# Search for gravitational waves from black hole hyperbolic encounters in LIGO-Virgo

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Motivation				

- Some models predict dense black hole clusters in our universe
  - Primordial black hole models
  - Models for the centers of galaxies and globular clusters
- Black holes in dense clusters will scatter off each other in hyperbolic orbits.
- Sufficiently close scatters will emit gravitational waves (GWs) detectable by current interferometers on Earth.
- Before this work, no targeted search for close hyperbolic encounters (CHEs) had been carried in GW detectors.

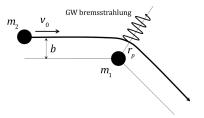


Figure 1: Schematic representation of a Hyperbolic Encounter.

# Steps to search for CHE in GW detectors

- Theoretical templates of the GWs emitted by CHE.
- Characterize the signal GWs from CHE leave in laser interferometers.
- Data processing to make the CHE signal stand out over detector noise.
- Develop a trigger to select the data that might contain a CHE. Our trigger will consist on:
  - Oata preselection based on correlations between interferometers.
  - Onvolutional Neural Network to determine which correlations can come from CHE.

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CHE tem	plates			

- Problem: Scattering of two gravitationally interacting masses  $m_1$  and  $m_2$  with spins  $\vec{S_1}$  and  $\vec{S_2}$ .
- No analytical solution in General Relativity (GR).
- Numerical Relativity is computationally cost prohibitive.
- We can use the Post Newtonian (PN) approximation because:
  - BHs do not get as close as in CBC (there is no merger).
  - Accurately following the phase for many cycles is not as critical.
- We take up to leading order spin effects  $ightarrow O(1/c^3) 
  ightarrow 1.5$ PN.

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# GW polarizations

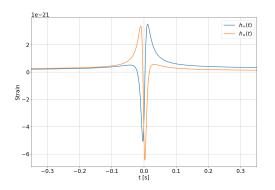


Figure 2: Representative gravitational waves polarization emitted by maximally spinning black holes with  $m_1 = 20M_{\odot}$ ,  $m_2 = 15M_{\odot}$ ,  $b = 70GM/c^2$ ,  $e_{t0} = 1.1$ ,  $\Phi_0 = 0$ ,  $\theta_1^i = 0.5$  rad,  $\phi_1^i = 0.35$  rad,  $\theta_2^i = 0.8$  rad,  $\phi_2^i = 1$  rad. t = 0 represents the time of closest approach.

- Quadrupolar nature of the GWs  $\rightarrow$   $f_{GW} = 2 f_{orbit}$ .
- $\bullet$  CHE perform "half" of an orbit  $\rightarrow$  GWs from CHE perform one oscillation.

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# Projected GWs into the detector

• The signal the GWs leave in the interferometers is simulated by projecting them into each detector and taking into account light travel time delays.

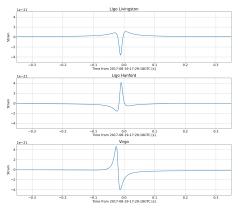


Figure 3: Result of projecting the GWs of Fig. 2 into the GW detectors, assuming that they come from  $\delta = 1.0$ rad,  $\alpha = 3.7$ rad, with  $\psi = 0.2$ rad and with the periastron time taking place at 17:29:18 UTC of 2017-08-19 at the center of the Earth.

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# Data Processing

#### • Filtering:

- $\bullet~$  Below 20Hz  $\rightarrow$  large increase in seismic and thermal noise.
- $\bullet\,$  Above 800Hz  $\rightarrow$  small amount of CHE signal.
- Remove noise dominated frequencies applying 20-800Hz band-pass.
- Whitening
  - Whiten to weigh down the strain in the noisiest regions and make all frequencies have same noise.
- Q transform:
  - GWs from CHEs only perform only one oscillation  $\rightarrow$  extended time-frequency structure.

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#### Data processing example

- Inject example CHE in the detector output:  $s(t) = s_{exp}(t) + h_{CHE}(t)$
- Filter, whiten and Q transform the data containing the injection.

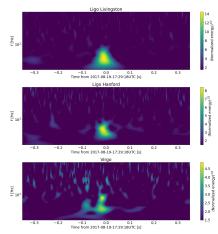


Figure 4: Square root of the normalized energy obtained by applying the data processing to the example CHE injected in the detector data.

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# Search for CHEs in LIGO-Virgo data

- CHEs look very similar to common blip glitches.
- Coincidences are very important to reject these glitches.
- More detectors online improve search sensitivity.
- Analyze public LIGO-Virgo O2 data with all three detectors in nominal operation  $\rightarrow$  15.3 effective days.
- Since CHEs only do one oscillation burst methods are competitive with and can outperform match filtering techniques for the search.
- Use a two step trigger to look for CHEs:
  - Correlation trigger.
  - Output Network.

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Correlatio	n trigger				

- We want a loose trigger to preselect events for the Neural Network.
- Look for correlations in the time-frequency domain with Pearson's coeficient:

$$r[x,y] = \frac{\text{cov}(x,y)}{\sigma_x \sigma_y} = \frac{\sum_{n=0}^{N-1} (x_n - \bar{x})(y_n - \bar{y})}{\sqrt{\sum_{n=0}^{N-1} (x_n - \bar{x})^2} \sqrt{\sum_{n=0}^{N-1} (y_n - \bar{y})^2}},$$

- Trigger discriminant:  $D = ar_{L1-H1} + (1-a)r_{L1-V1}$
- In O2 we determine that for CHE signals, optimally  $a\sim 0.67$
- Preselect images with D > 0.3



- What fraction of CHEs can we detect as a function of their SNR?
- We inject CHEs with random parameters and run the correlation trigger over them.

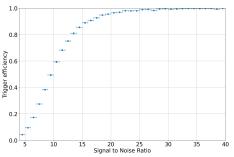


Figure 5: Trigger efficiency as a function of the signal to noise ratio of the injected events.

• We are accepting most events with SNR > 10.

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Trigger False /	Alarm Rate			

- We estimate the number of expected noise events by running the trigger on shifted data.
- As expected, events with larger discriminant are increasingly uncommon.

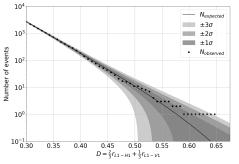


Figure 6: Number of events above a certain discriminant observed and expected.

$$\sigma = \sqrt{N_{expected}}.$$

• Total preselected (D > 0.3) events in the data: 2704

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## Neural Network

- We want to classify which of the 2704 preselected images come from noise and which from CHE.
- Neural networks (NN) are mathematical models to analyze data using Machine Learning.

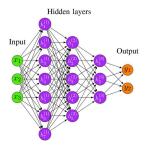


Figure 7: NN sketch. [F. Emmert-Streib (2020)]

- NN input: pixels of the image to be classified.
- NN output: probability of the image to be a CHE.
- Hidden layers:
  - Used for internal calculations only.
  - Contain free parameters to fit training data.
- We use one of the most refined architectures for image classification: Residual Convolutional Neural Network

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### Image samples

- Put the normalized energy of the three detectors in the same image.
- Image duration: 0.7s



(a) CHE image.

(b) Noise image.



(c) Data Image.

Figure 8: Examples of images used to train and validate our neural network as well as to make predictions on the data.

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#### Neural Network training and performance

- Training the NN means finding the internal parameters such that:
  - $\bullet~$  If input is a CHE image  $\rightarrow~$  return 1 ~
  - If input is a noise image  $\rightarrow$  return 0
- NN output is then interpreted as CHE probability.
- Train the NN on 64028 noise images and 45356 CHE images.
- We consider an image a CHE candidate if  $p_{CHE} > 0.9$

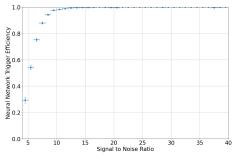


Figure 9: NN trigger efficiency as a function of the signal to noise ratio of the CHE test events.

• We are able to detect most CHEs with signal to noise ratio above 5

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## Neural Network results on Data

• Distribution of data with CHE probability can be compared with expectation if only noise was present.

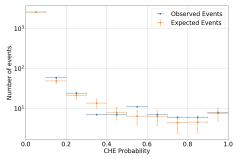


Figure 10: Number of observed events and expected noise events as a function of CHE probability given by the NN. The number of expected noise events is the number of events we would expect to see if the data only contained noise.  $\Delta N_{\text{expected}} = \sqrt{N_{\text{expected}}}$ 

- Consistently more events than expected at high CHE probability.
- At  $p_{CHE} > 0.9 \rightarrow$  We expect 7.5  $\pm$  2.7 noise events and we observe 8.

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## Most significant events

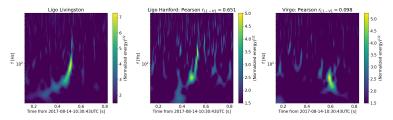


Figure 11: Event with  $p_{CHE} = 0.997$  and correlation discriminant D = 0.469. This event is GW170814, the coalescence of two black holes of masses  $31M_{\odot}$  and  $25M_{\odot}$ .

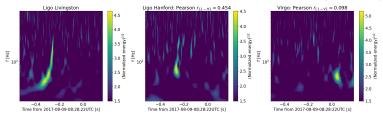


Figure 12: Event with  $p_{CHE} = 0.991$  and correlation discriminant D = 0.337. This event is GW170809, the coalescence of two black holes of masses  $35M_{\odot}$  and  $24M_{\odot}$ .

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#### Possible CHEs

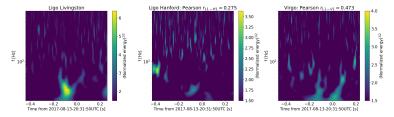


Figure 13: Event with  $p_{CHE} = 0.980$  and correlation discriminant D = 0.341.

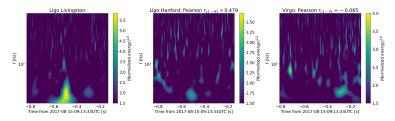


Figure 14: Event with  $p_{CHE} = 0.976$  and correlation discriminant D = 0.321.

- Determined GWs emitted by CHEs and characterized how they look in real GW detectors.
- Looked for CHEs in LIGO-Virgo data using burst methods and machine learning.
- Found 2 of the 3 binary black hole mergers contained in the data analyzed.
- Obtained 6 candidate CHEs on which to perform further analysis.
- The number of events is consistent with just noise.

# **Backup Slides**

Backup Slides

# Hamiltonian Formulation of the problem

• Hamiltonian of the system:

$$H(\vec{r}, \vec{p}, \vec{S}_1, \vec{S}_2) = H_{\mathrm{N}}(\vec{r}, \vec{p}) + H_{\mathrm{1PN}}(\vec{r}, \vec{p}) + H_{\mathrm{SO}}(\vec{r}, \vec{p}, \vec{S}_1, \vec{S}_2) + O\left(rac{1}{c^4}
ight) \,,$$

where

$$\begin{split} & \mathcal{H}_{\mathrm{N}}(\vec{r},\vec{p}) = \frac{p^{2}}{2} - \frac{1}{r} \,, \\ & \mathcal{H}_{\mathrm{1PN}}(\vec{r},\vec{p}) = \frac{1}{c^{2}} \left( \frac{1}{8} (3\eta - 1)(p^{2})^{2} - \frac{1}{2} \left[ (3 + \eta)p^{2} + \eta(\hat{n} \cdot \vec{p})^{2} \right] \frac{1}{r} + \frac{1}{2r^{2}} \right) \,, \\ & \mathcal{H}_{\mathrm{SO}}(\vec{r},\vec{p},\vec{S}_{1},\vec{S}_{2}) = \frac{1}{c^{2}r^{3}} (\vec{r} \times \vec{p}) \cdot \vec{S}_{\mathrm{eff}} \,, \end{split}$$

• To get the equations of motion we use Poisson's brackets:

$$\{r_i, p_j\} = \delta_{ij},$$
  
 $\{S_{1i}, S_{1j}\} = \epsilon_{ijk}S_{1k},$   
 $\{S_{2i}, S_{2j}\} = \epsilon_{ijk}S_{2k}.$ 

# Solution for the orbit

- After some manipulation, the equations of motion can be efficiently solved to obtain the orbit.
- The orbit will depend on:
  - Black hole masses m1, m2
  - Black hole initial spins  $\vec{S}_1$ ,  $\vec{S}_2$
  - Initial eccentricity et0
  - Impact parameter b
  - Initial orbital azimutal angle  $\Phi_0$
  - Orbital inclination angle  $\Theta$

$$b = 70 \ GM/c^2$$
  $v_{max} = 0.36 \ c$ 

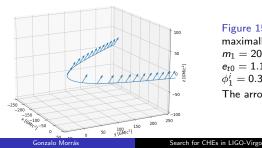


Figure 15: Example of an orbit for maximally spinning black holes with  $m_1 = 20M_\odot$ ,  $m_2 = 15M_\odot$ ,  $b = 70Gm/c^2$ ,  $e_{t0} = 1.1$ ,  $\Phi_0 = 0$ ,  $\theta_1^i = 0.5$  rad,  $\phi_1^i = 0.35$  rad,  $\theta_2^i = 0.8$  rad,  $\phi_2^i = 1$  rad. The arrow represents  $\vec{S}_{\rm eff}$ .

# GWs derived from the orbit

- GWs can computed from the orbit.
- Use formula with up to leading order spin effects:

$$\begin{split} h_{\times} &= 4 \frac{Gm\eta}{c^{4}R'} \Big[ -(p\cdot n)(q\cdot n)z + (p\cdot v)(q\cdot v) - \frac{\delta}{c} [ \langle [[3(N\cdot n)\dot{r} - (N\cdot v)](q\cdot n) - 3(N\cdot n)(q\cdot v)](p\cdot n) - 3(N\cdot n)(q\cdot v)(p\cdot v))z \\ &+ 2(p\cdot v)(q\cdot v)(N\cdot v)] + \frac{1}{6c^{2}} [ (5(1-3\eta)(N\cdot v)^{2}(p\cdot v)(q\cdot v) + ([(6\eta-2)(N\cdot v)^{2}(q\cdot n) + (48\eta-16)(N\cdot v)(N\cdot n)(q\cdot v)])z \\ &+ (2p\cdot v)(q\cdot v)(N\cdot v)] + \frac{1}{6c^{2}} [ (5(1-3\eta)(N\cdot v)^{2}(p\cdot v)(q\cdot v) + ([(6\eta-2)(N\cdot v)^{2}(q\cdot n) + (48\eta-16)(N\cdot v)(N\cdot n)(q\cdot n)] \\ &\times (p\cdot n) + (48\eta-16)(N\cdot v)(N\cdot n)(p\cdot v)(q\cdot n) + ((-14+42\eta)(N\cdot n)^{2}-4+6\eta)(q\cdot v)(p\cdot v))z + (-9\eta+3)(q\cdot v) \\ &\times (p\cdot v)v^{2} + (29+(7-21\eta)(N\cdot n)^{2}] (q\cdot n)(p\cdot n)z^{2} + ((-9\eta+3)(N\cdot n)^{2}-10-3\eta)(q\cdot n)(p\cdot n)z^{2} + ([(-36\eta+12) \\ &\times (N\cdot v)(N\cdot n)(q\cdot n) + ((15-45\eta)(N\cdot n)^{2}+10+6\eta)(q\cdot v)](p\cdot n) + [(15-45\eta)(N\cdot n)^{2}+10+6\eta](p\cdot v)(q\cdot n))\dot{r}z \\ &+ ((45\eta-15)(N\cdot n)^{2}-9\eta+3)(q\cdot n)(p\cdot n)^{2}z] + \frac{c^{2}(q\cdot n)}{2} [ 2\chi_{2\chi}2(p\cdot (s_{2}\times N)) - X_{1\chi_{1}}(p\cdot (s_{1}\times N))] \Big], \\ h_{+} &= 2 \frac{Gm\eta}{c^{4}R'} \Big[ [((q\cdot n)^{2}-(p\cdot n)^{2})z + (p\cdot v)^{2}-(q\cdot v)^{2}] - \frac{\delta}{2} [((N\cdot n)\dot{r}-(N\cdot v))z(p\cdot n)^{2}-6z(N\cdot n)(p\cdot n)(p\cdot v) + (-3(N\cdot n)\dot{r})(N\cdot v)(x) + (16-4\gamma)(N\cdot v)^{2}(q\cdot n)^{2}+6z(N\cdot n)(q\cdot n)(q\cdot v) + 2((p\cdot v)^{2}-(q\cdot v)^{2})(N\cdot v) \Big] + \frac{1}{6c^{2}} [6(N\cdot v)^{2}((p\cdot v)^{2}-(q\cdot v)^{2})(1-3\eta) \\ &+ ((5\eta-2)(N\cdot v)^{2}(p\cdot n)^{2}+(6\eta-32)(N\cdot v)(N\cdot n)(p\cdot v)(p\cdot n) + (-6\eta+2)(N\cdot v)^{2}(q\cdot n)^{2}+(-9\eta+3)(N\cdot v) \\ &\times (N\cdot n)(q\cdot v)(q\cdot n) + [(-14+42\eta)(N\cdot n)^{2}-4+6\eta](p\cdot v)^{2}+[(-42\eta+14)(N\cdot n)^{2}+4-6\eta](q\cdot v)^{2}z) \\ &+ (((-9\eta+3)(p\cdot v)^{2}+(-3+9\eta)(q\cdot v)^{2})v^{2}+((-23\eta+\eta)(N\cdot n)^{2}+1-29+(21\eta-7)(N\cdot n)^{2}](q\cdot n)^{2}z)^{2} \\ &+ (((-9\eta+3)(p\cdot v)^{2}+(-3+9\eta)(q\cdot v)^{2}v^{2}+((-3\eta+\eta)(N\cdot n)^{2}+1-29+(21\eta-7)(N\cdot n)^{2}](q\cdot n)^{2}z^{2} \\ &+ ((-9\eta+3)(N\cdot n)^{2}-10+3\eta)(p\cdot n)^{2}+((-3+9\eta)(N\cdot n)^{2}+10+3\eta)(q\cdot n)^{2}z^{2} \\ &+ (((-9\eta+3)(N\cdot n)^{2}-10+3\eta)(N\cdot n)^{2}+10+3\eta)(q\cdot n)^{2}+10+3\eta)(q\cdot n)^{2}z^{2} \\ &+ ((-9\eta+3)(N\cdot n)^{2}-10+3\eta)(N\cdot n)^{2}+10+3\eta)(q\cdot n)^{2}+10+3\eta)(q\cdot n)^{2}z^{2} \\ &+ ((-9\eta+3)(N\cdot n)^{2}-10+3\eta)(N\cdot n)^{2}+10+3\eta)(q\cdot n)^{2}+1(-10+3\eta)(N\cdot n)^{2} \\ &+ ((-9\eta+3)(N\cdot n)^{2}-10+3\eta)(N\cdot n)^{2}+10+3\eta)(q\cdot n)^{2}+1(-10+3\eta)(N\cdot n)^{2} \\ &+ ((-9\eta+3)(N\cdot n)^$$

## Antenna Patterns

$$\begin{split} F_+(\theta,\phi,\psi) &= \frac{1}{2} (1 + \cos^2 \theta) \cos 2\phi \cos 2\psi - \cos \theta \sin 2\phi \sin 2\psi \,, \\ F_\times(\theta,\phi,\psi) &= \frac{1}{2} (1 + \cos^2 \theta) \cos 2\phi \sin 2\psi + \cos \theta \sin 2\phi \cos 2\psi \,. \end{split}$$

- $\theta$  and  $\phi$  give the direction of the GW in the frame in which the arms of the interferometer are in the x and y axis.
- $\psi$  measures the polarization of the wave.
- Detectors have different locations and orientations  $\rightarrow \theta$ ,  $\phi$  and  $\psi$  are different in each one for a GW coming from a location in the sky with right ascension  $\alpha$  and declination  $\delta$ .
- GWs arrive at different times in different locations due to the light travel time.

# Injected signal

- Current detectors are dominated by noise.
- Data processing suppresses the noise, enhancing the signal.

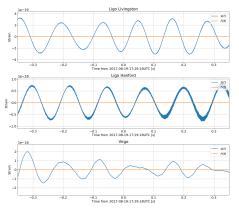


Figure 16: In blue, result s(t) of injecting the example GWs into the experimental strain of the detectors. In orange, the GW signal h(t). The noise is so dominating that the features of the gravitational wave signal can not be seen.

# CHE frequency dependance

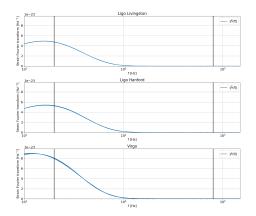


Figure 17: Fourier transform of the gravitational wave shown in Fig. 3. We have marked with black lines the 20-800Hz frequency window we are using to band-pass the strain.

# Filtering

- Below 20Hz and above 800Hz data is dominated by noise.
- Remove noise dominated frequencies applying 20-800Hz band-pass.

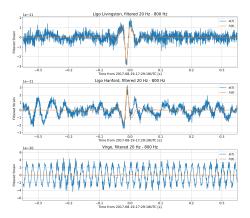


Figure 18: Result of band-pass filtering the example signals within a 20-800Hz window.

# Whitening

- The signal in frequencies with more noise is less significant.
- Whiten to make all frequencies have same noise  $\tilde{s}_{whiten}(f) = \frac{\tilde{s}(f)}{\sqrt{S_n(f)}}$ .

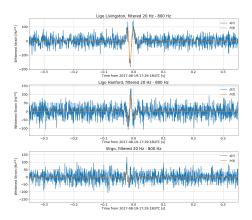


Figure 19: Result of whitening the example filtered signals.

## $3\sigma$ correlated event

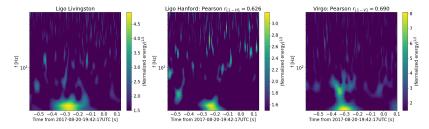


Figure 20: Event with the highest value of the trigger discriminant found in the analyzed data, with a value D = 0.647. From our false alarm rate analysis we expect an event like this every 240 days but we got one in 15.3 days of data.

# Other CHE candidates I

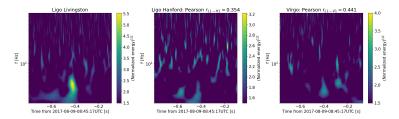


Figure 21: Event with  $p_{CHE} = 0.974$  and correlation discriminant D = 0.382.

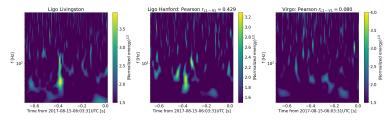


Figure 22: Event with  $p_{CHE} = 0.942$  and correlation discriminant D = 0.314.

# Other CHE candidates II

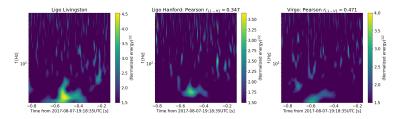


Figure 23: Event with  $p_{CHE} = 0.927$  and correlation discriminant D = 0.388.

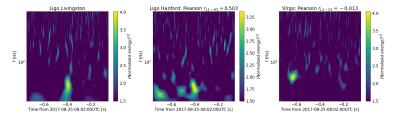


Figure 24: Event with  $p_{CHE} = 0.914$  and correlation discriminant D = 0.337.