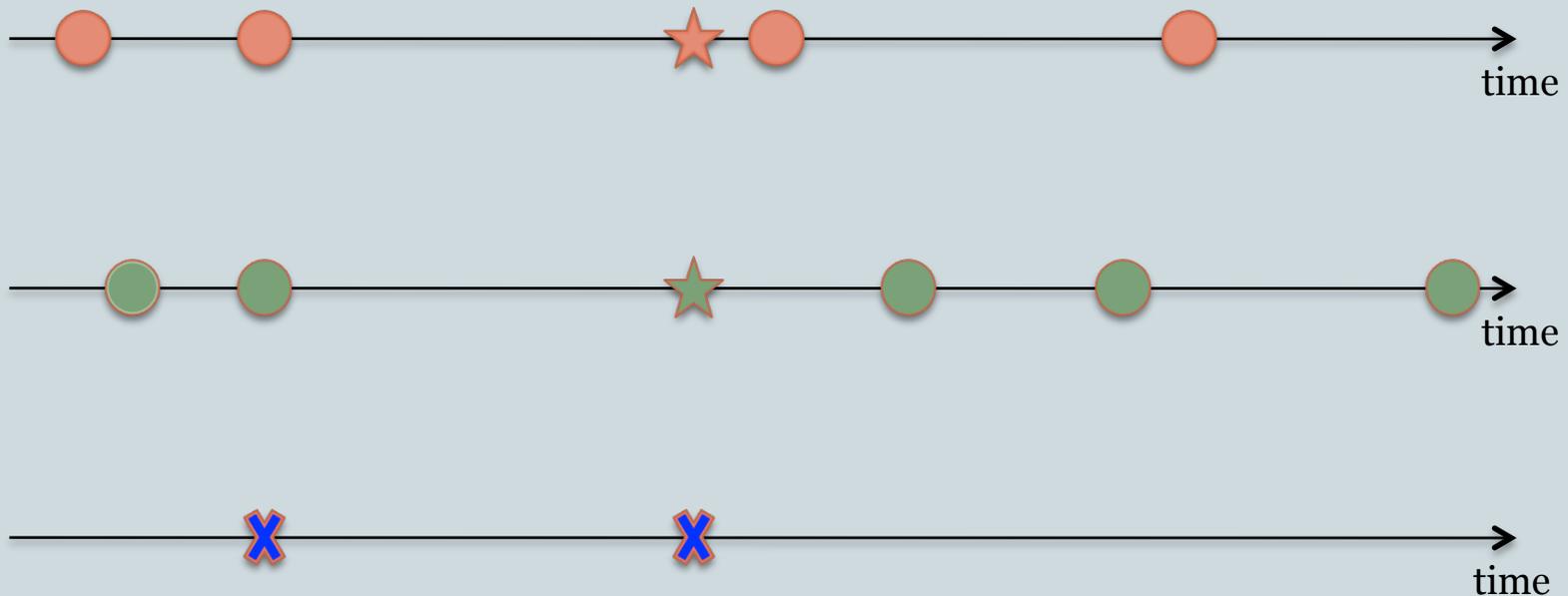
- triggers from detector 1
- triggers from detector 2



# Time-slides



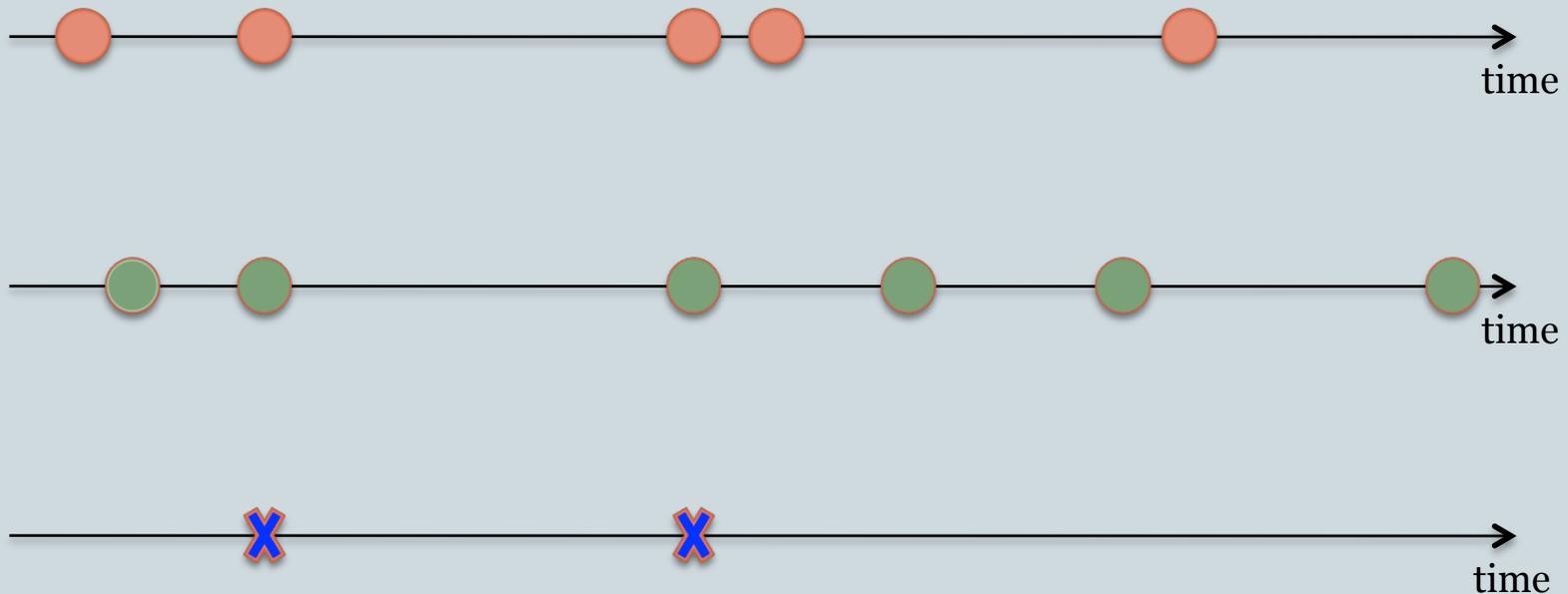
- triggers from detector 1
- triggers from detector 2
- ✕ coincident triggers
- ★ trigger due to GWs
- ★ trigger due to GWs



# Time-slides



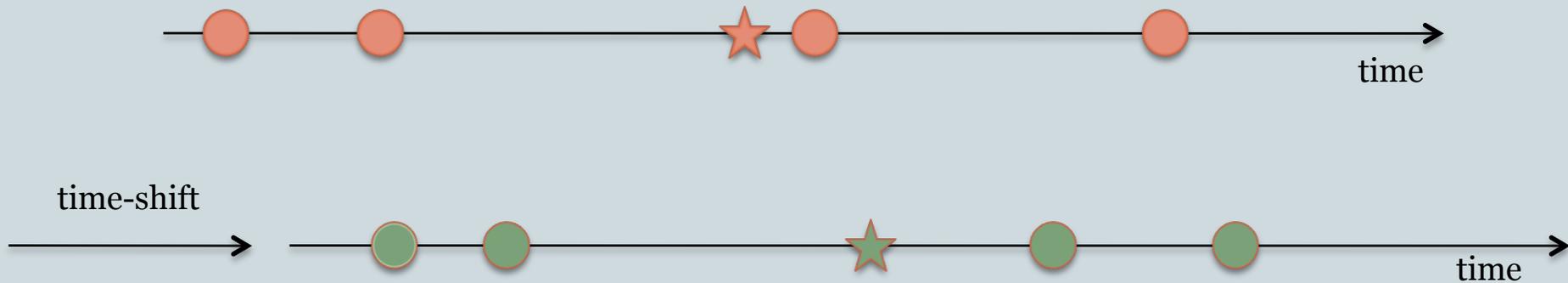
- triggers from detector 1
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# Time-slides



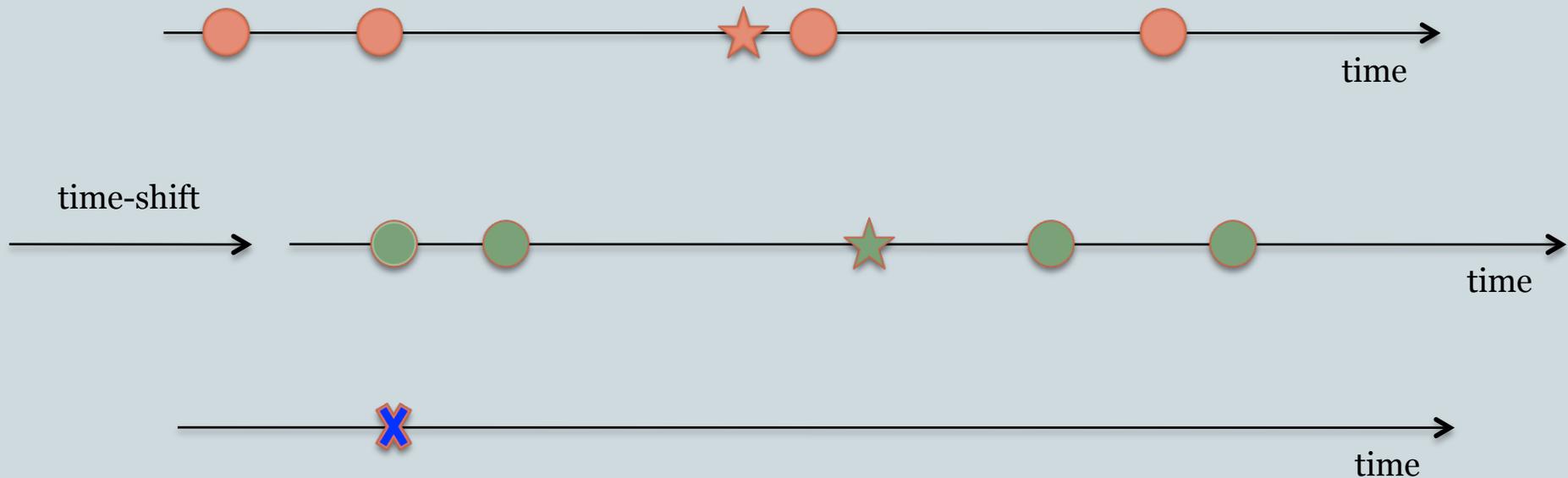
- ★ trigger from detector 1 due to a GW
- ★ trigger from detector 2 due to a GW



# Time-slides



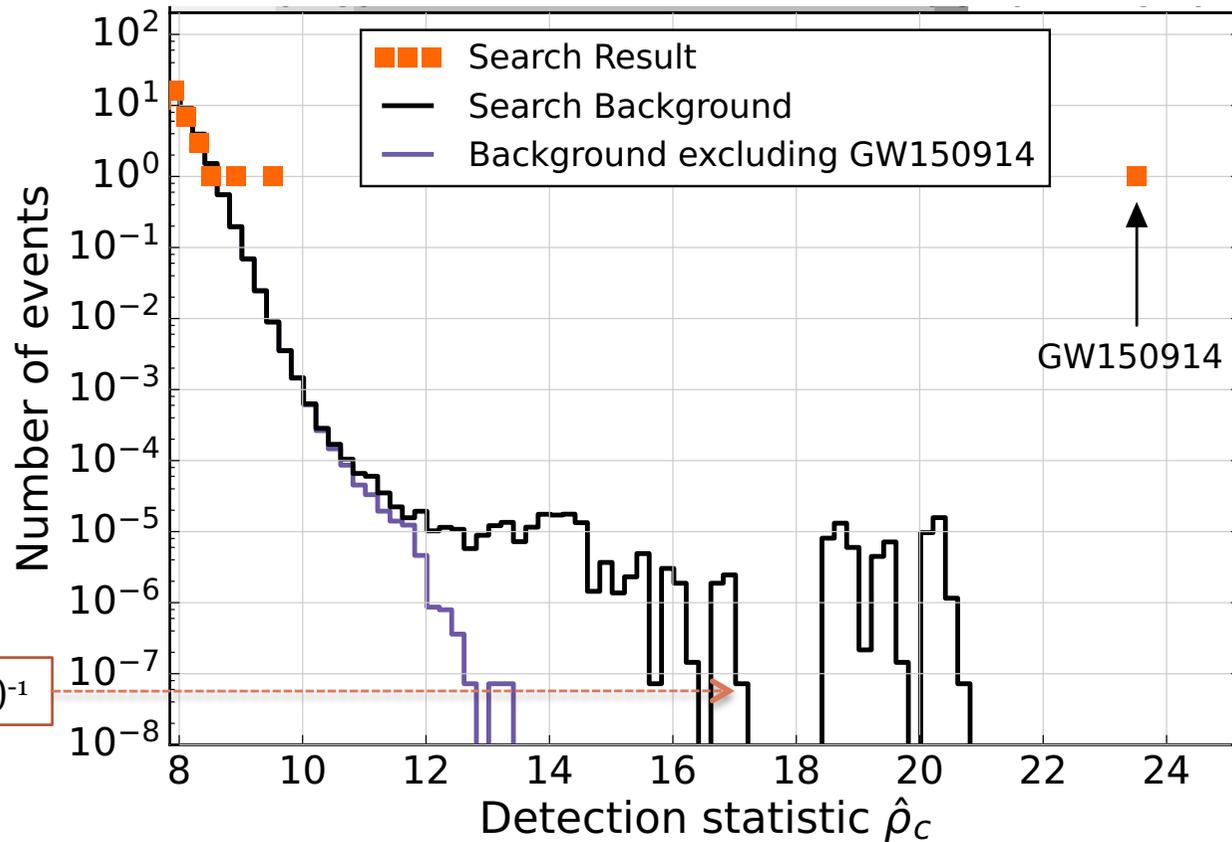
- ★ trigger from detector 1 due to a GW
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- ✕ coincident triggers



# The first GW signal (GW150914)



$1.4 \times 10^7$  time slides corresponding to 608 000 yrs of simulated background.

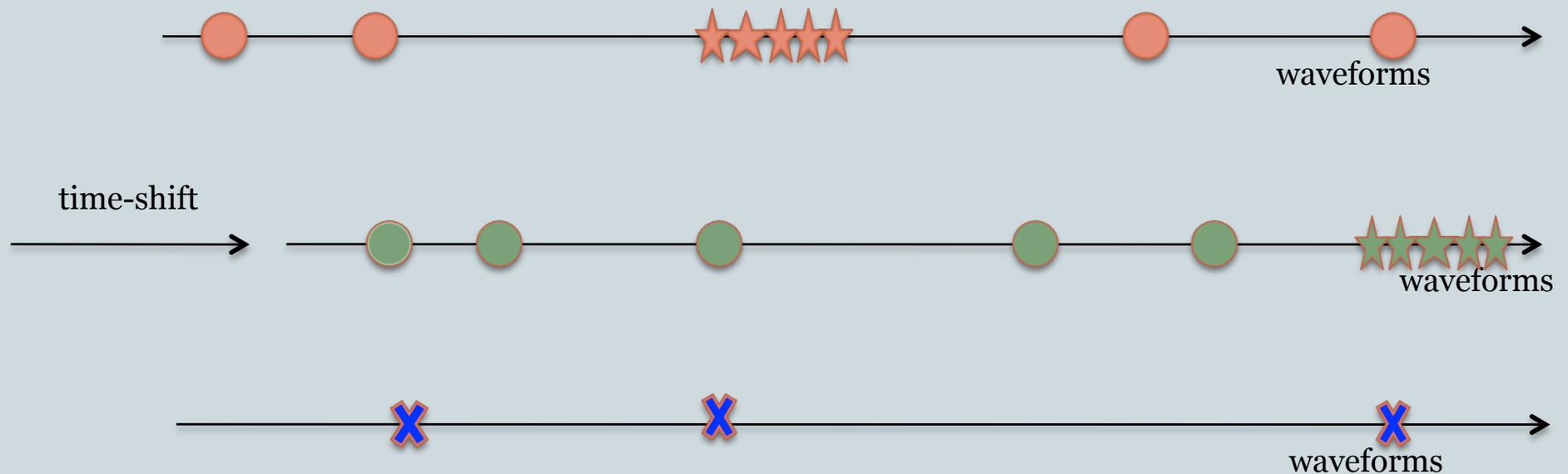


$7 \times 10^{-8} \approx (1.4 \times 10^7)^{-1}$

# Time-slides: a conservative estimate of the background



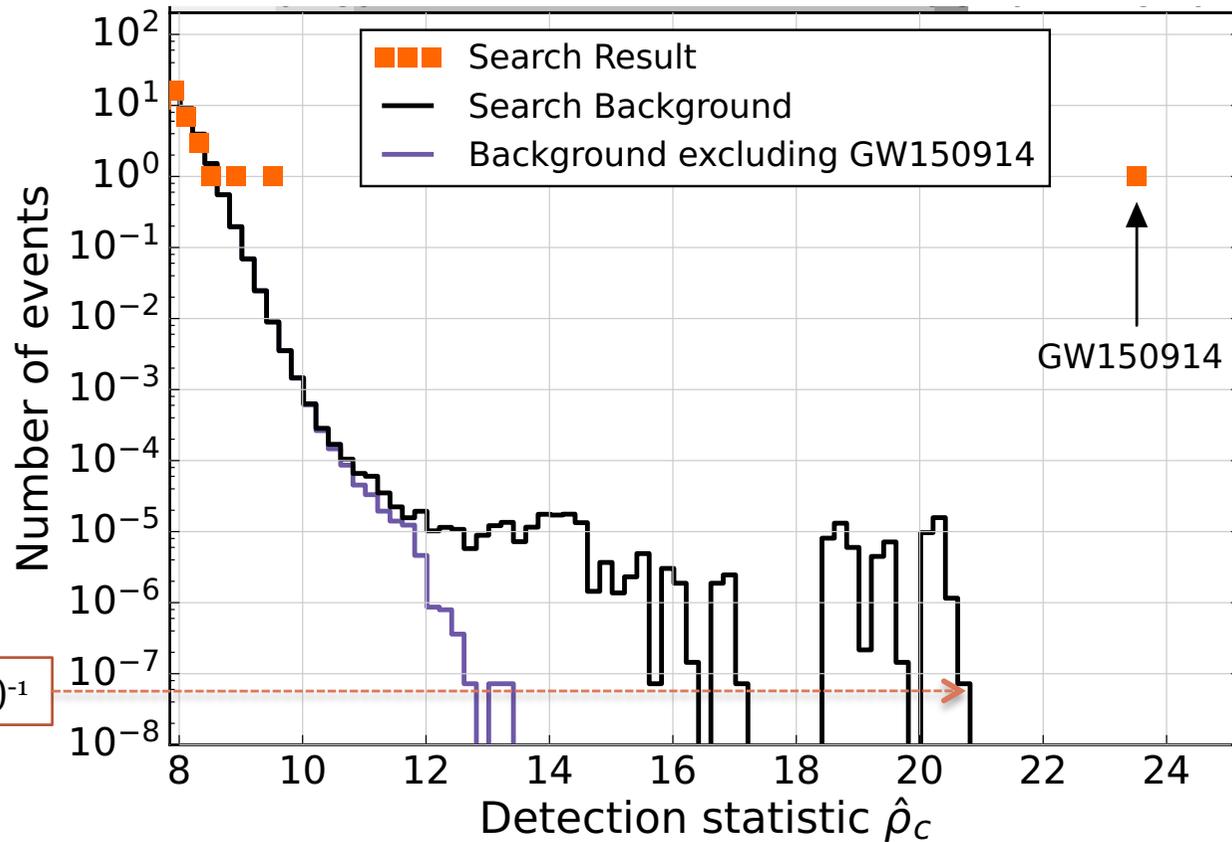
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- ✕ coincident triggers



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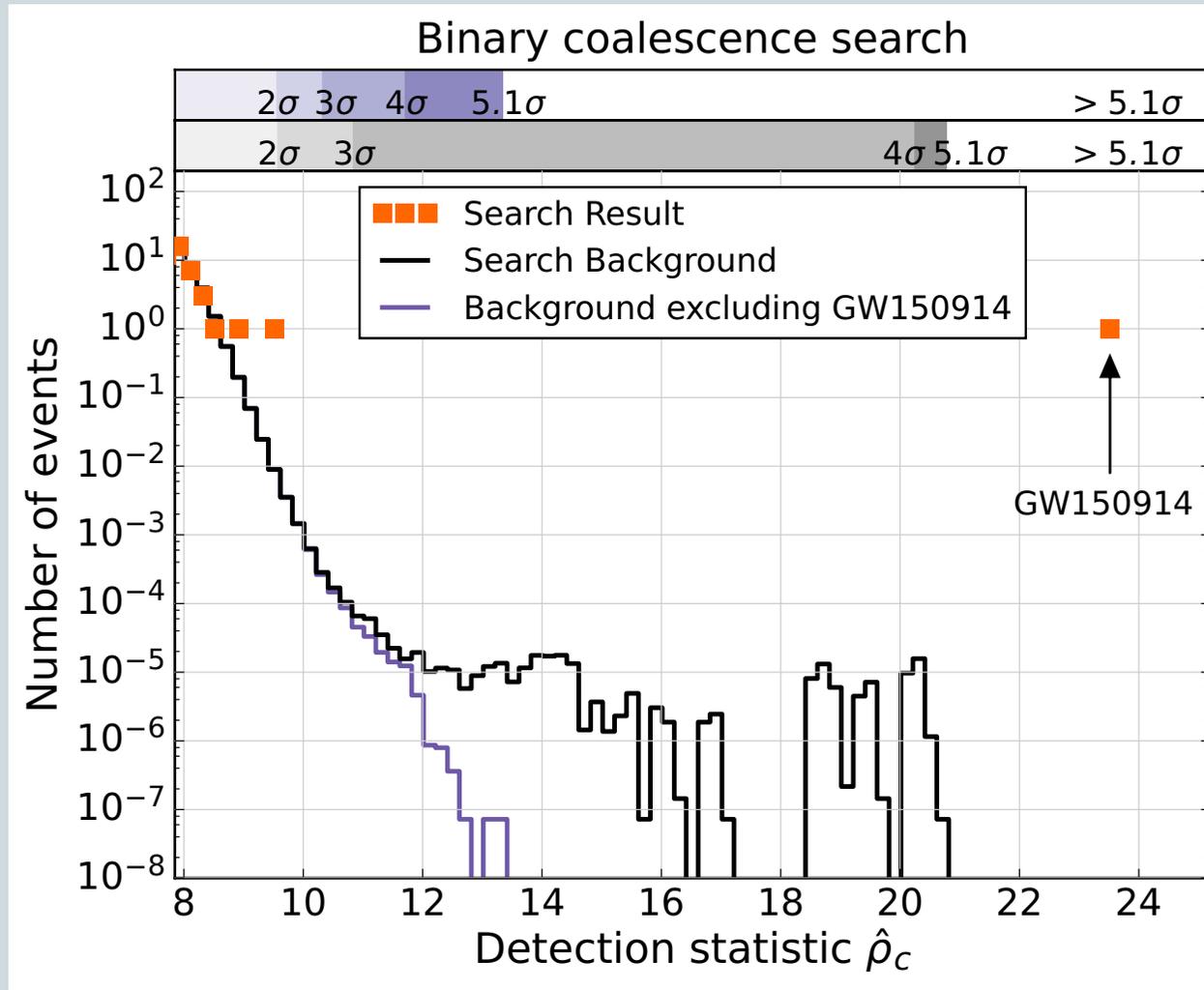
$7 \times 10^{-8} \approx (1.4 \times 10^7)^{-1}$

# Significance numbers for GW150914



- Analysis time (livedtime): 14.14 days
- Could not get as significant detection statistic value in  $1.375 \times 10^7$  realizations of the experiment.
- This corresponds to an analysis background time of  $1.375 \times 10^7 \times 14.14 \text{ days} / 365 \text{ days} = 608\,000 \text{ yrs}$
- Let's take the most significant event at  $\rho_c \sim 21$
- False alarm rate (FAR) = 1 event / background-time =  $1.6 \times 10^{-6} \text{ yr}^{-1}$
- FAR  $\rightarrow$  3 x FAR X because 3 independent searches were performed (trials factor). FAR =  $4.9 \times 10^{-6} \text{ yr}^{-1}$
- Poisson process with  $\lambda = \text{FAR} \times \text{livedtime} = 2 \times 10^{-7}$
- The probability to measure one event or more in a Poisson process with that average rate  $\lambda$  is FAP =  $2 \times 10^{-7}$  (it's the same as  $\lambda$  because  $\lambda \ll 1$ )
- The Gaussian sigma level corresponding to such FAP is 5.1

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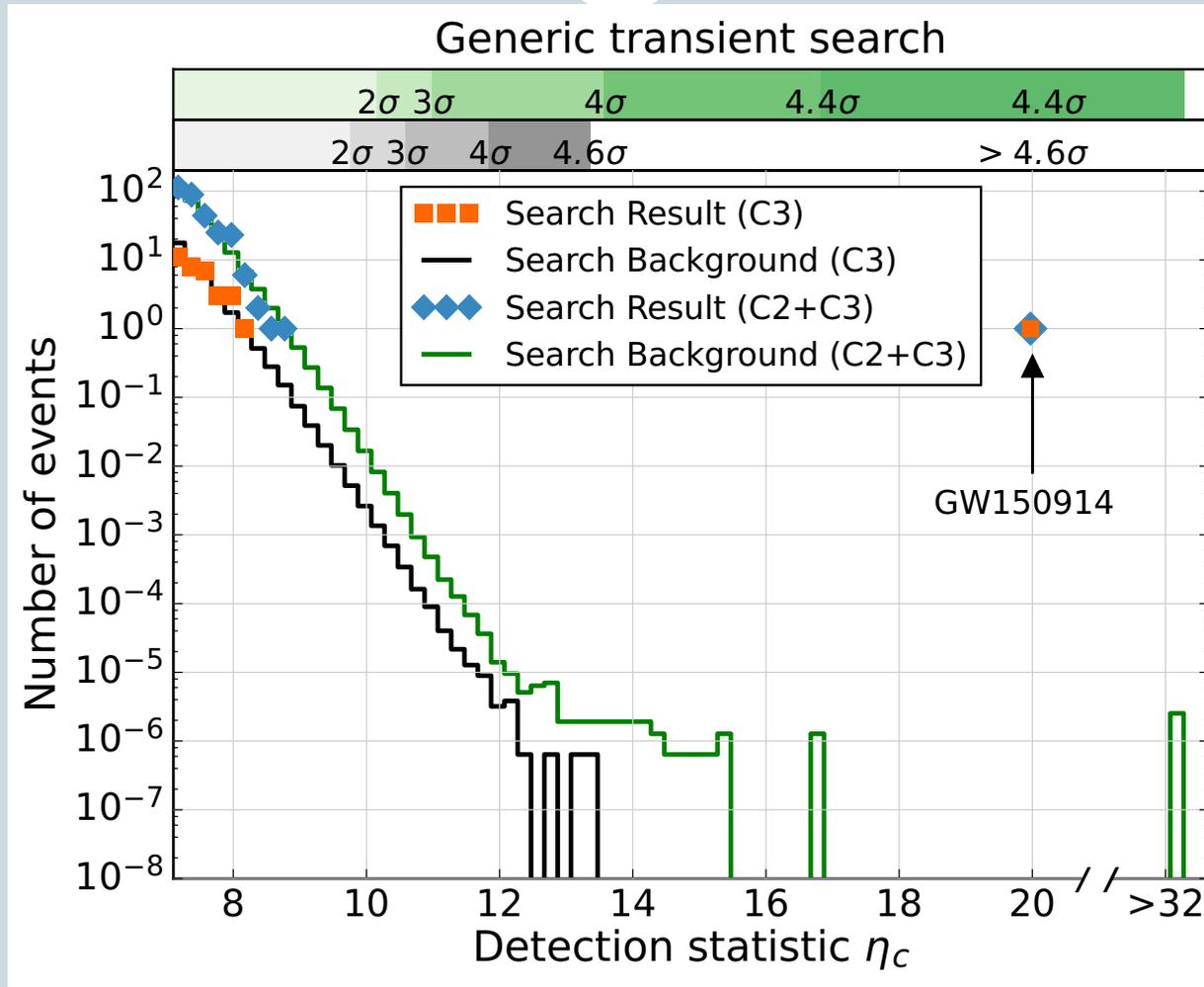


# Further inspection means

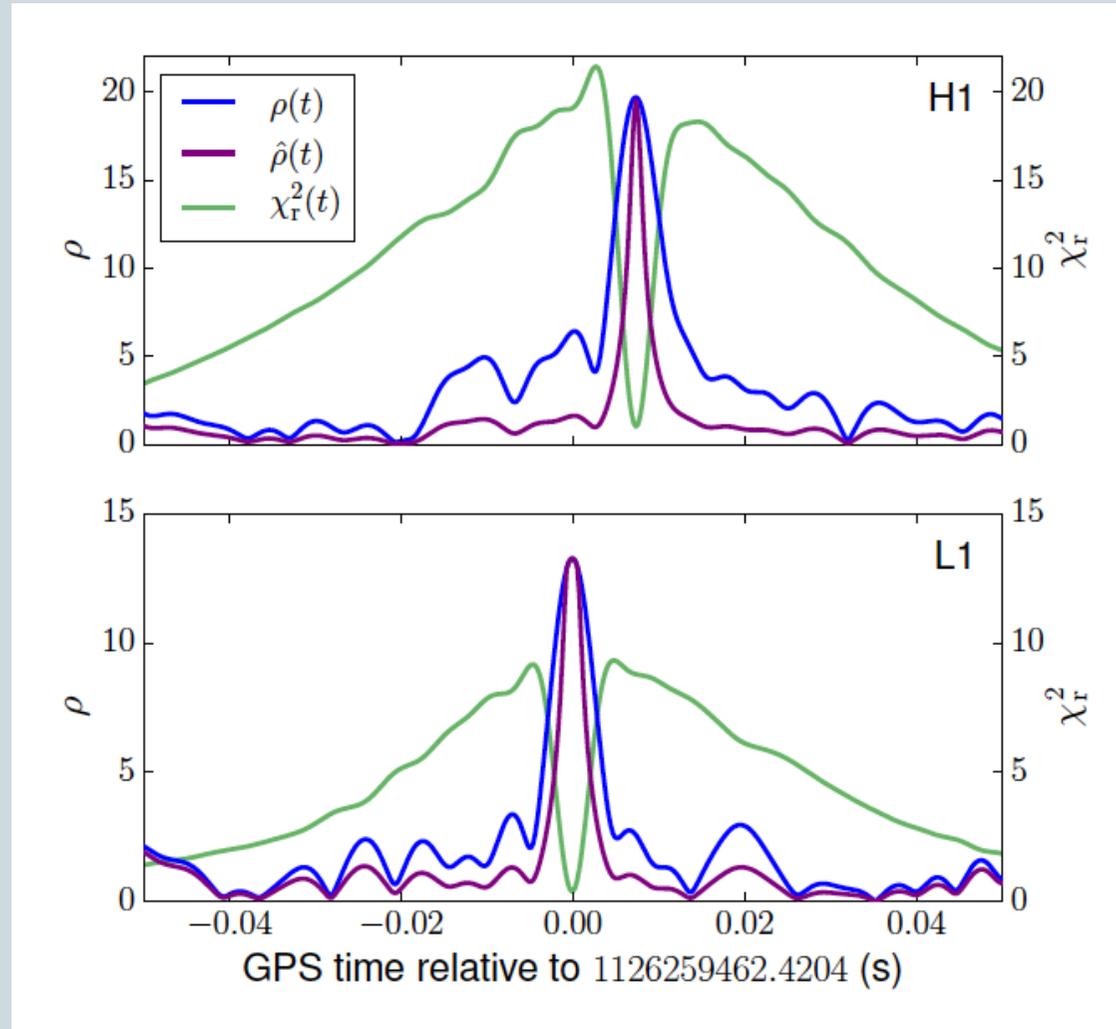


- Statistical significance of the candidate (cross check with other pipelines)
- Status of the interferometers
- Check for environmental or instrumental causes
- Check intermediate stages of the analysis
- Check for coincidences with non-GW searches: other E/M or particle detectors when relevant

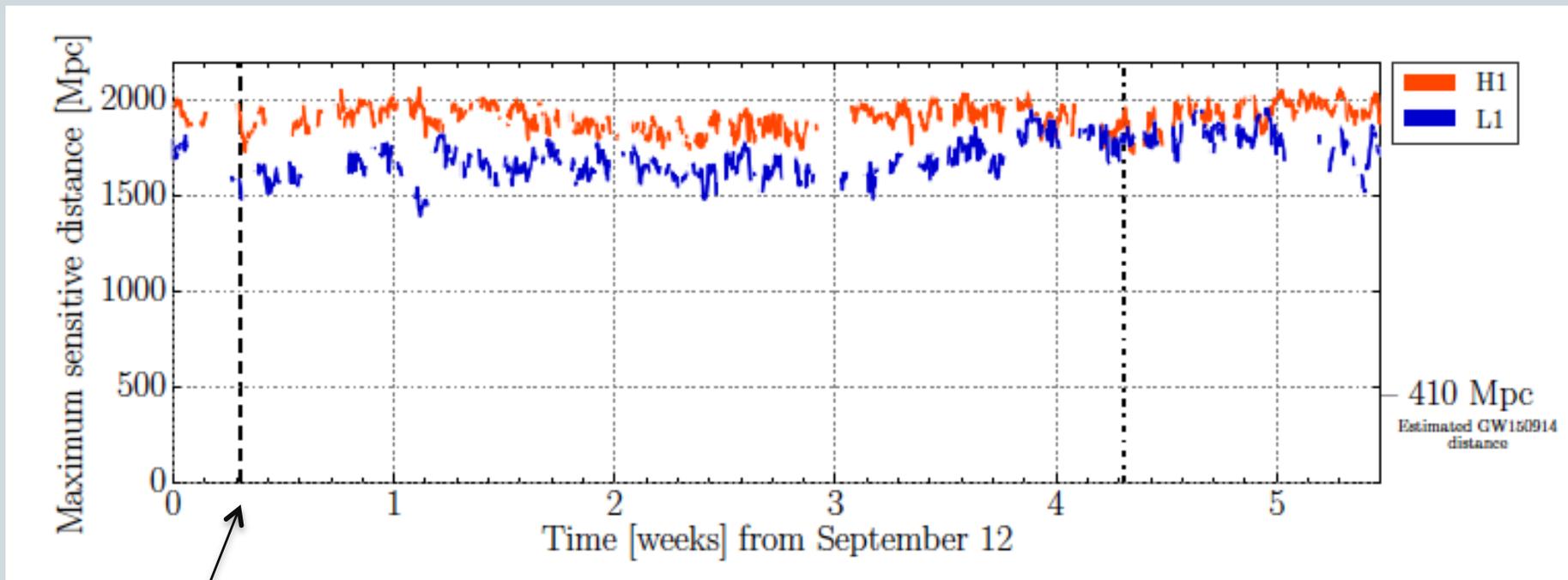
# GW150914 as seen by the generic short-transient signals search



# Intermediate analysis products



# State of the interferometers: stability



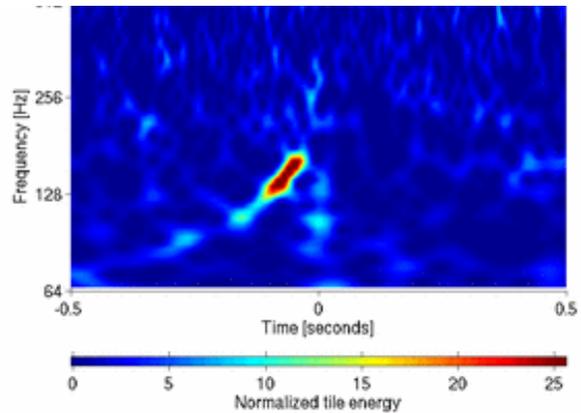
GW150914

# Candidate appearance, examples from LIGO science runs

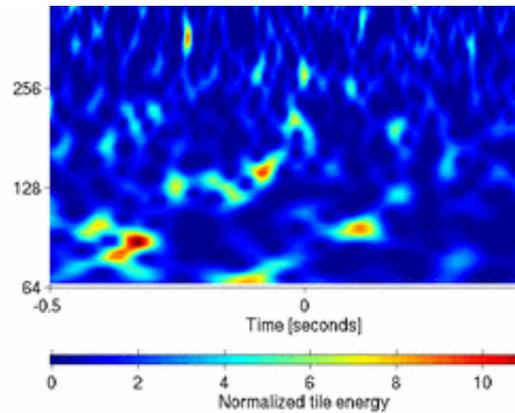


## Trigger from simulated signal

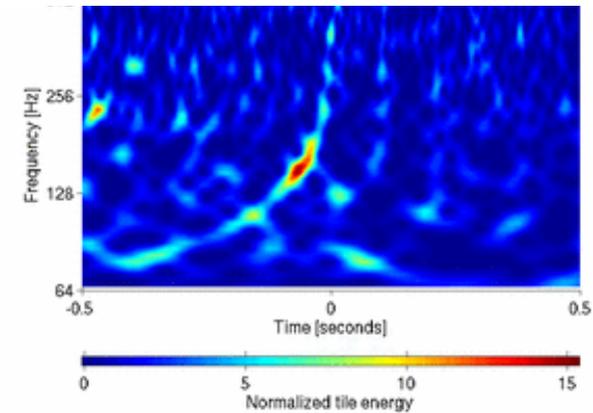
**GW channel: H1**



**GW channel: H2**

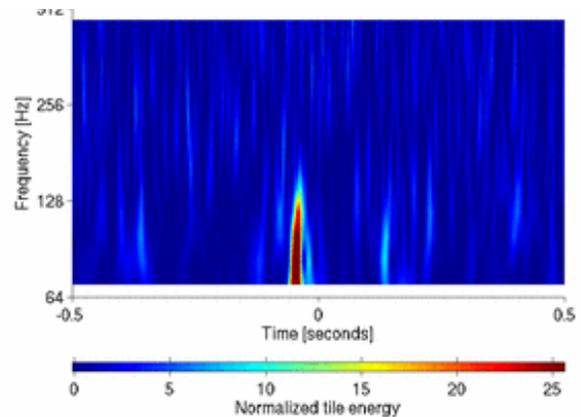


**GW channel: L1**

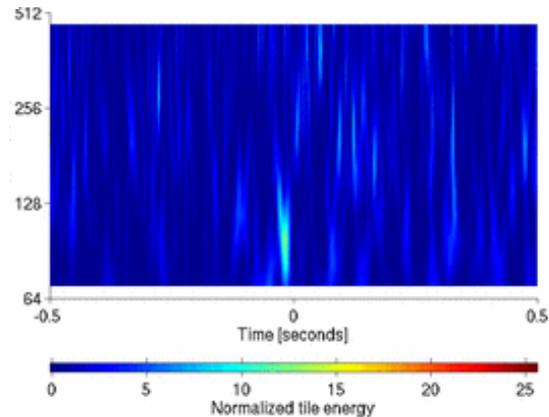


## Background trigger

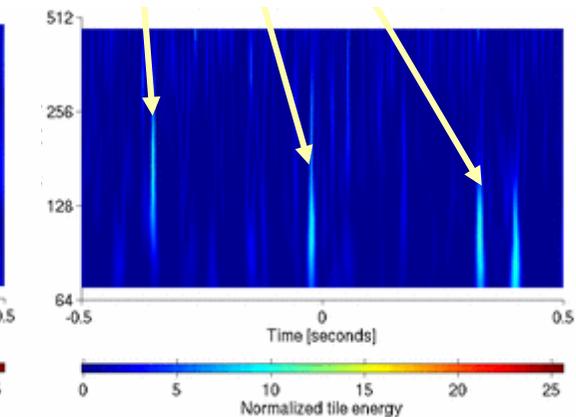
**GW channel: H1**



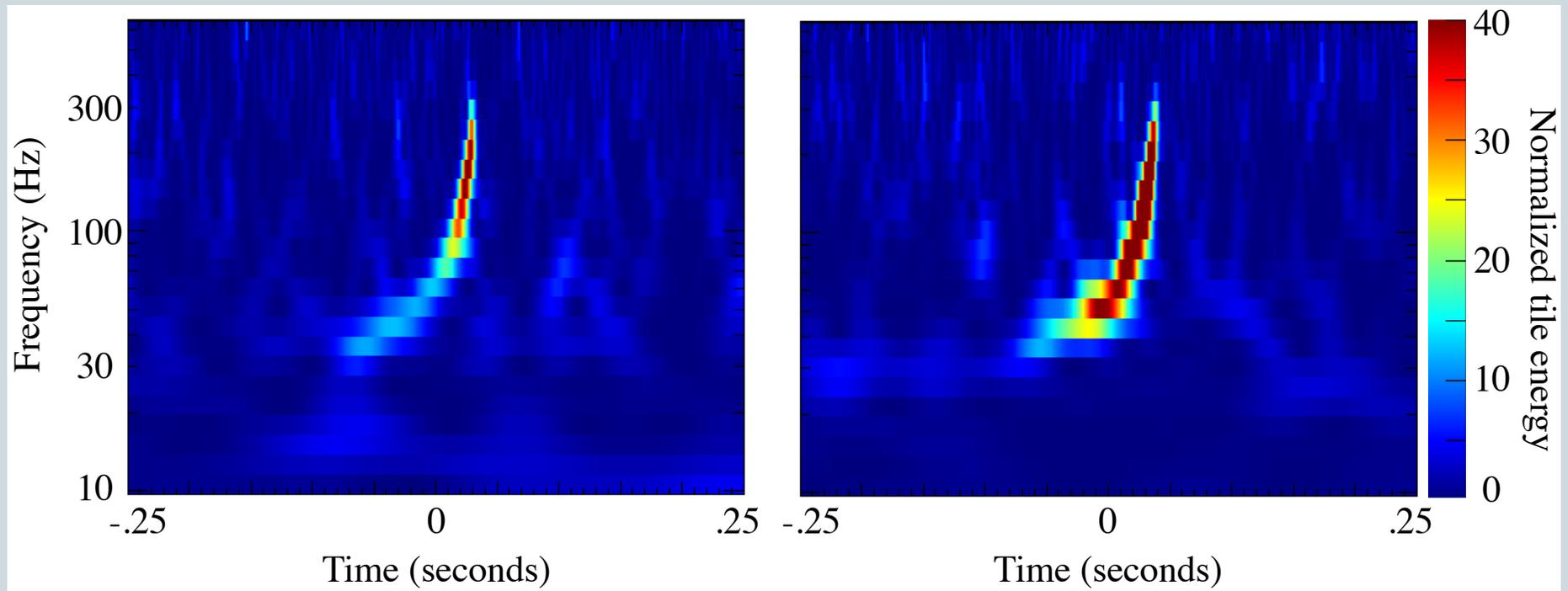
**GW channel: H2**



**GW channel: L1**



# GW150914 candidate appearance



# GW150914, the first GW detection



Washington, 12 February 2016

