

# Detection of Stochastic Gravitational Wave Background Based on Topological Data Analysis

M.Sc. Thesis defense

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Computational Cosmology Group  
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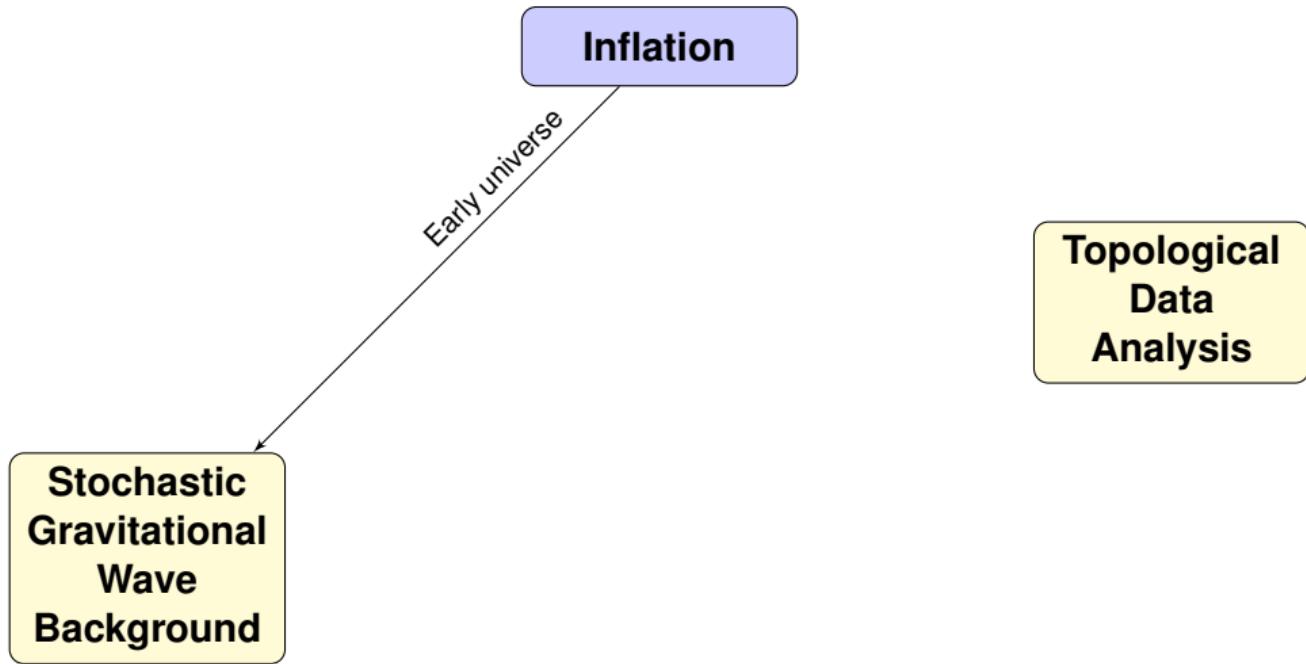


# Overview

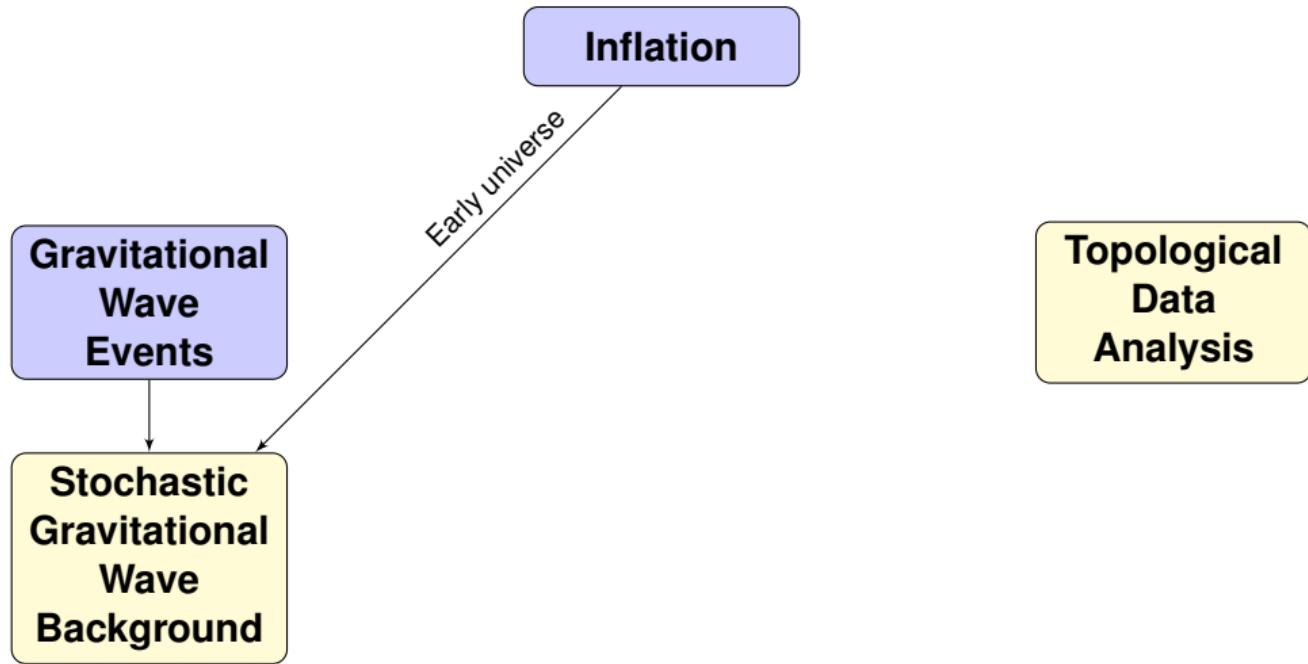
**Stochastic  
Gravitational  
Wave  
Background**

**Topological  
Data  
Analysis**

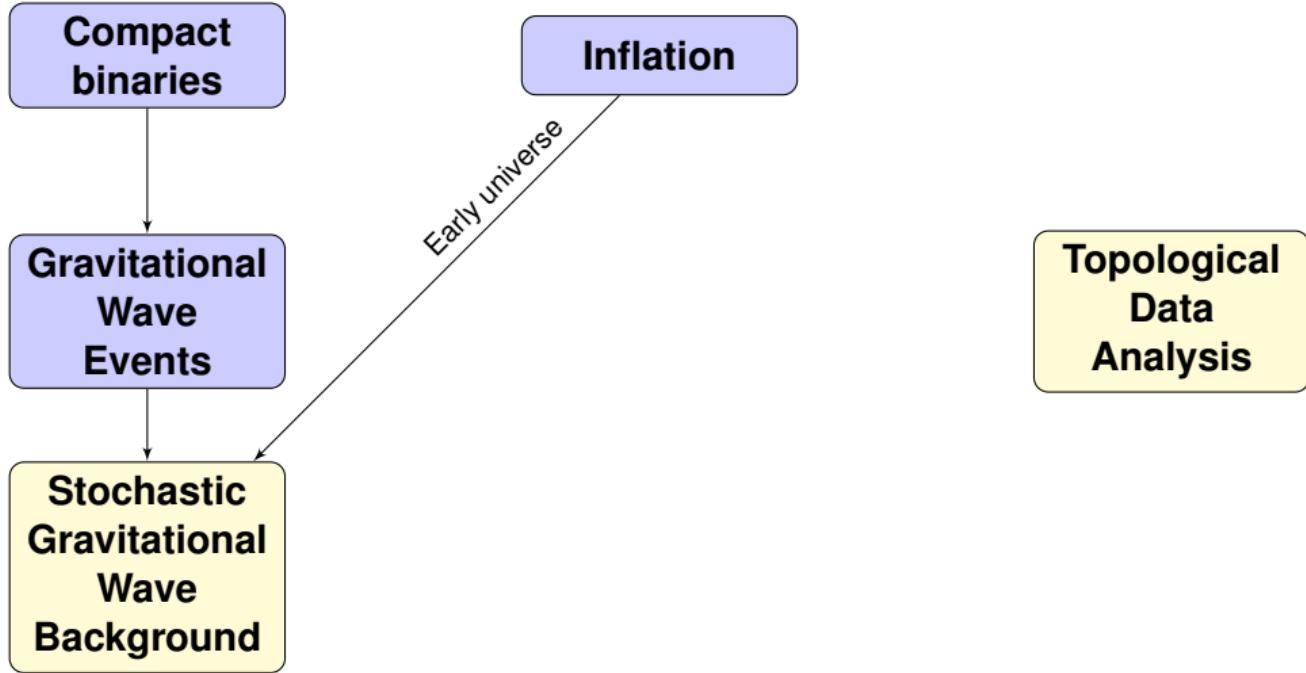
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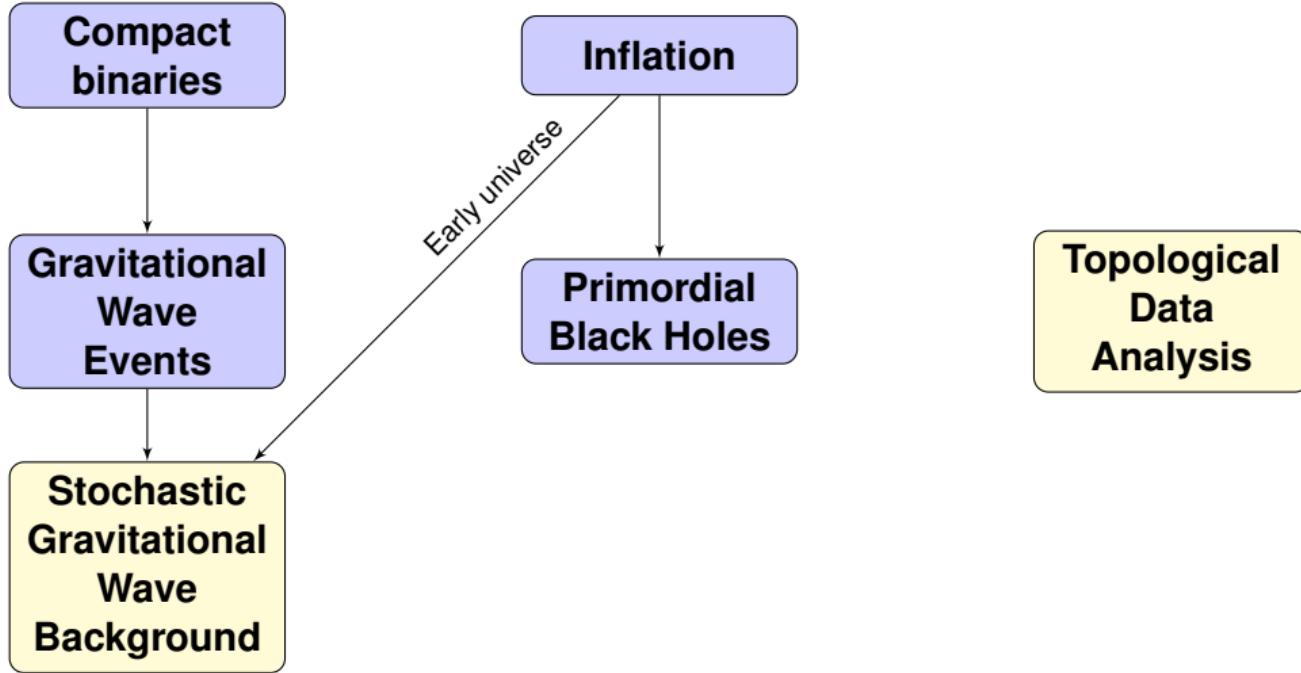
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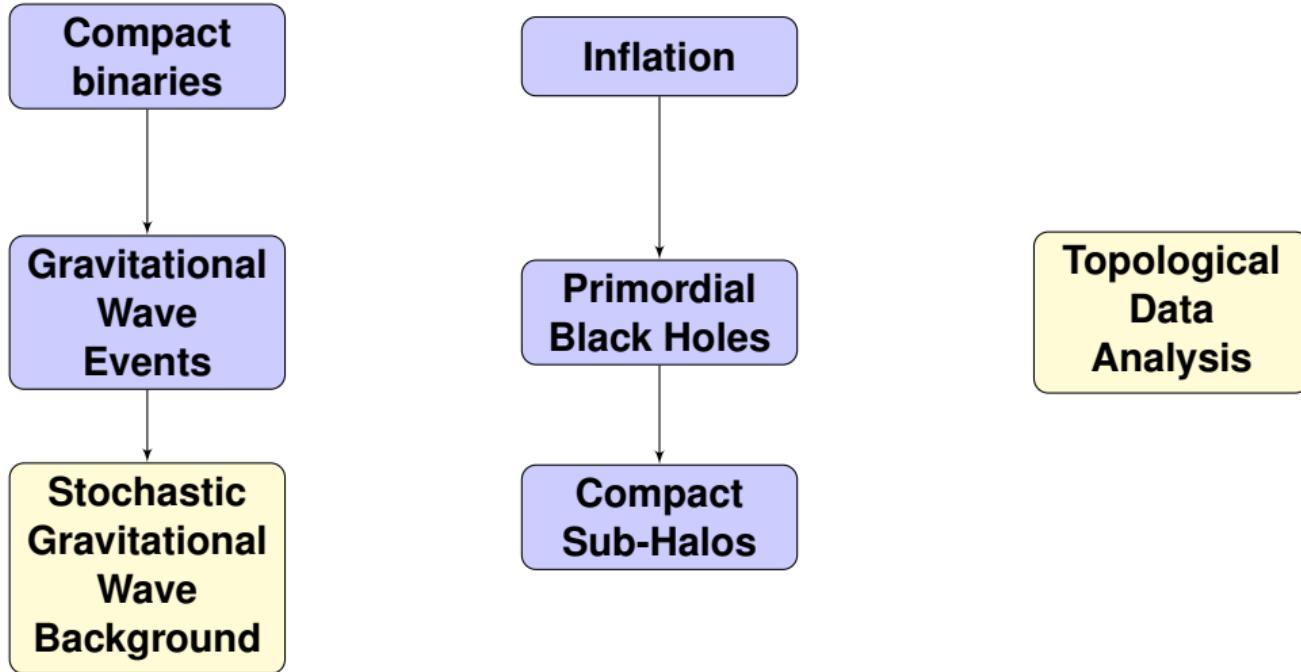
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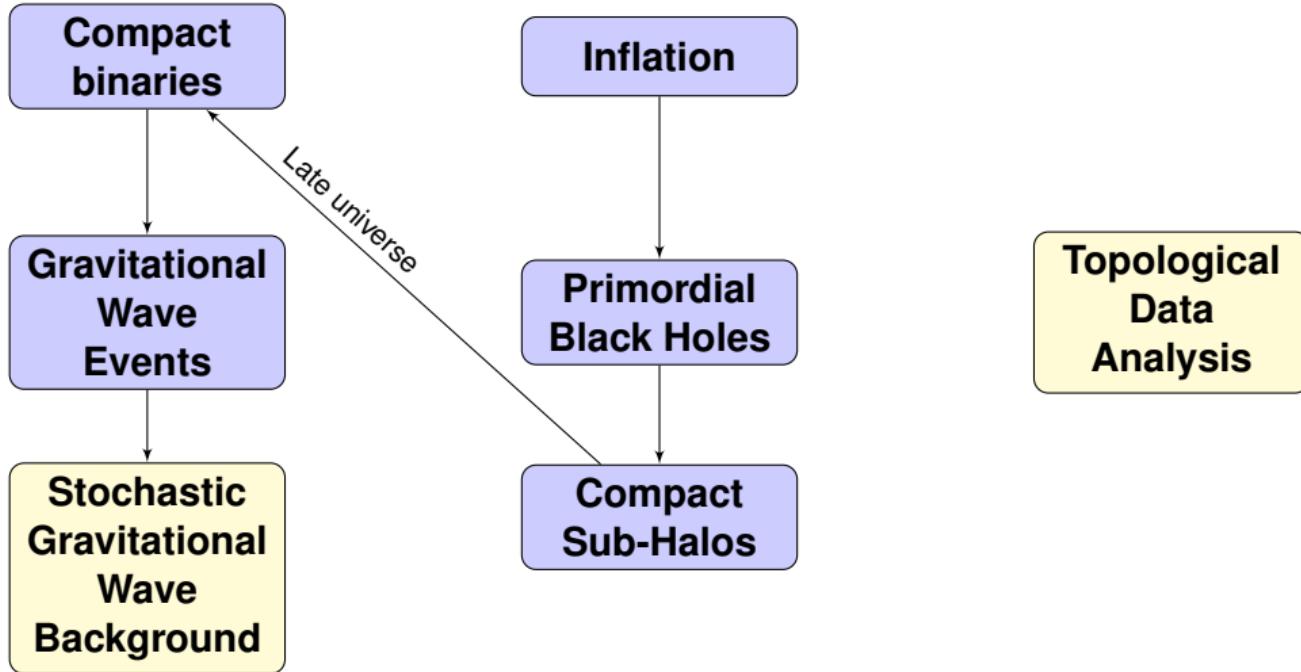
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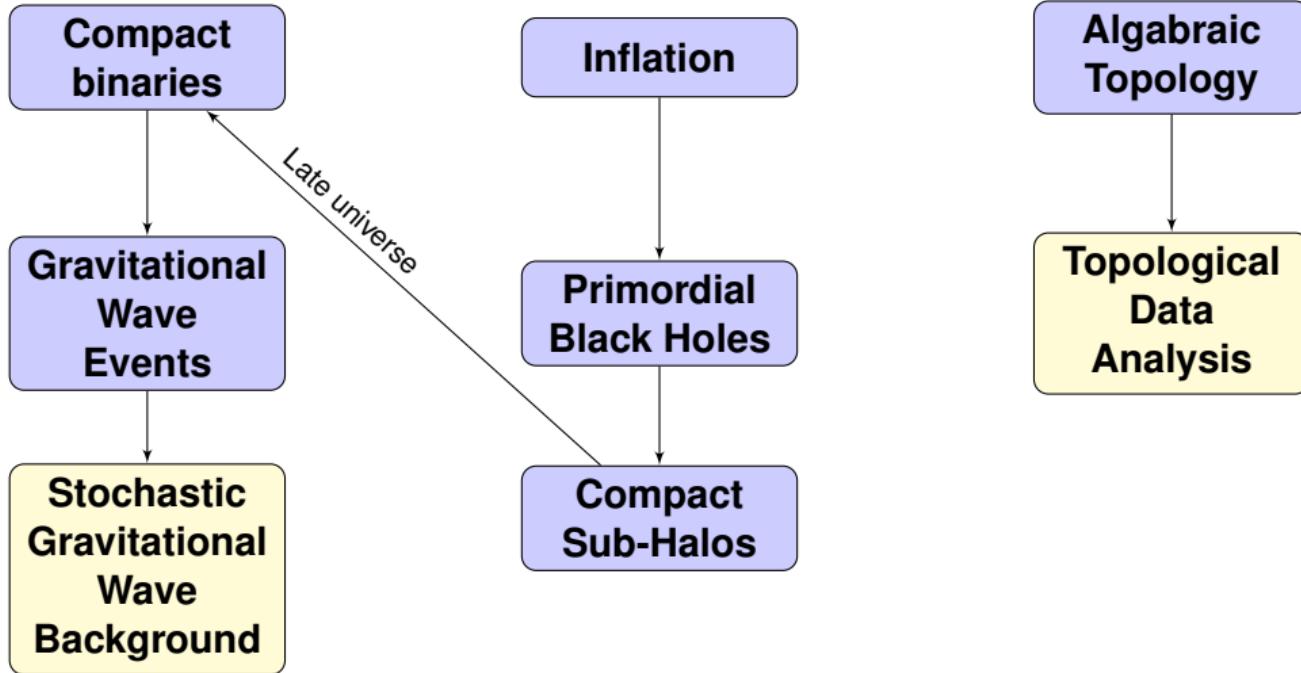
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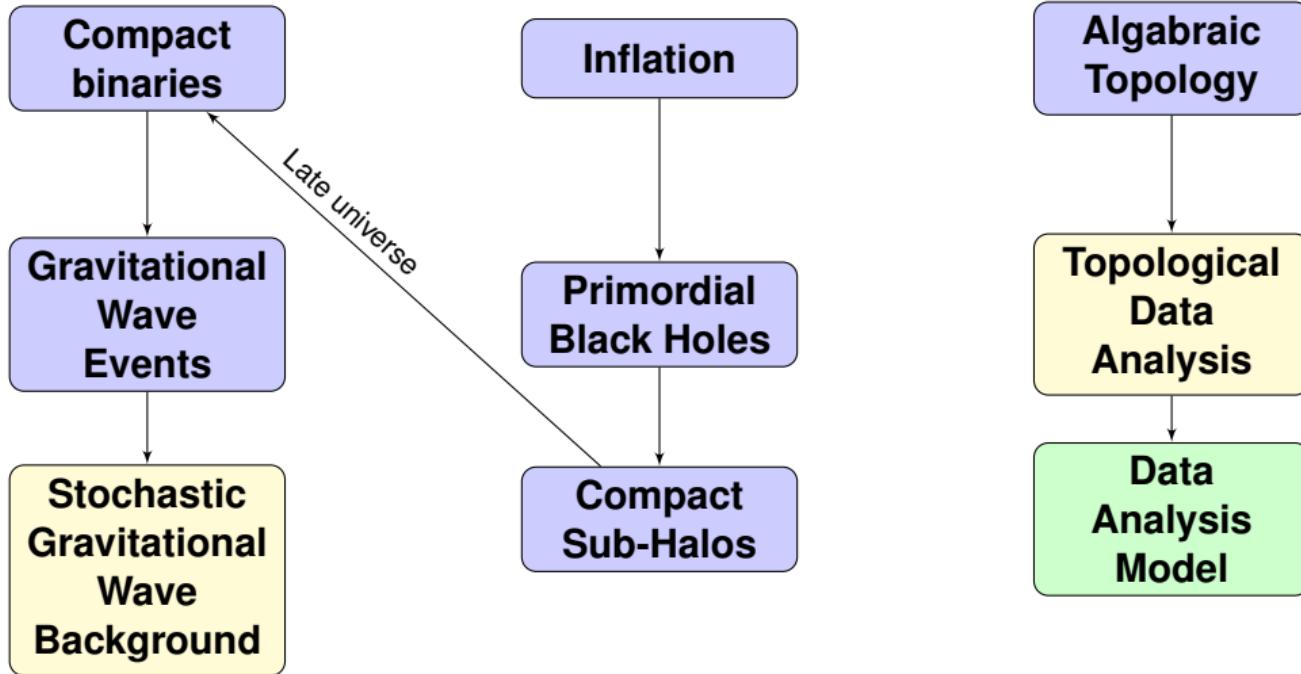
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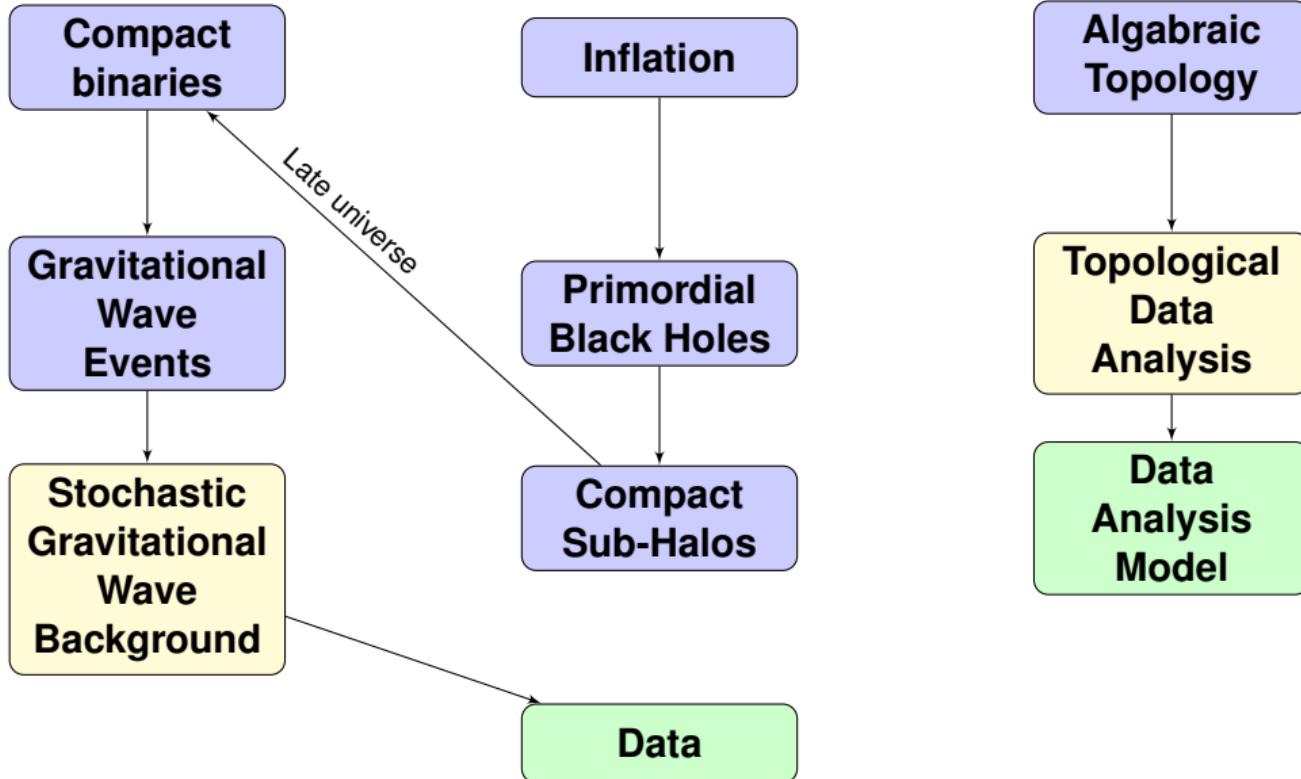
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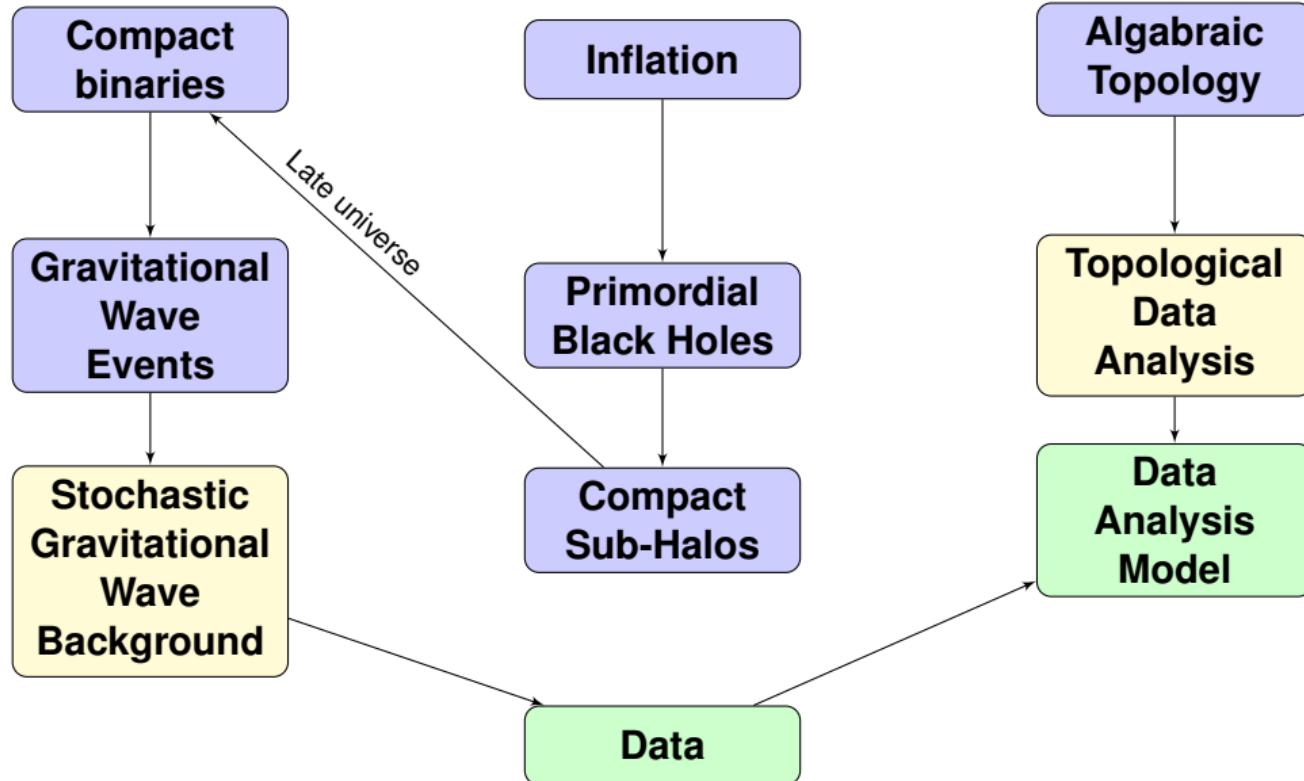
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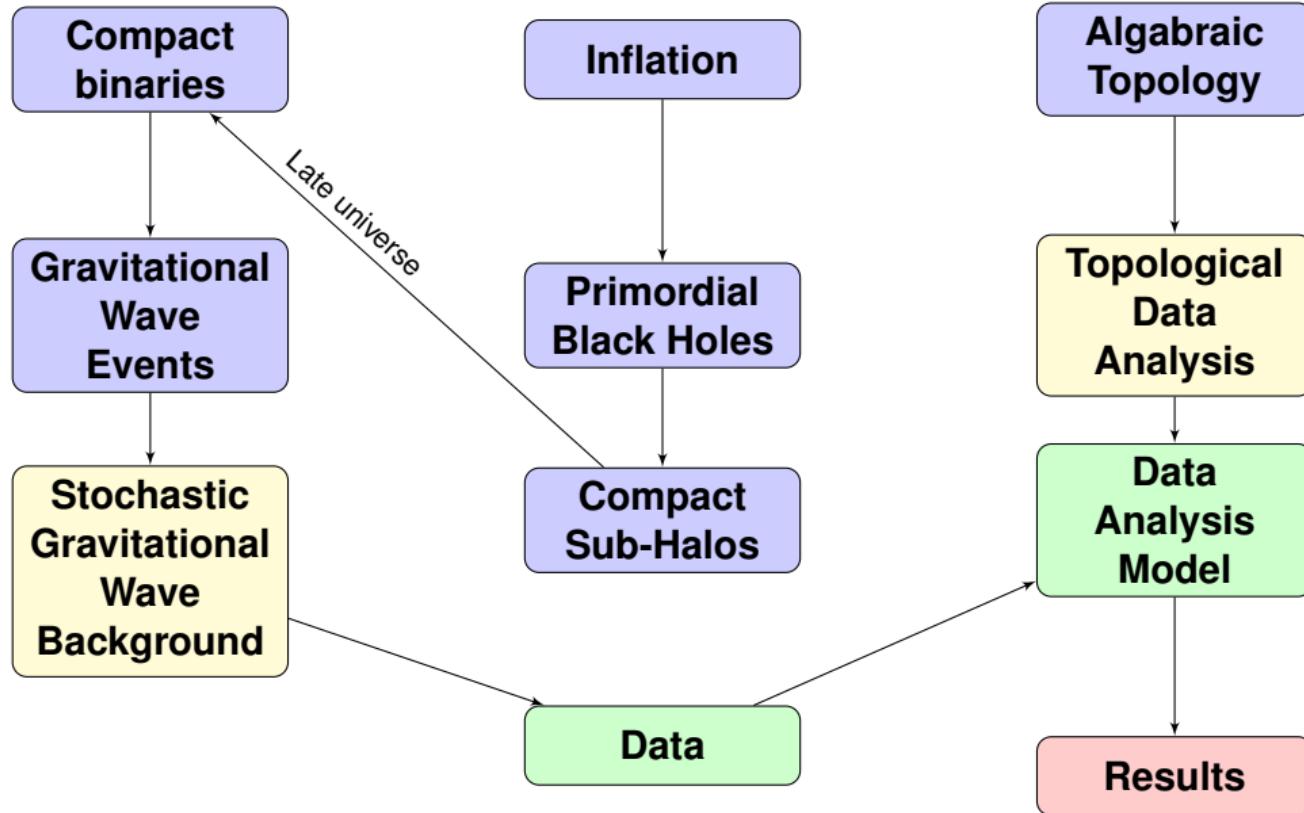
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# Outline

1 Gravitational Waves

2 Primordial Black Holes

3 Simulations

4 Topological Data Analysis

5 Results

6 Conclusion and Outlook

## Section 1

# Gravitational Waves

# Where do the gravitational waves come from?

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Image credit: [Youtube: minutephysics](#)

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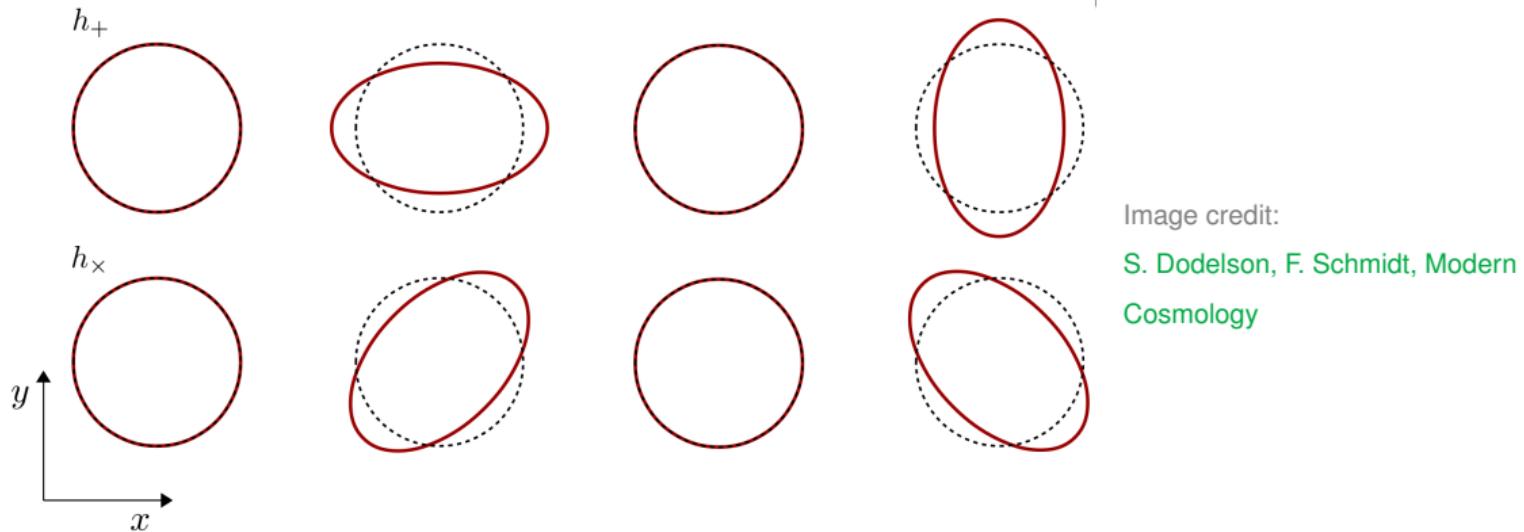
In Transverse-Traceless Gauge:

$$h_{\mu\nu} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & h_+ & h_\times & 0 \\ 0 & h_\times & -h_+ & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

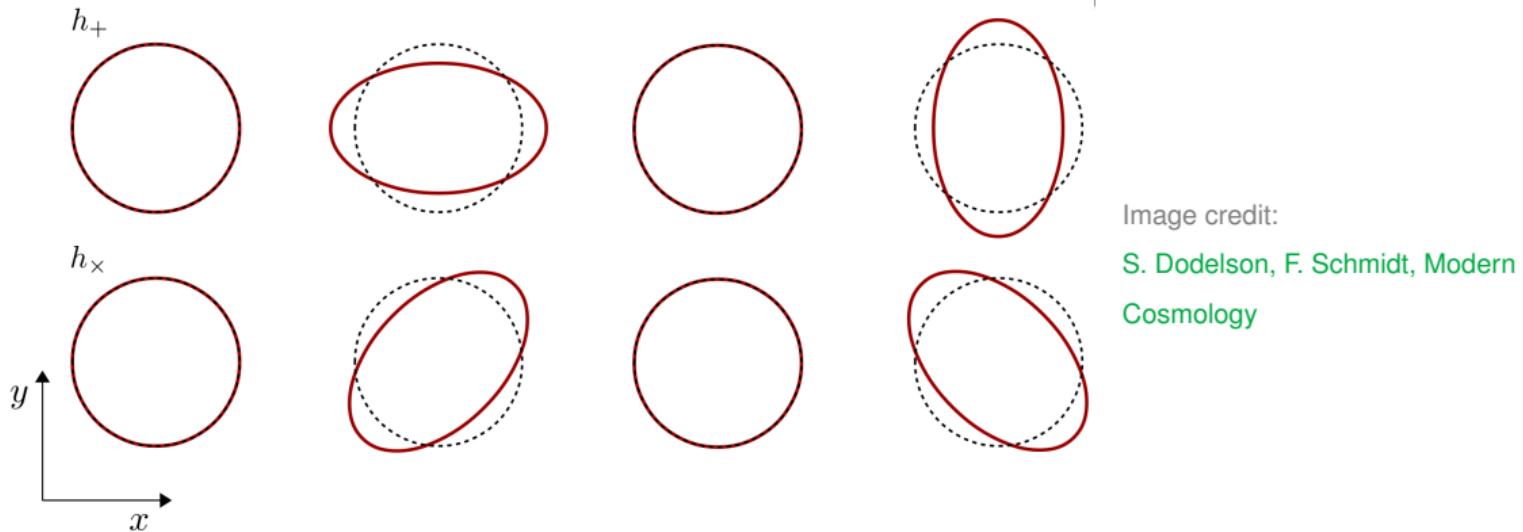


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# Polarizations are the signature of gravitational waves

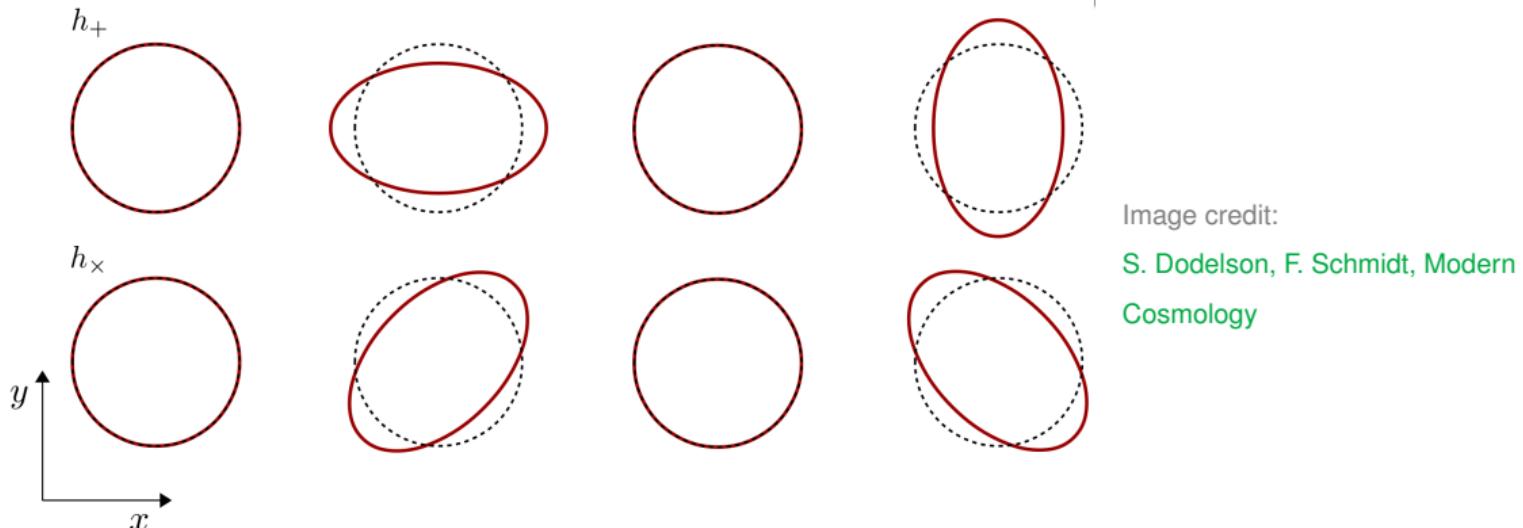


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Other Gravitational wave features:

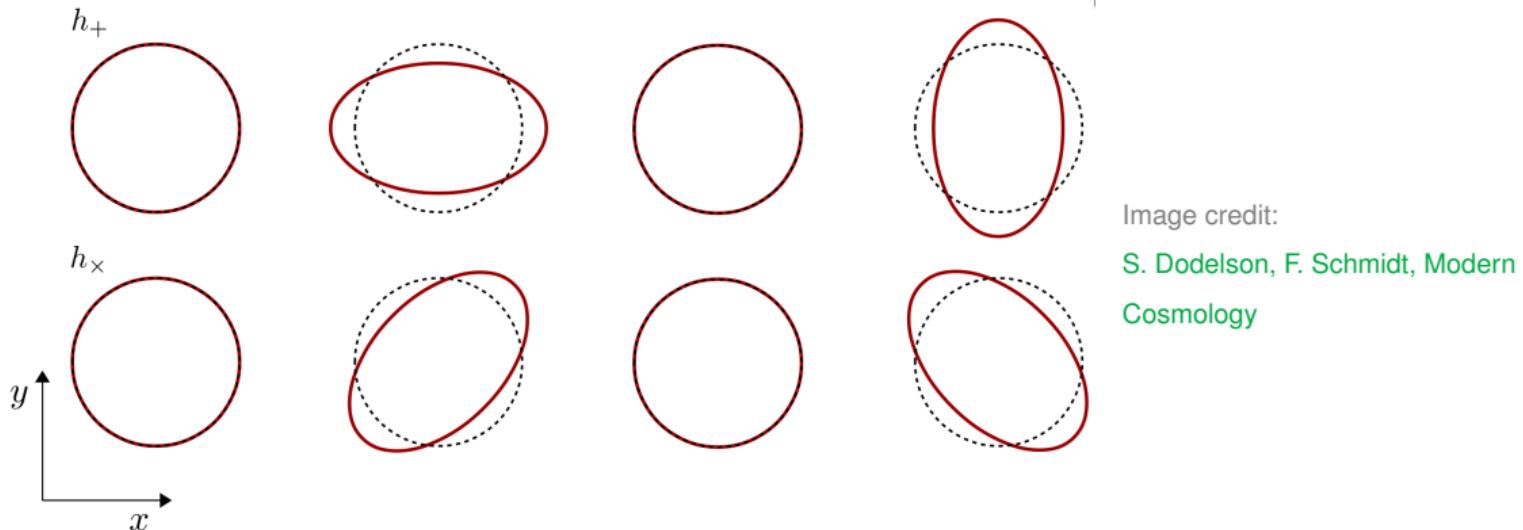
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Other Gravitational wave features:

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# Sources of Gravitational Waves

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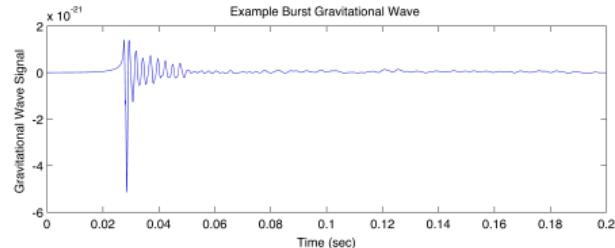


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Continuous and Inspiral:

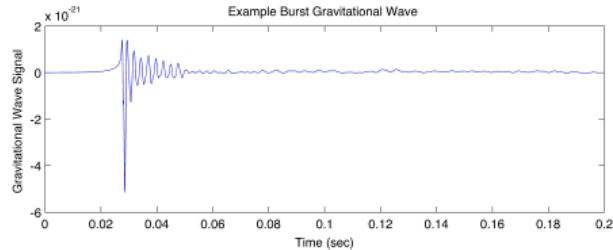
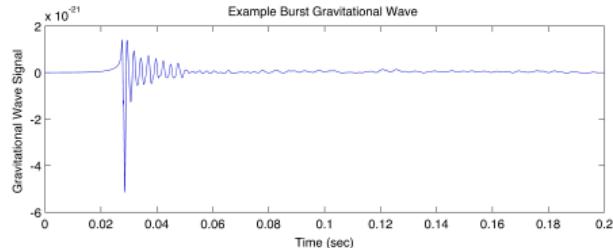


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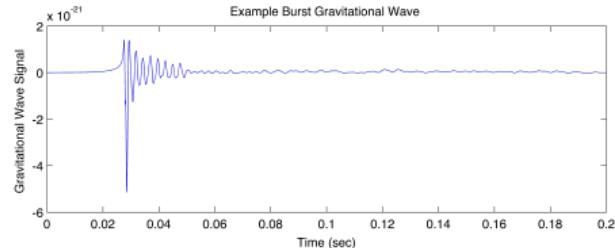
- ▶ Binary mergers (NS, SBH, PBH, SMBH, ...)

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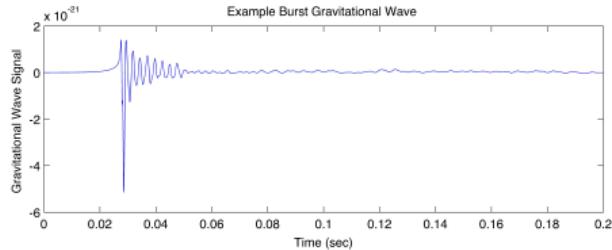
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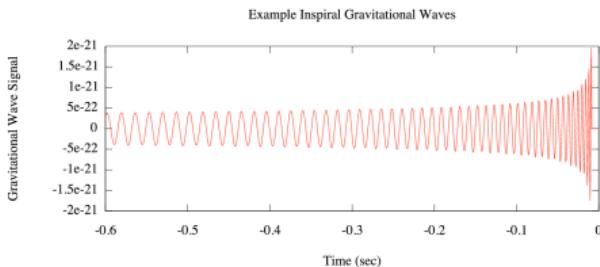
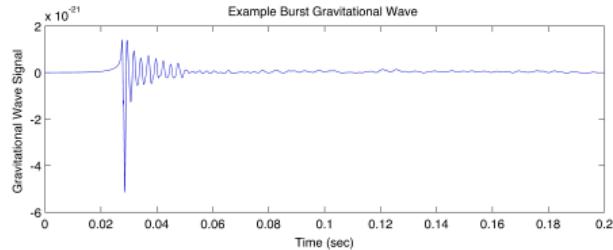


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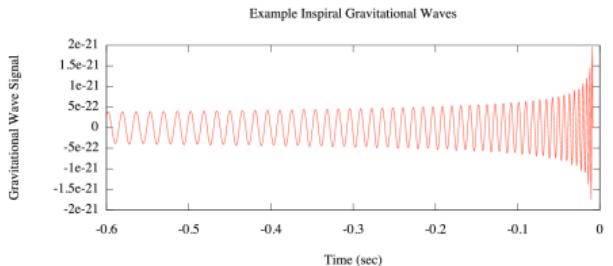
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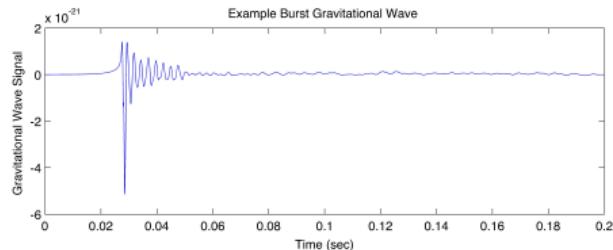
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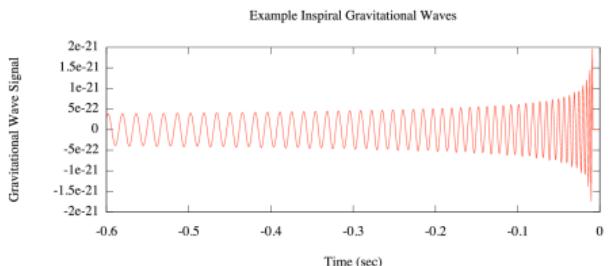
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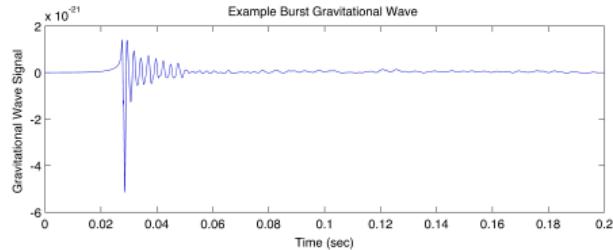
Stochastic:

- ▶ Primordial tensor perturbations

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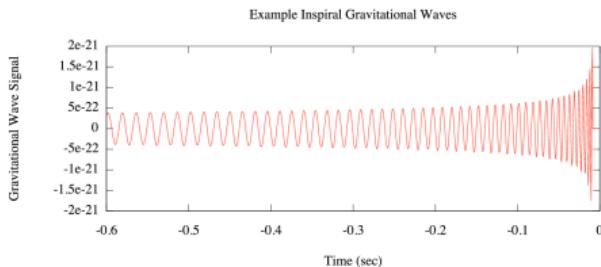
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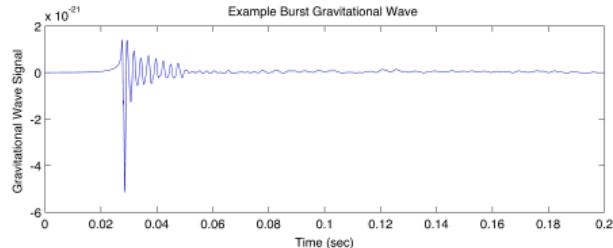
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Image credit: [LIGO Scientific Collaboration](#)

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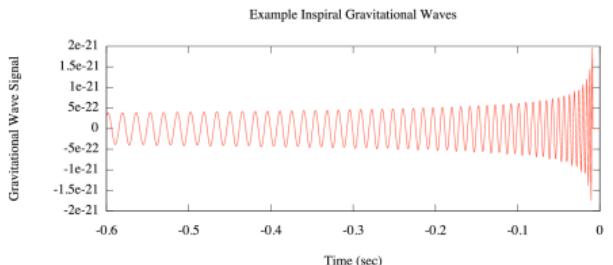
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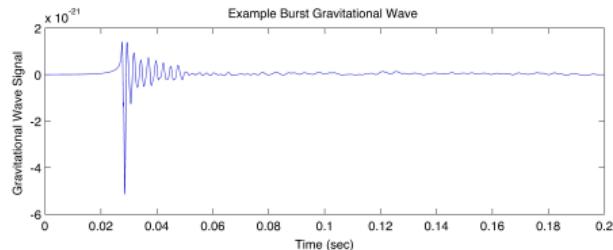
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Image credit: [LIGO Scientific Collaboration](#)

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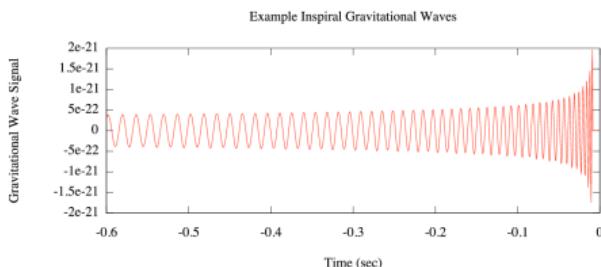
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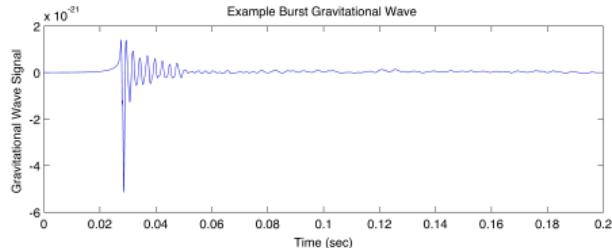
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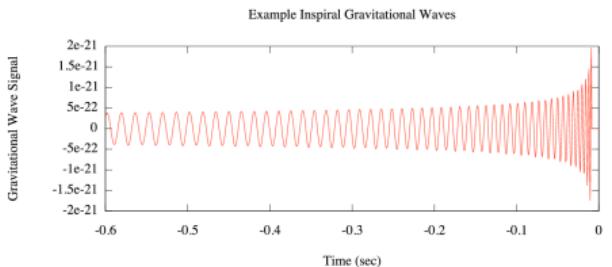
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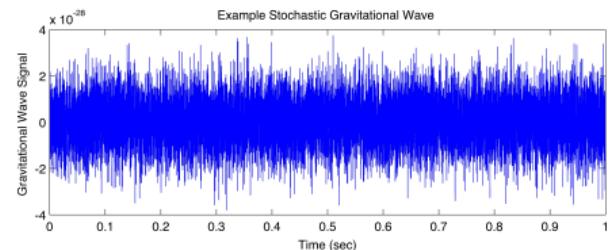


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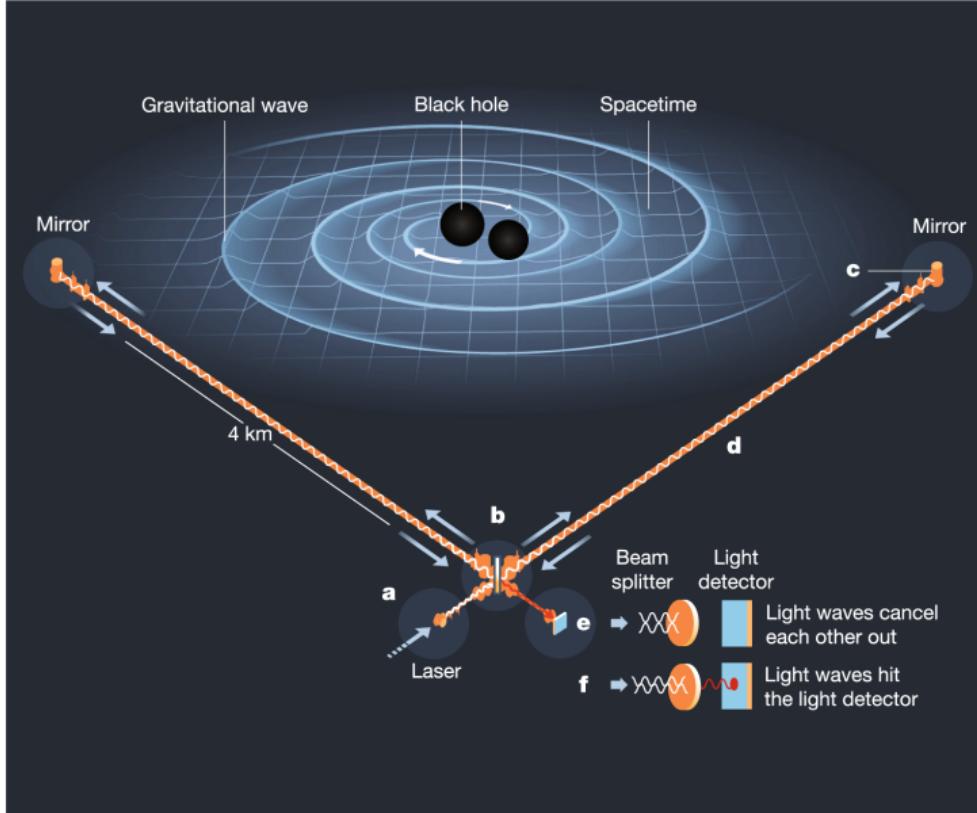
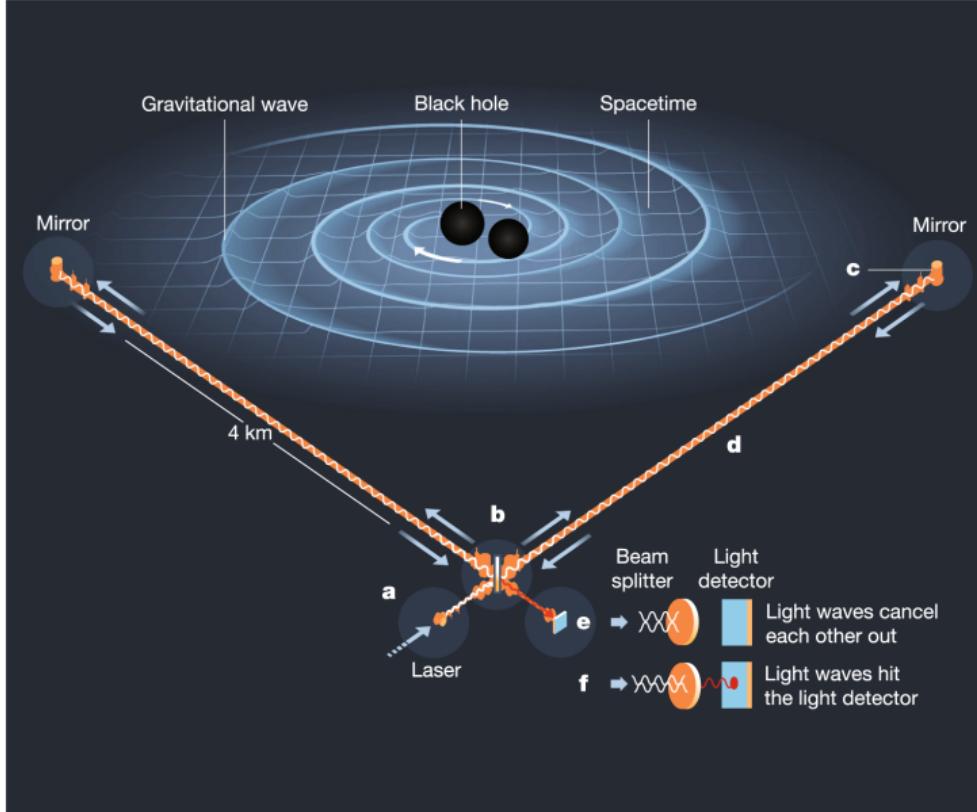


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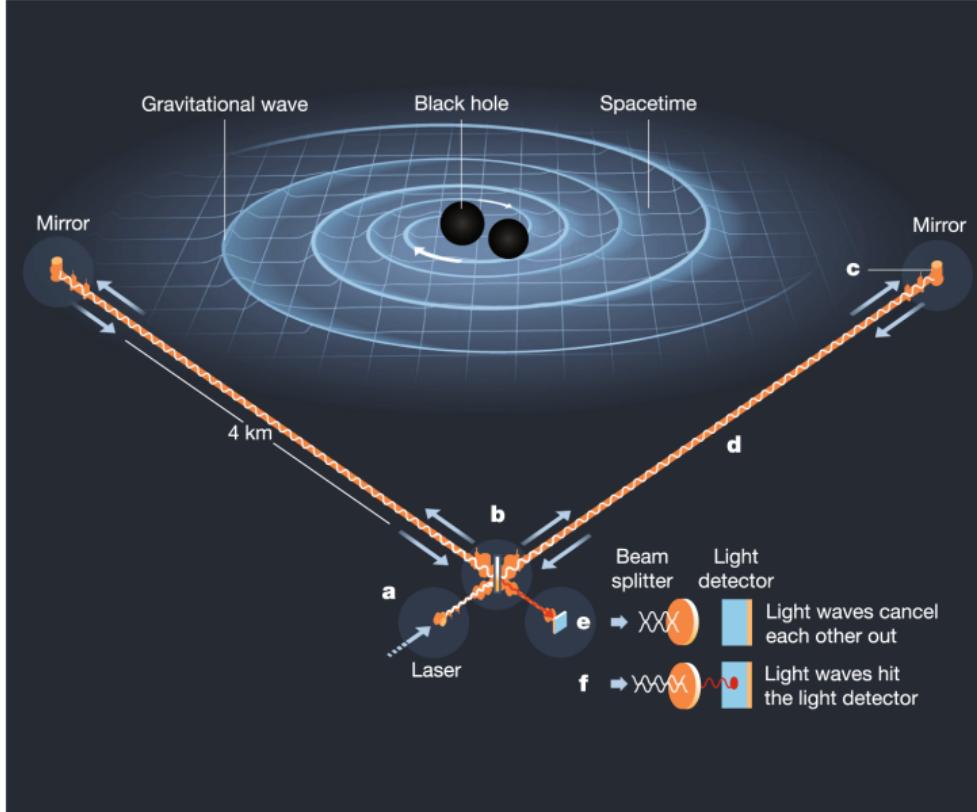
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Other detection methods:

Image credit: [The Royal Swedish Academy of Sciences](#)

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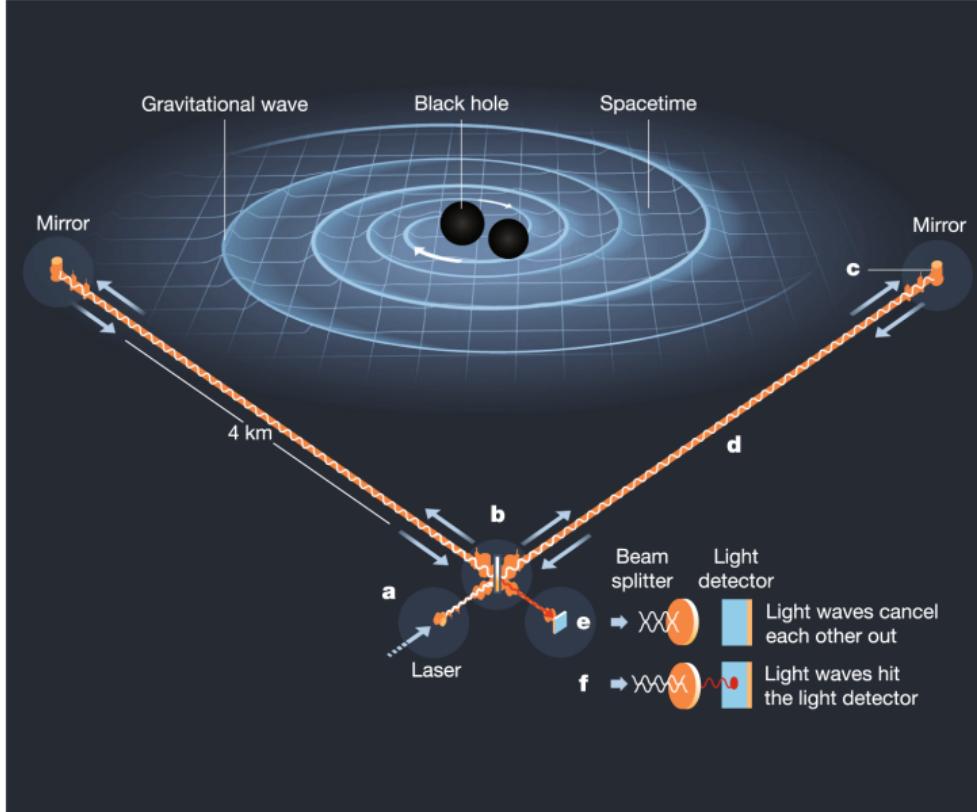


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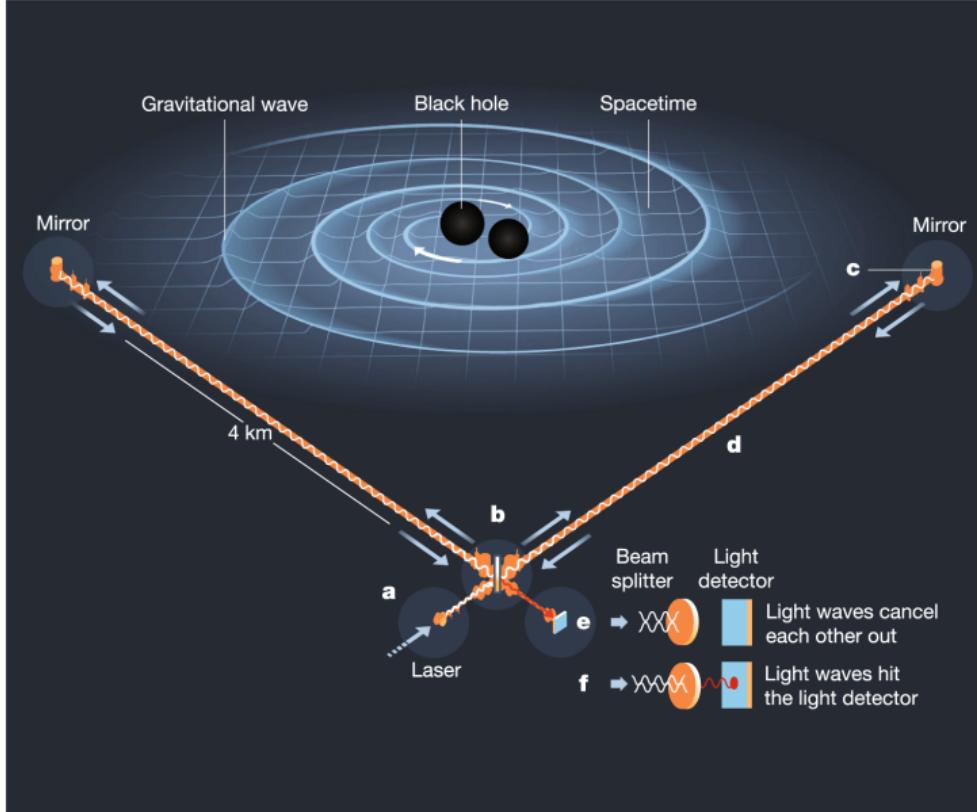


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- ▶ Pulsar timing arrays

Image credit: [The Royal Swedish Academy of Sciences](#)

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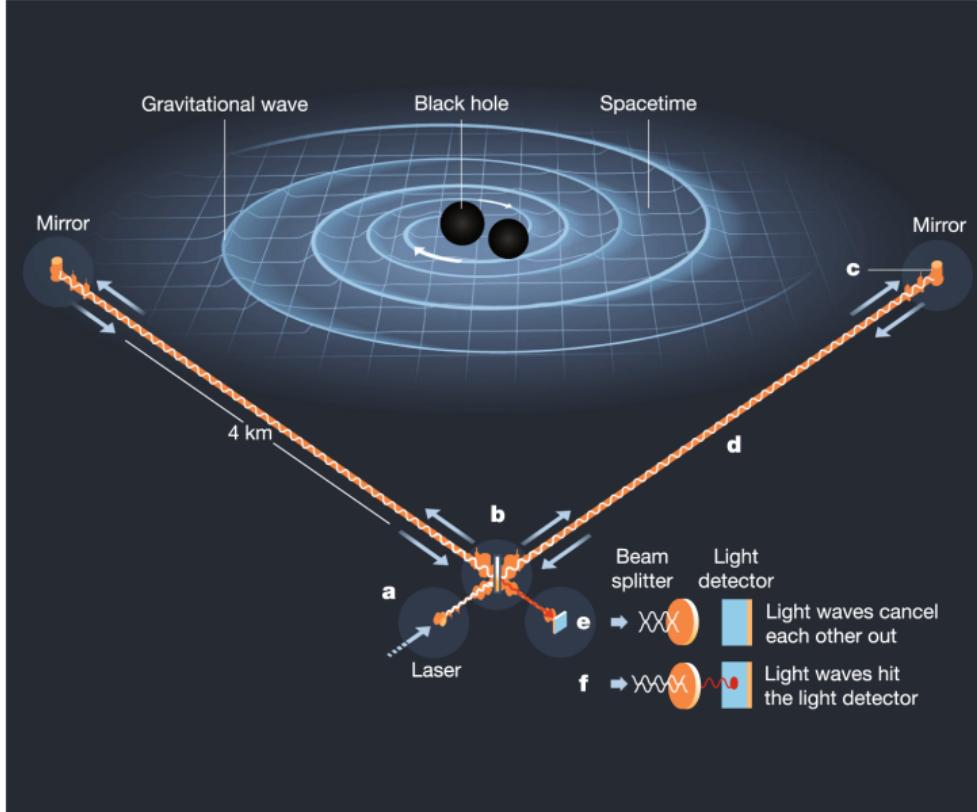


Other detection methods:

- ▶ Weber bar
- ▶ Pulsar timing arrays
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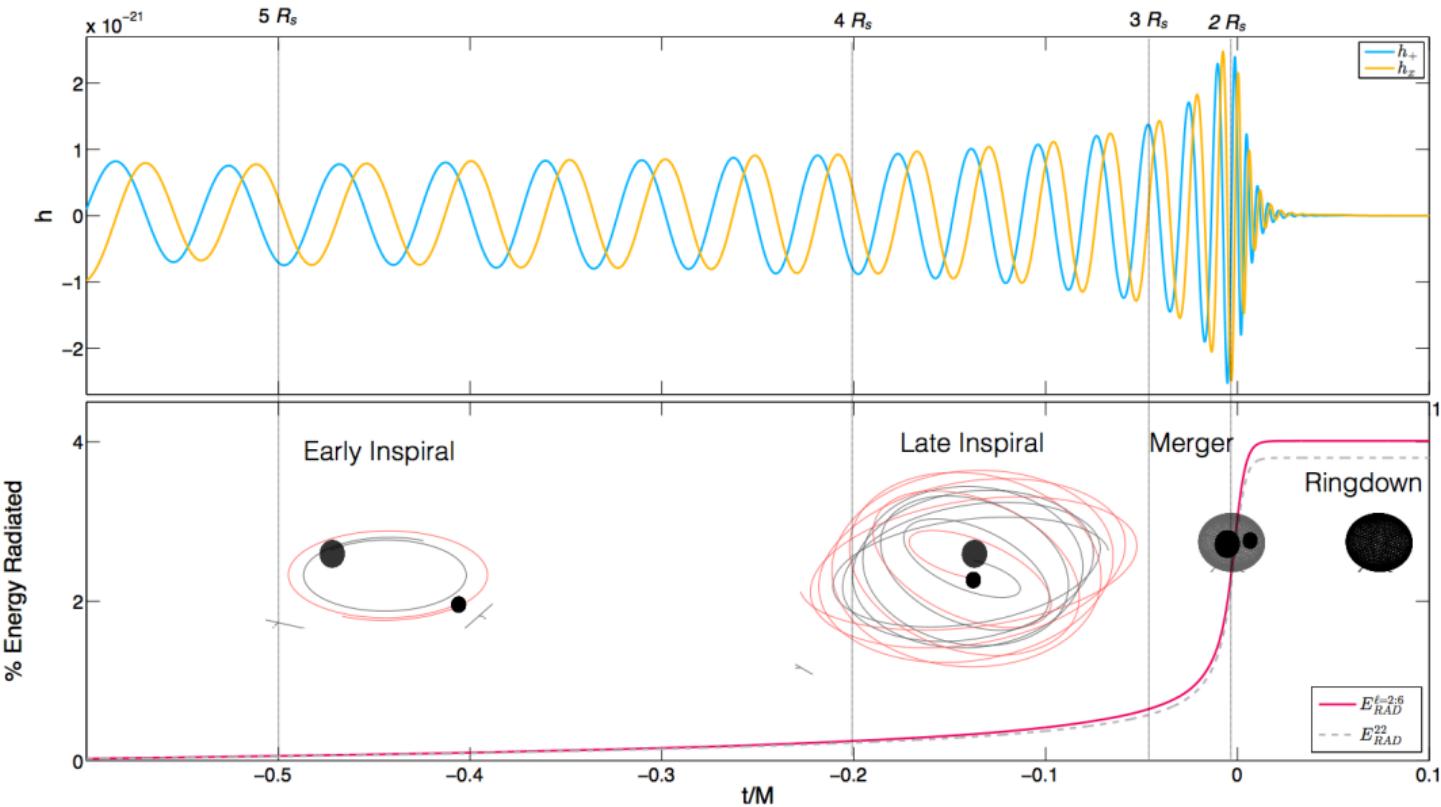


Other detection methods:

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- ▶ Other non-direct ways

Image credit: [The Royal Swedish Academy of Sciences](#)

# Gravitational waves from black hole mergers



# Stochastic gravitational wave background



Image credit: [Megna pools](#)

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## Section 2

# **Primordial Black Holes**

## Primordial black hole formation

In the early universe ( $\sim$  before the first second), if some part of the cosmos that has the same density as a black hole becomes causally connected (a.k.a. the corresponding mode enters the horizon) that part of the cosmos collapses to a black hole.

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$$M_{PBH} \sim \frac{c^3 t}{G} \sim 10^{-15} \left( \frac{t}{10^{-23} s} \right) gr$$

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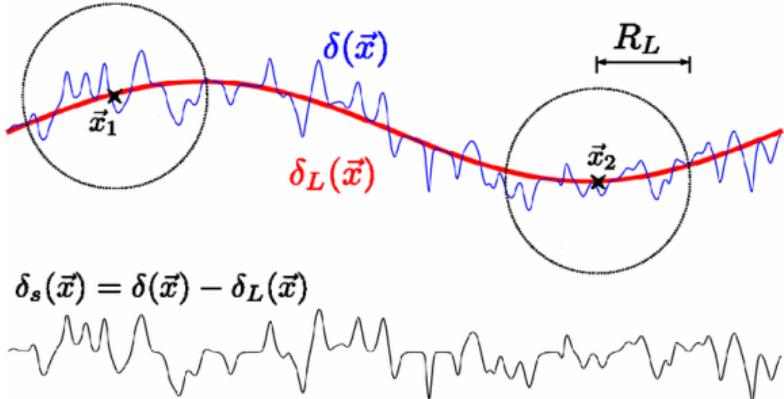


Image credit: F. Schmidt, et al.

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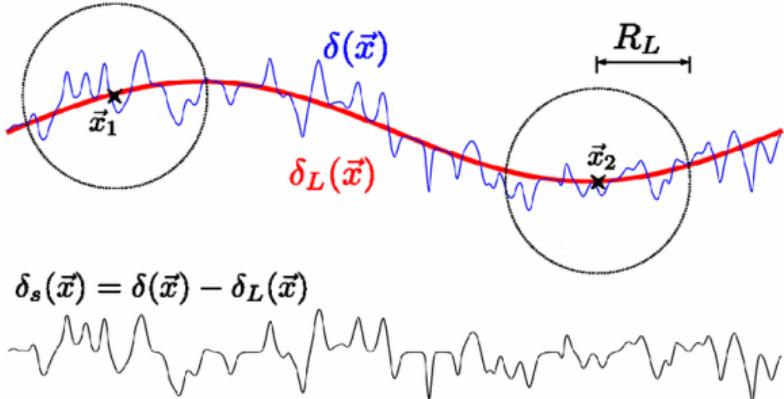


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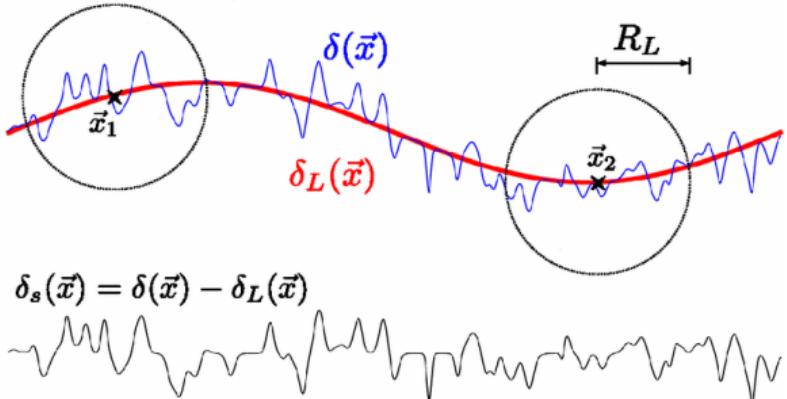


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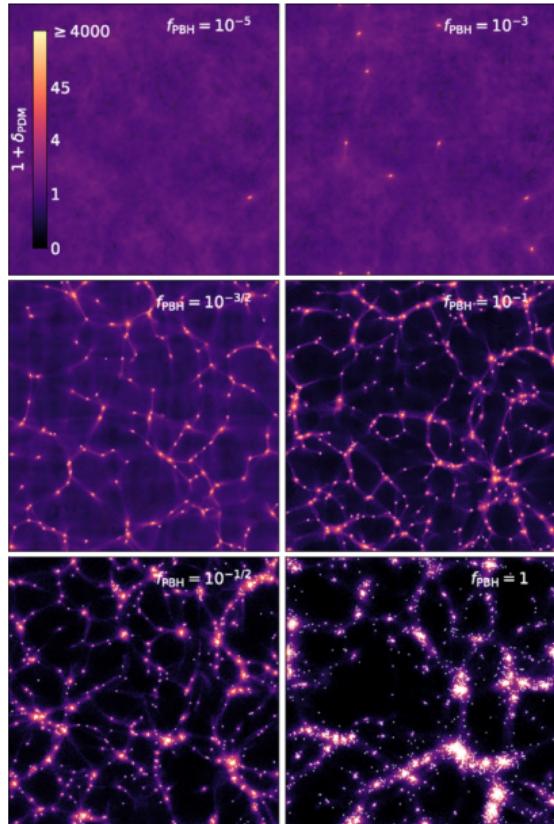


Image credit: D. Inman, et al.

## Section 3

# Simulations

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3. **Semi-analytical approach:** For quickly testing various complicated models without using too much computational resource (e.g. [Braglia, et al.], [Durrer], [Mukherjee, et al.])

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$$f_r = (1+z)f$$

## Cosmology part

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# Astrophysics/Cosmology part

Following the picture in [M. Braglia, J. Garcia-Bellido, and S. Kuroyanagi, Testing primordial black holes with multi-band observations of the stochastic gravitational wave background, *Journal of Cosmology and Astroparticle Physics*, vol.2021, no.12, p.012, 2021.]

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To fix  $R_{\text{clust}}$  :

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$$f_{PBH}(m, \{\boxed{\sigma_{PBH}}, \mu\}) = F_0 \frac{1}{\sqrt{2\pi}\sigma_{PBH}m} \exp\left[-\frac{(\log_{10} \frac{m}{\mu})^2}{2\sigma_{PBH}^2}\right]$$

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$$\mathcal{G}(f_r) = \begin{cases} f_r^{-1/3} & f_r < f_{\text{merg}} \\ \frac{f_r^{2/3}}{f_{\text{merg}}} & f_{\text{merg}} \leq f_r < f_{\text{ring}} \end{cases}$$

## Source part

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# From spectral energy density to wave forms

page 11

Stochastic gravitational wave background



"We are like an insect sitting at the edge of a pool, trying to figure out who jumped in where and when and what's happening all over the pool."

- Richard Feynman

Image credit: Megna pools

$$\Omega_{SGWB}(f) = \frac{1}{\rho_c} \frac{d\rho_{GW}(f)}{d \ln f}$$

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$h_{ab}(t, \vec{x})$

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# From spectral energy density to wave forms

$$h(t) = \int_{-\infty}^{\infty} df h(f) e^{2\pi i f(t-\phi)}$$

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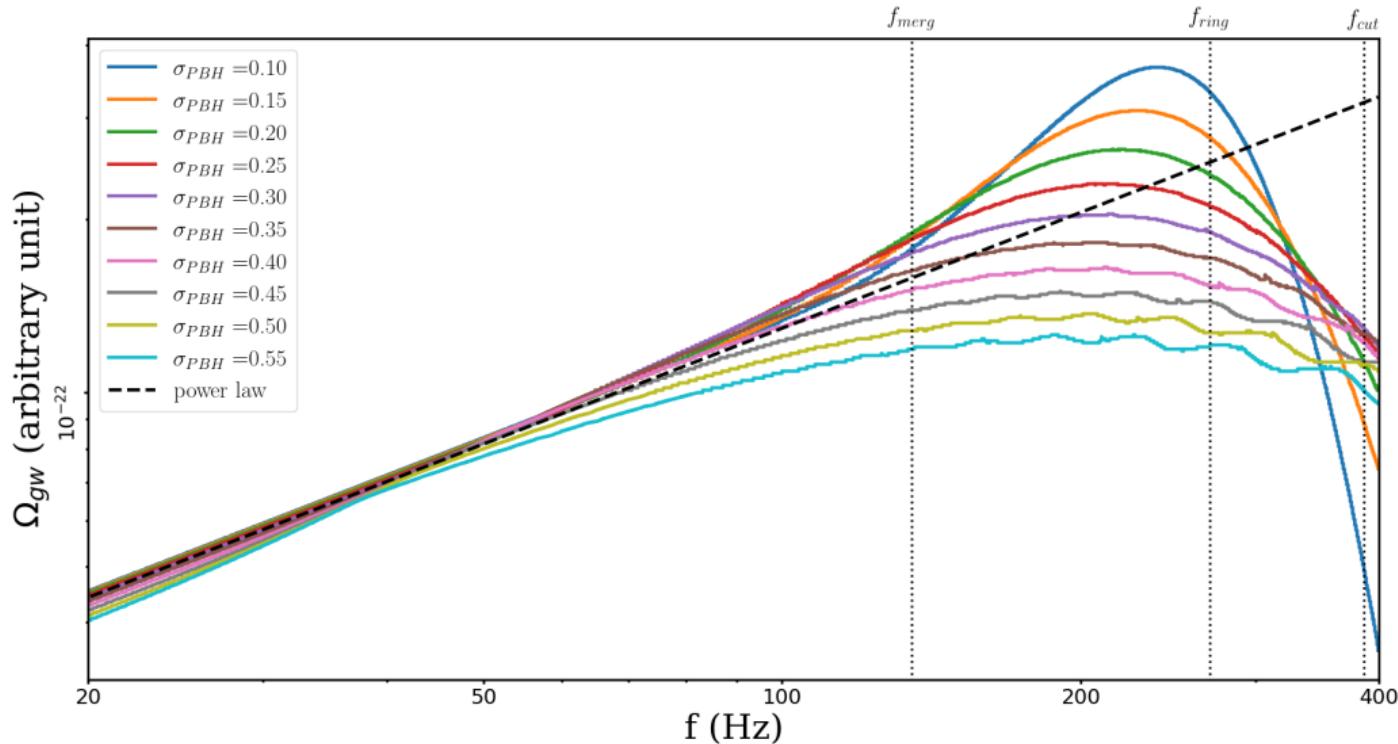
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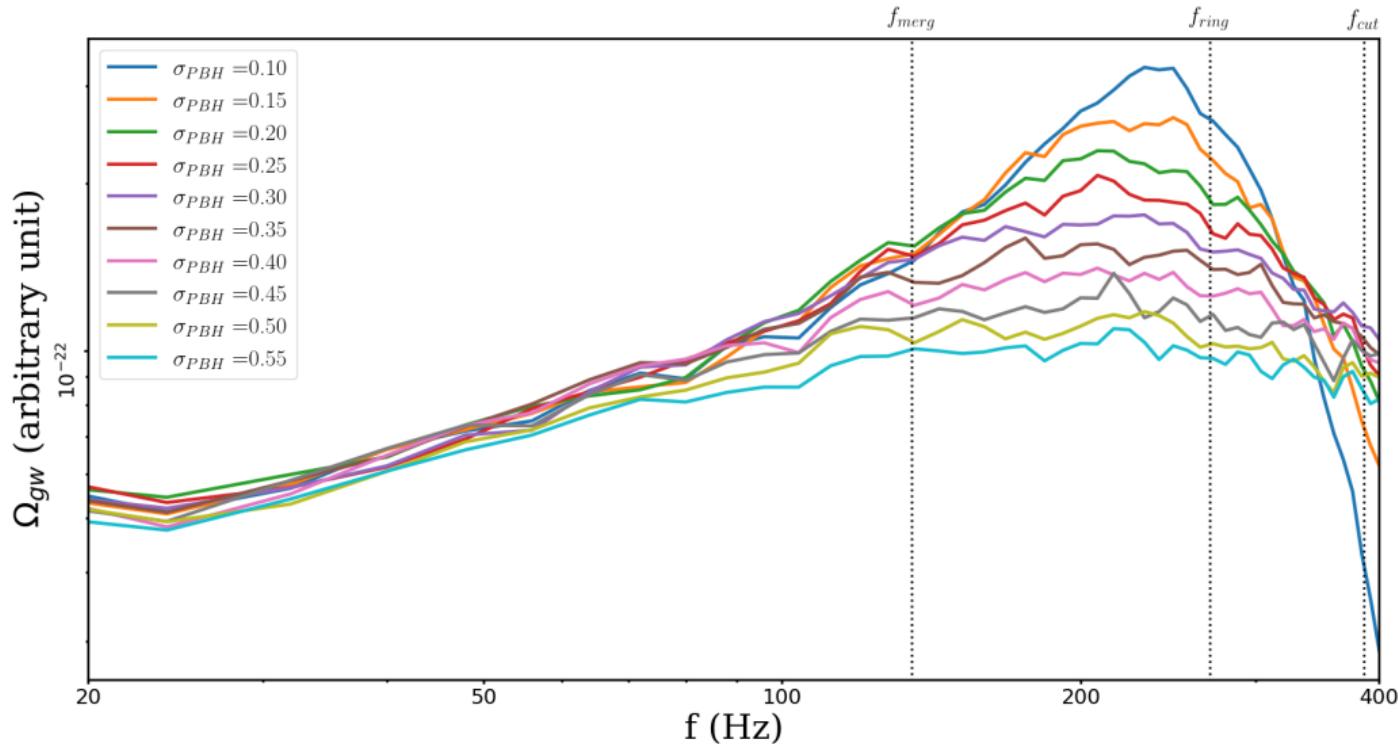
To achieve this we used *LALSuite* package with a few added features:

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- ▶ Options to access these new functionalities with an app

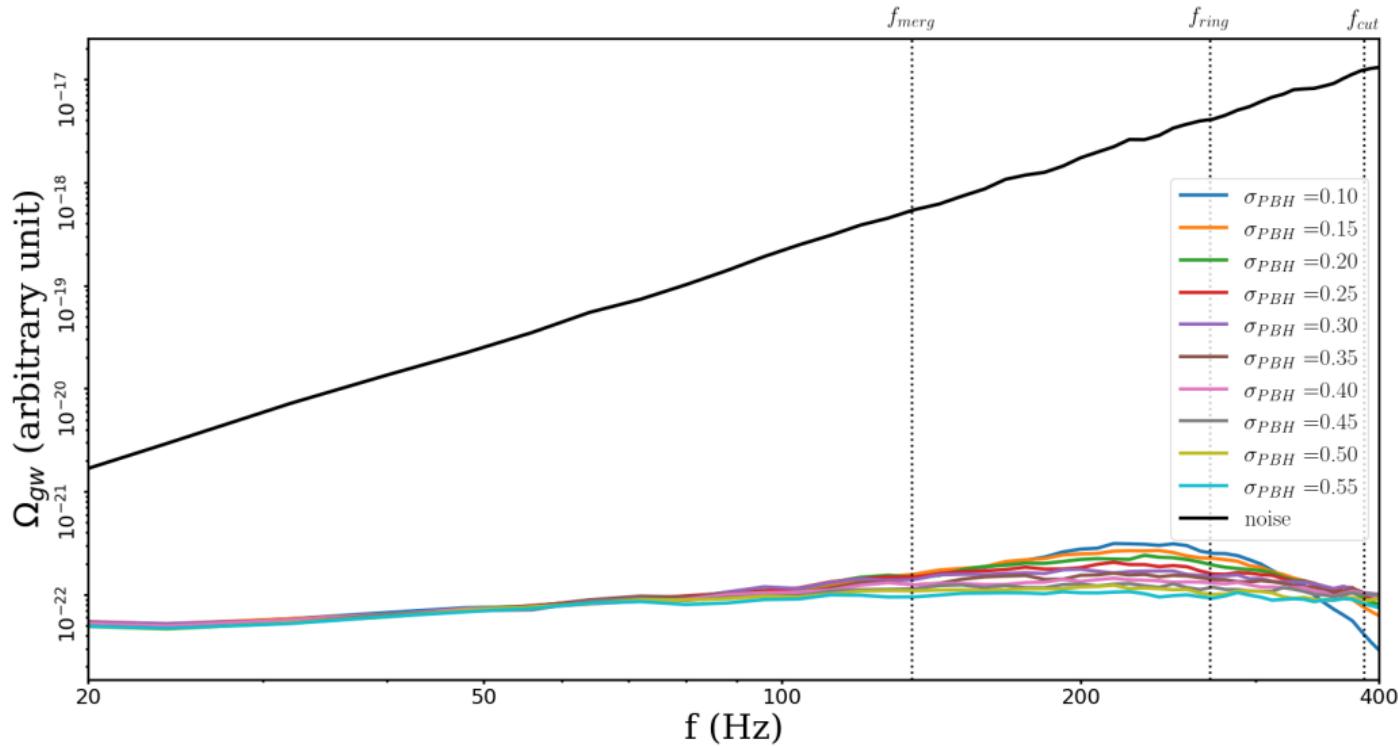
# Spectral energy density plots (theoretical)



# Spectral energy density plots (numerical)



# Spectral energy density plots (noisy)



# Recap

We want to detect:

- ▶ A stochastic gravitational wave background
- ▶ Coming from binary black hole mergers
- ▶ Our chosen black holes to study have a primordial origin

To achieve these, we need:

1. Learn about primordial black holes and their population statistics
2. Build a model for their merger rates based on the population statistics
3. Simulate a stochastic gravitational wave background
4. Build a model to analyse the data
5. Run the analyses

## Section 4

# Topological Data Analysis

# What is topology?



Image credit: [Physics world](#)

# What is topology?



Topology:

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- ▶ General Topology or Point Set Topology
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- ▶ ***Algebraic Topology***

# Homeomorphism

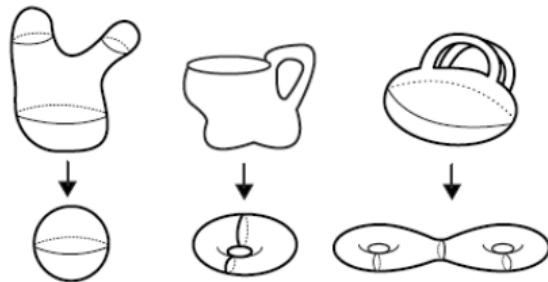


Image credit: [Annenberg Learner](#)

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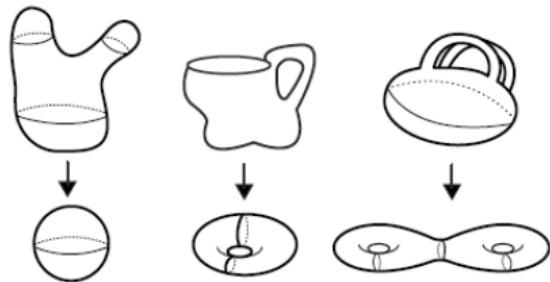


Image credit: [Annenberg Learner](#)

Me: Mom can we  
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Donuts at home:



Image source: [Reddit](#)

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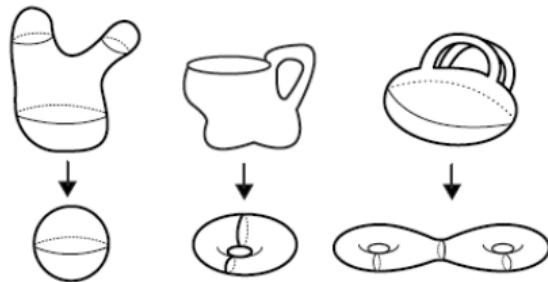


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$$\beta_k(\psi)$$

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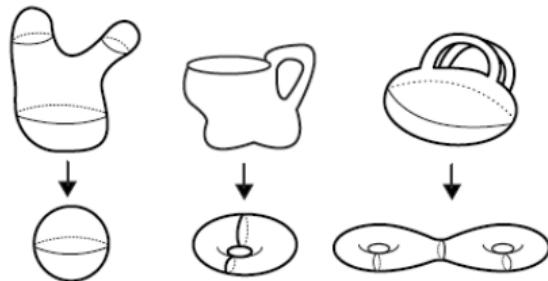


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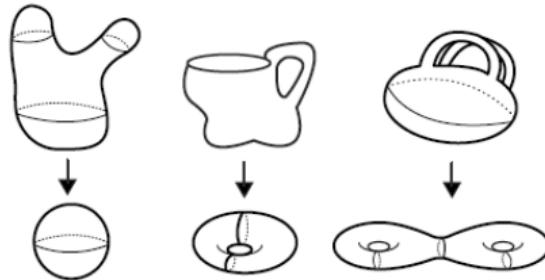


Image credit: [Annenberg Learner](#)

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Topological space \ Betti number	•	/	○	(○)	(○)	(○)
$\beta_0$	1	1	1	1	1	1
$\beta_1$	0	0	1	0	2	4
$\beta_2$	0	0	0	1	1	1

Image credit: [Masoomy, et al.](#)

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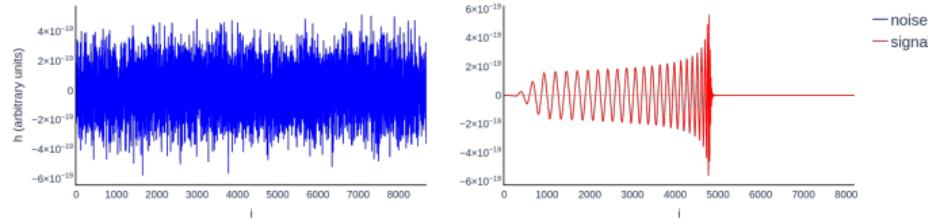
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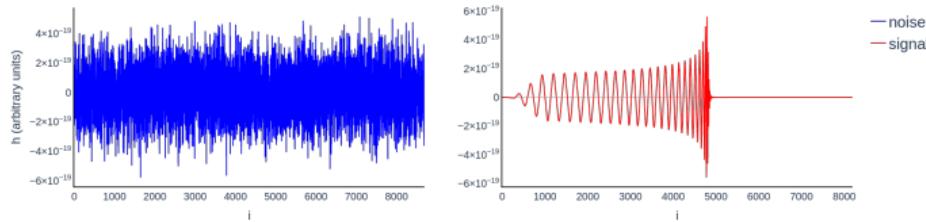
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# Topology of time series



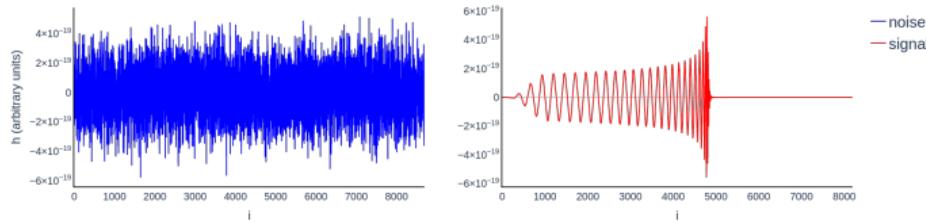
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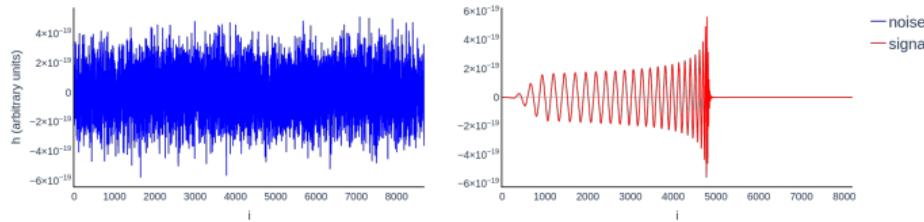
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$$x(t) = (x_1, x_2, \dots, x_i, x_{i+1}, \dots, x_n)$$

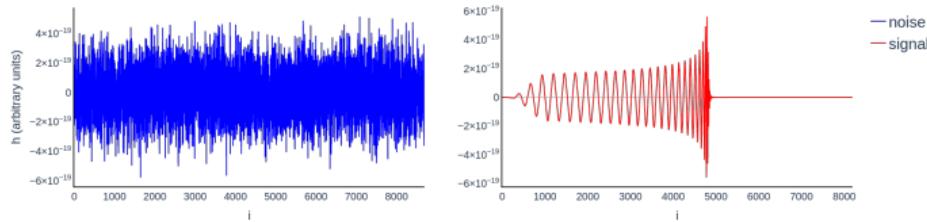
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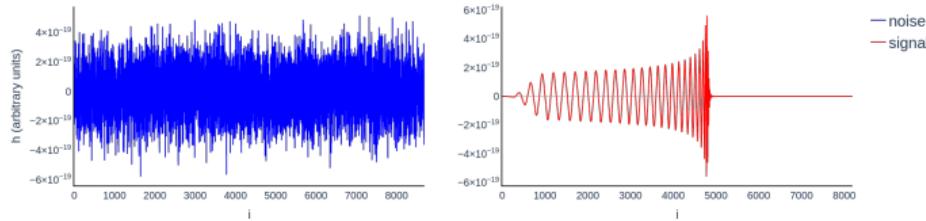
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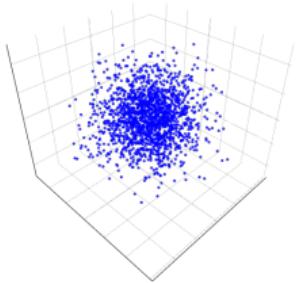
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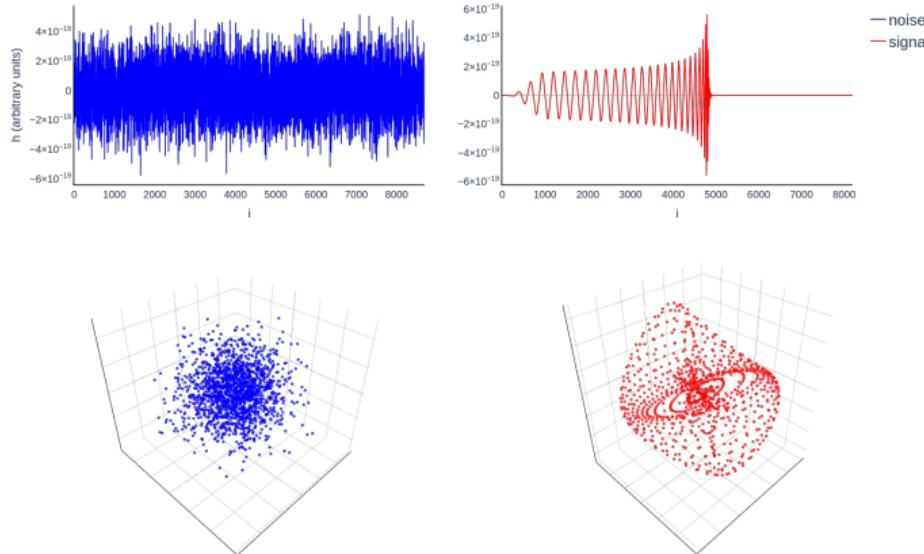


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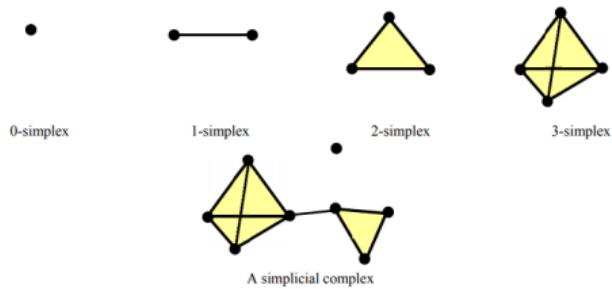


Image credit: [Zulkepli, et al.](#)

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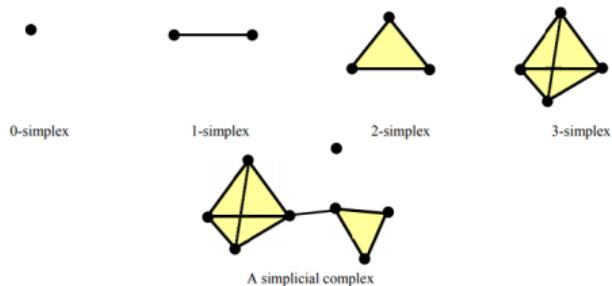


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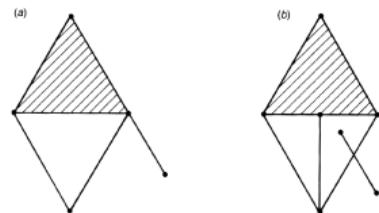


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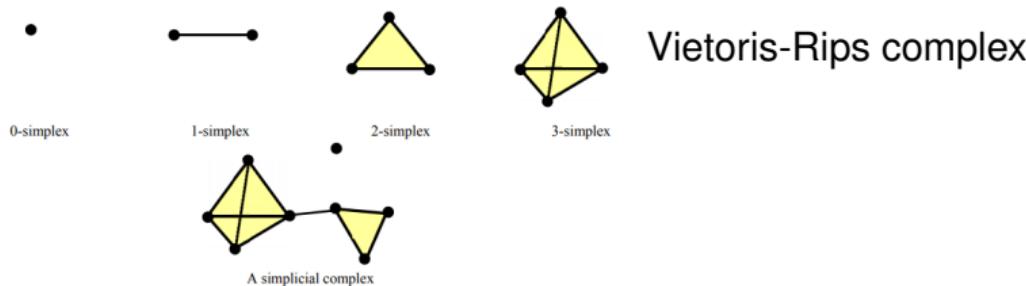


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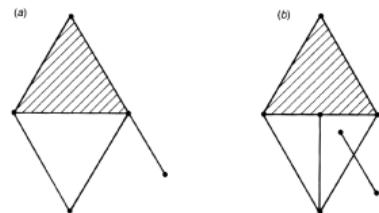


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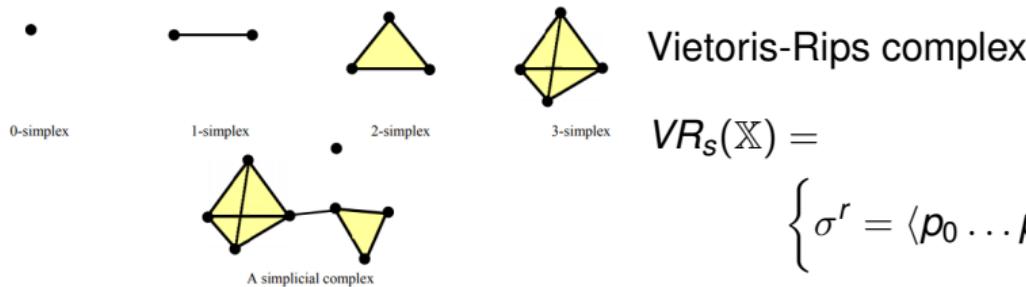


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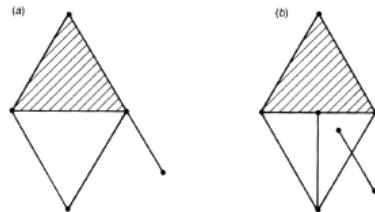


Image credit: [Nakahara](#)

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In simple terms, homology theory tries to connect the dots in a topological space together and figure out what shape they make

To achieve this, homology theory uses *Simplexes* and *Simplicial complexes*

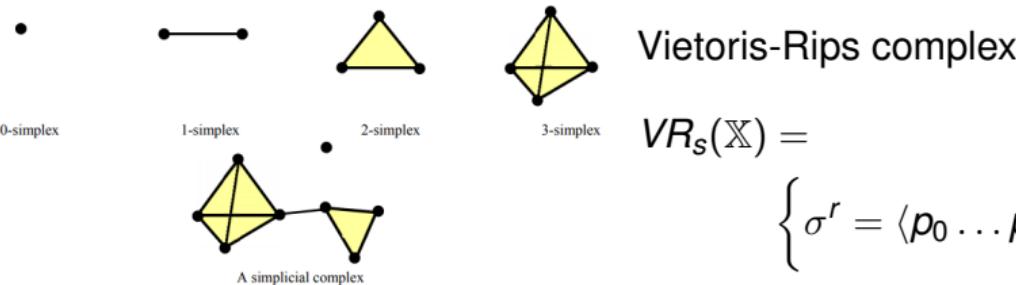


Image credit: Zulkepli, et al.

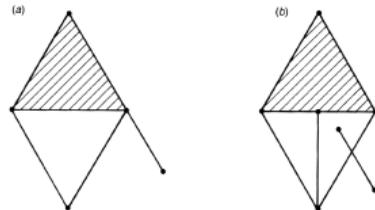


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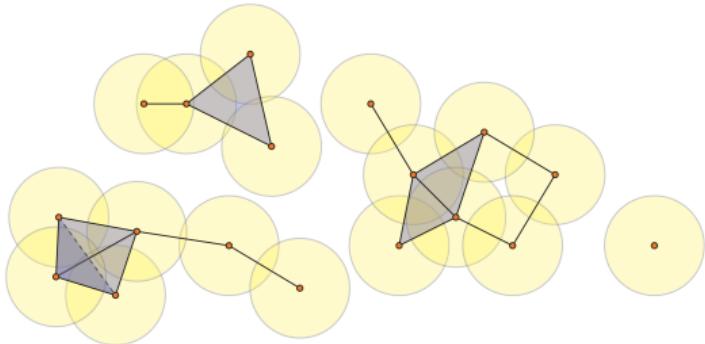
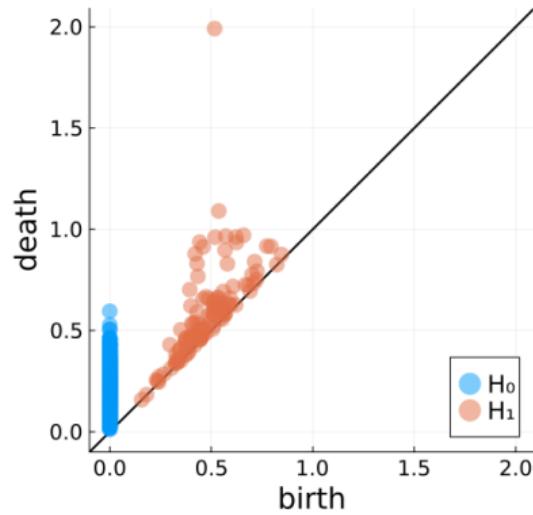
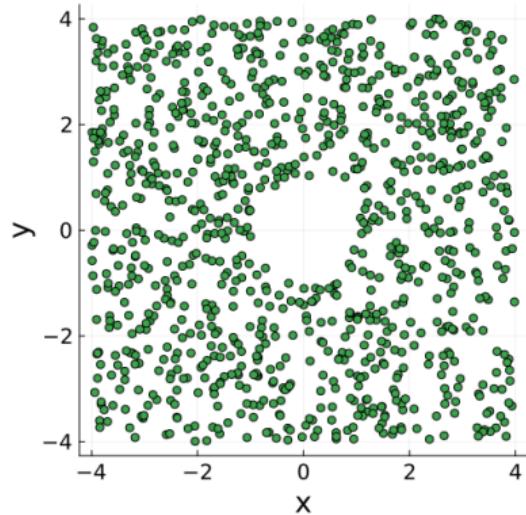
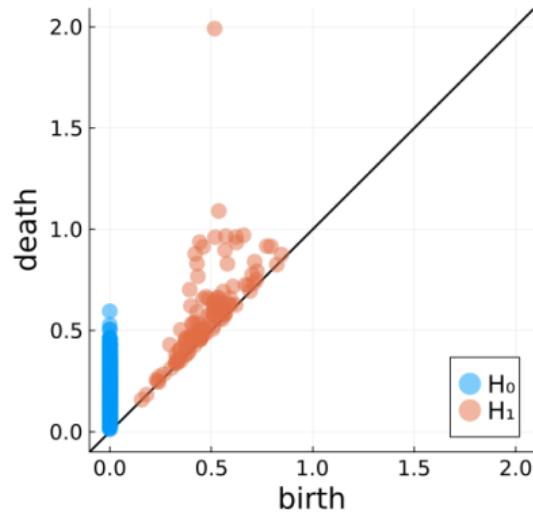
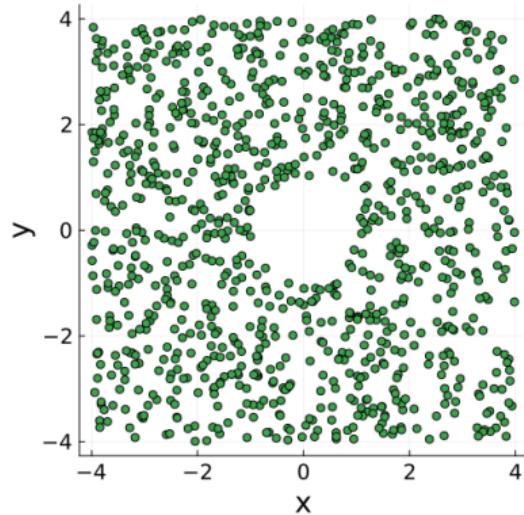


Image credit: Topaz, et al.

# Persistence homology

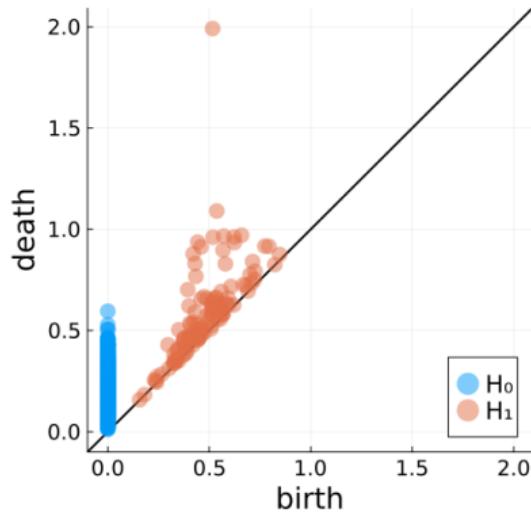
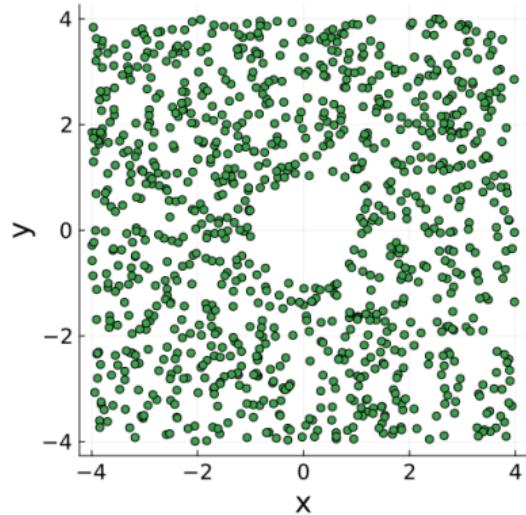


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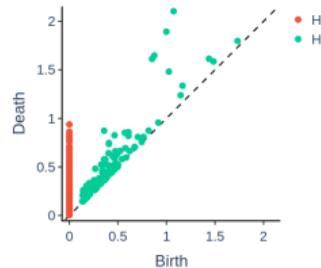
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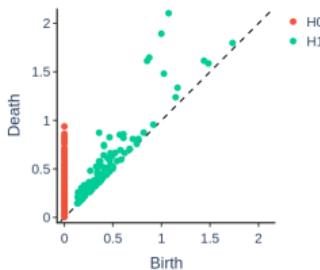
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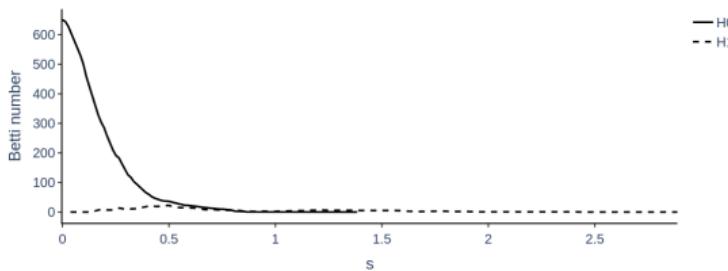


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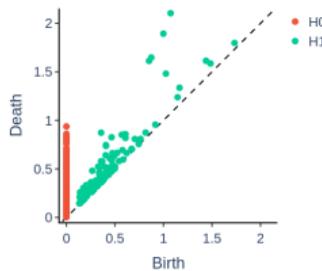
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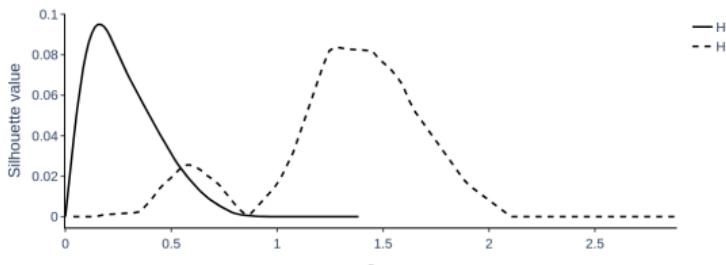
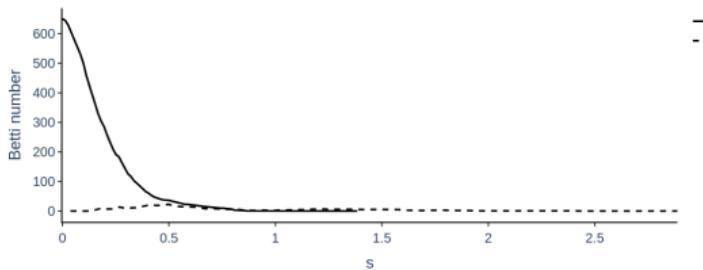


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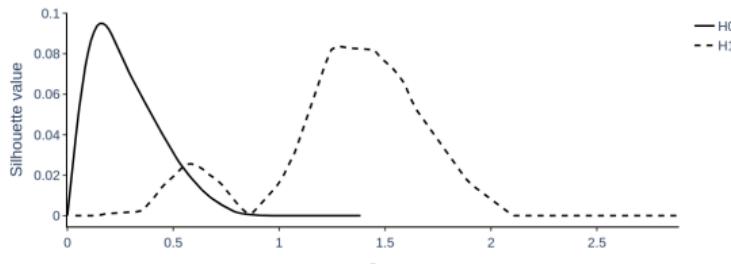
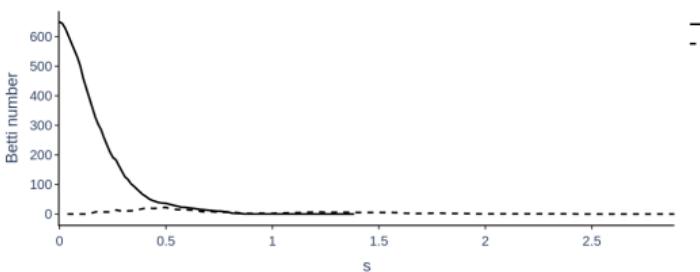
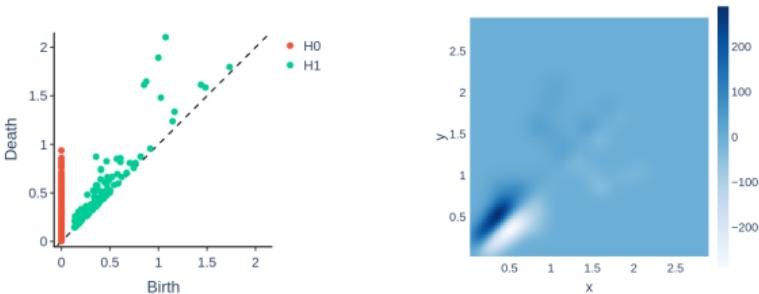
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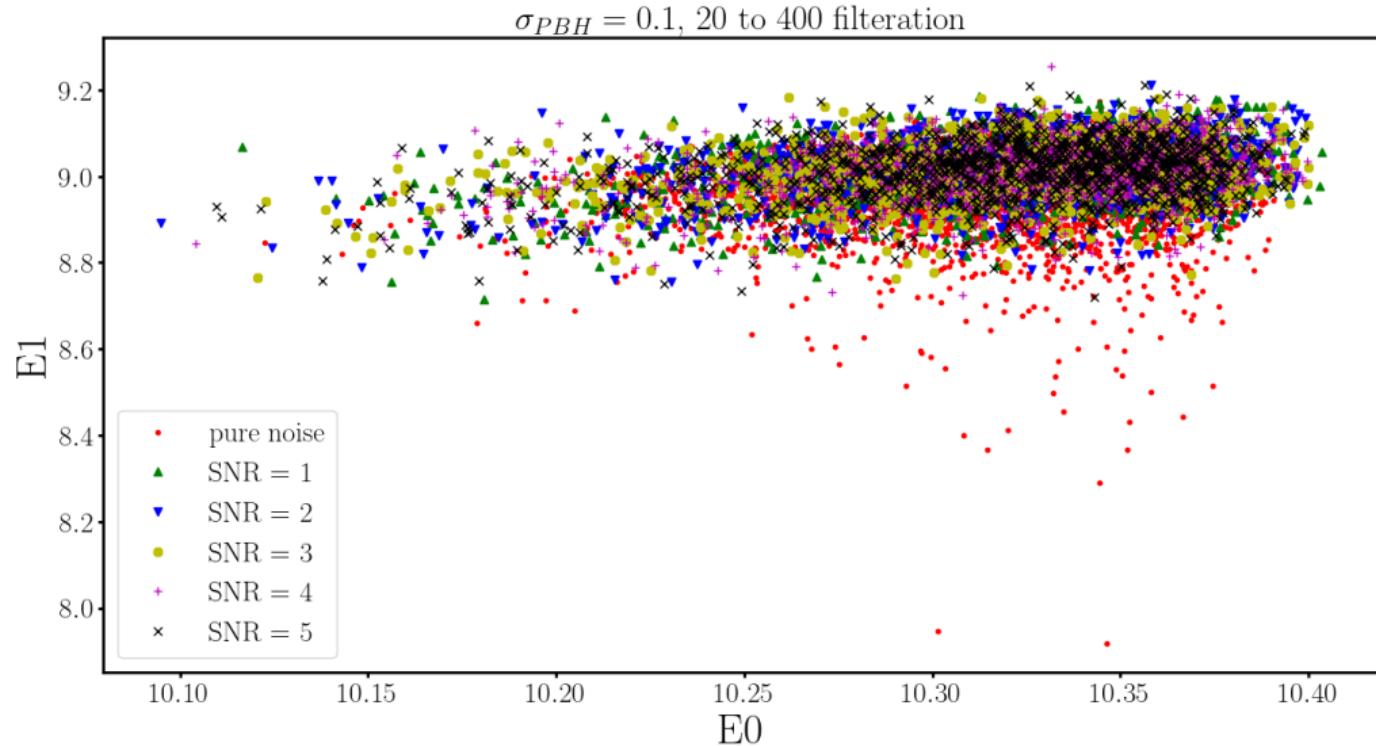
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- ▶ Find topological features to detect a signal amidst different levels of noise
- ▶ Find topological features to discriminate between different values of  $\sigma_{PBH}$
- ▶ A classification method to classify signals with different values of  $\sigma_{PBH}$

## Section 5

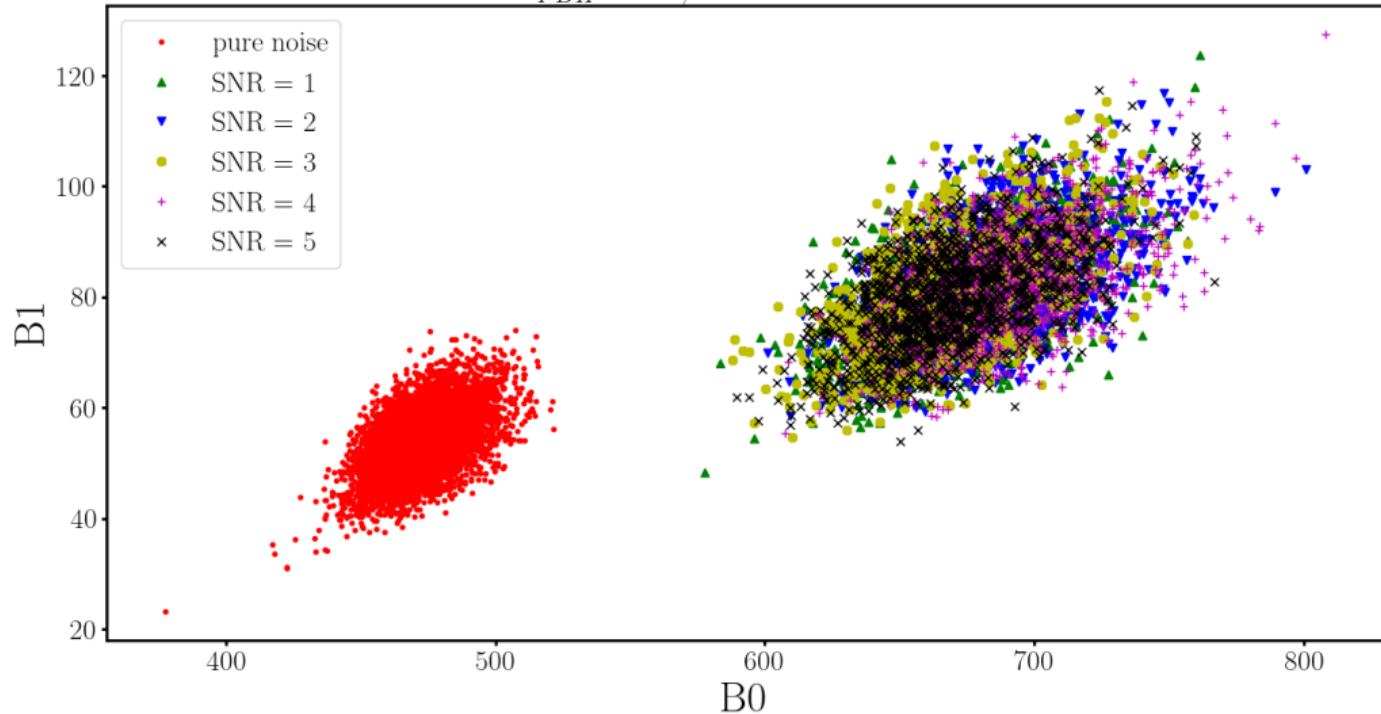
# Results

# Entropy measures

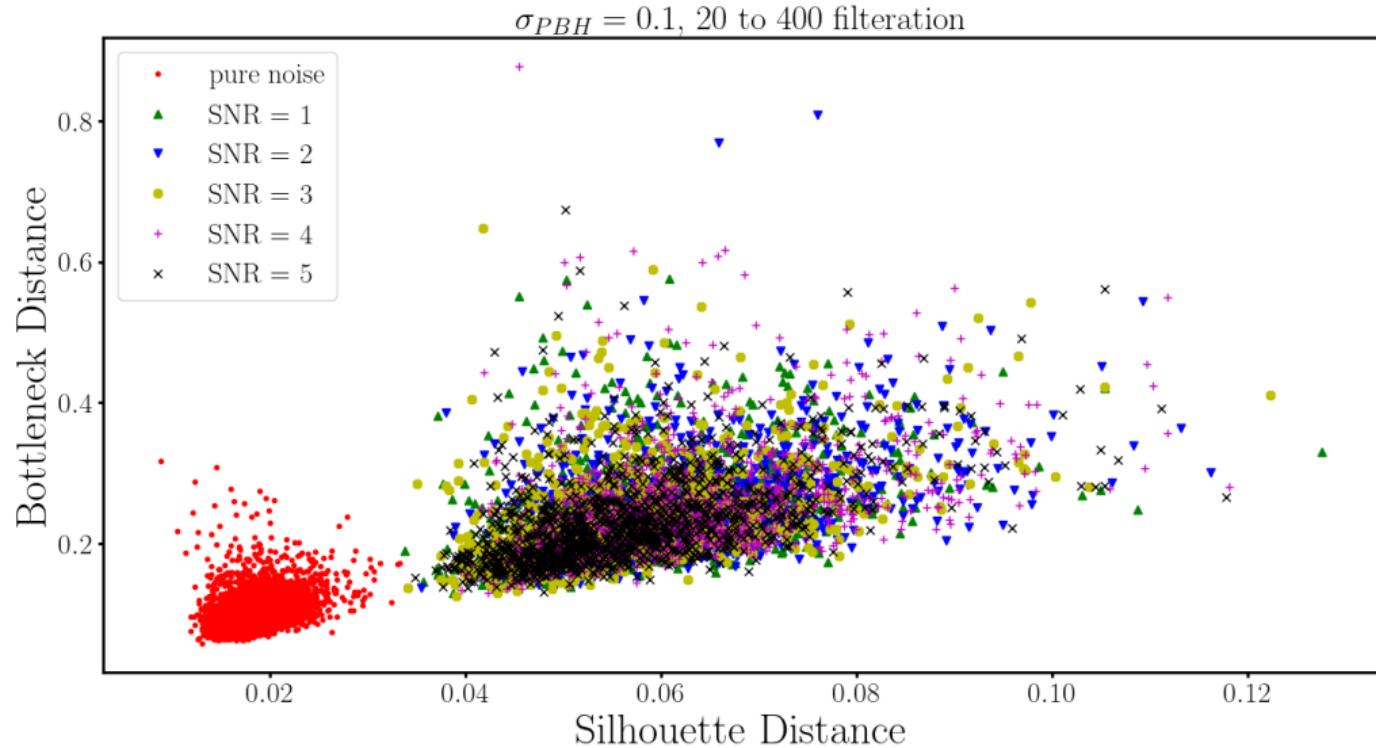


# Distance measures

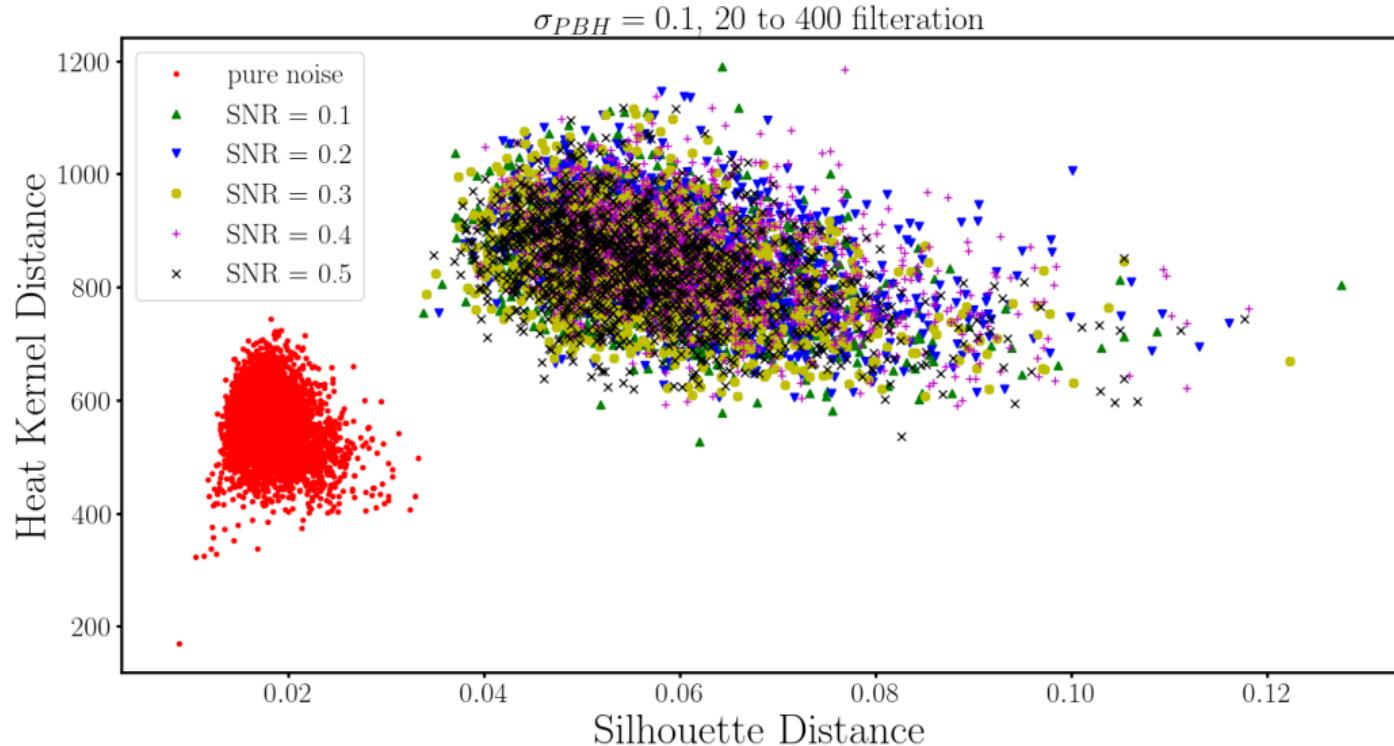
$\sigma_{PBH} = 0.1, 20 \text{ to } 400$  filtration



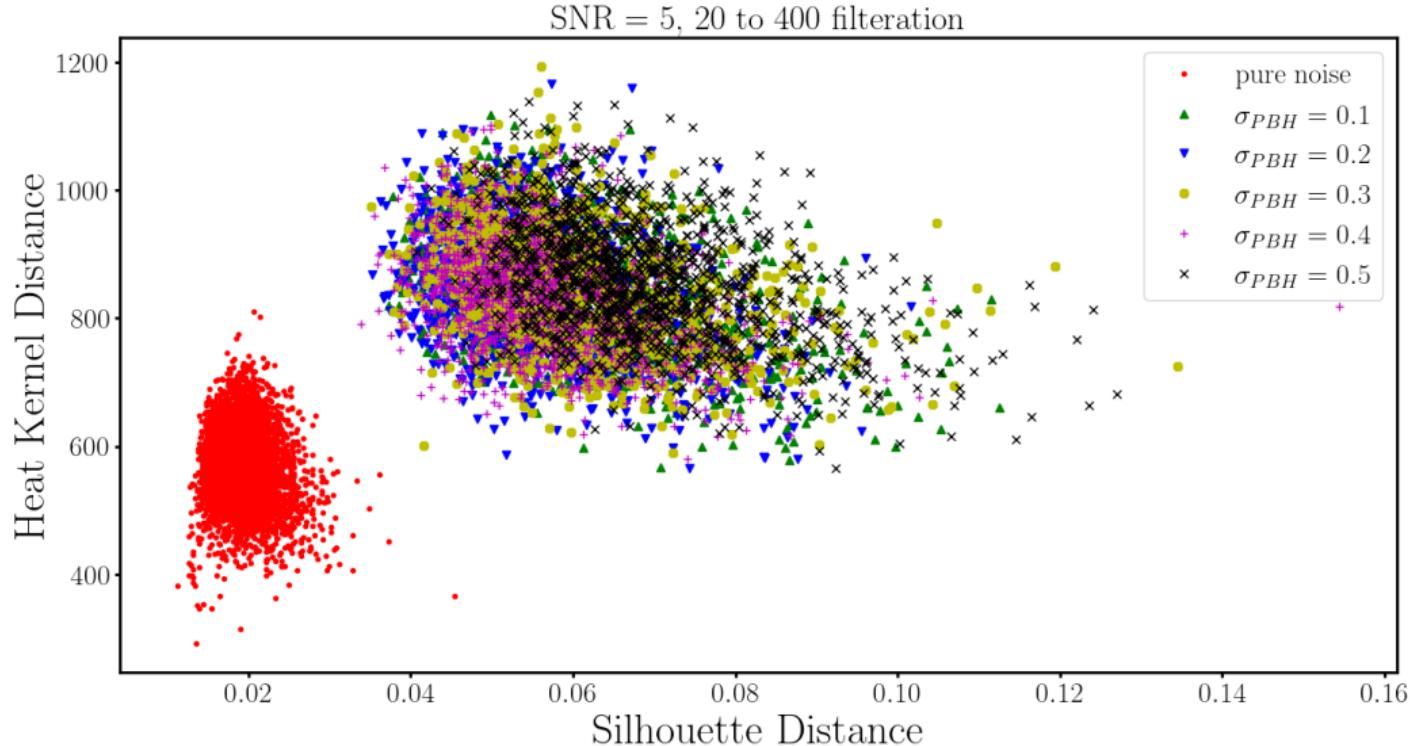
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# First attempt to classification



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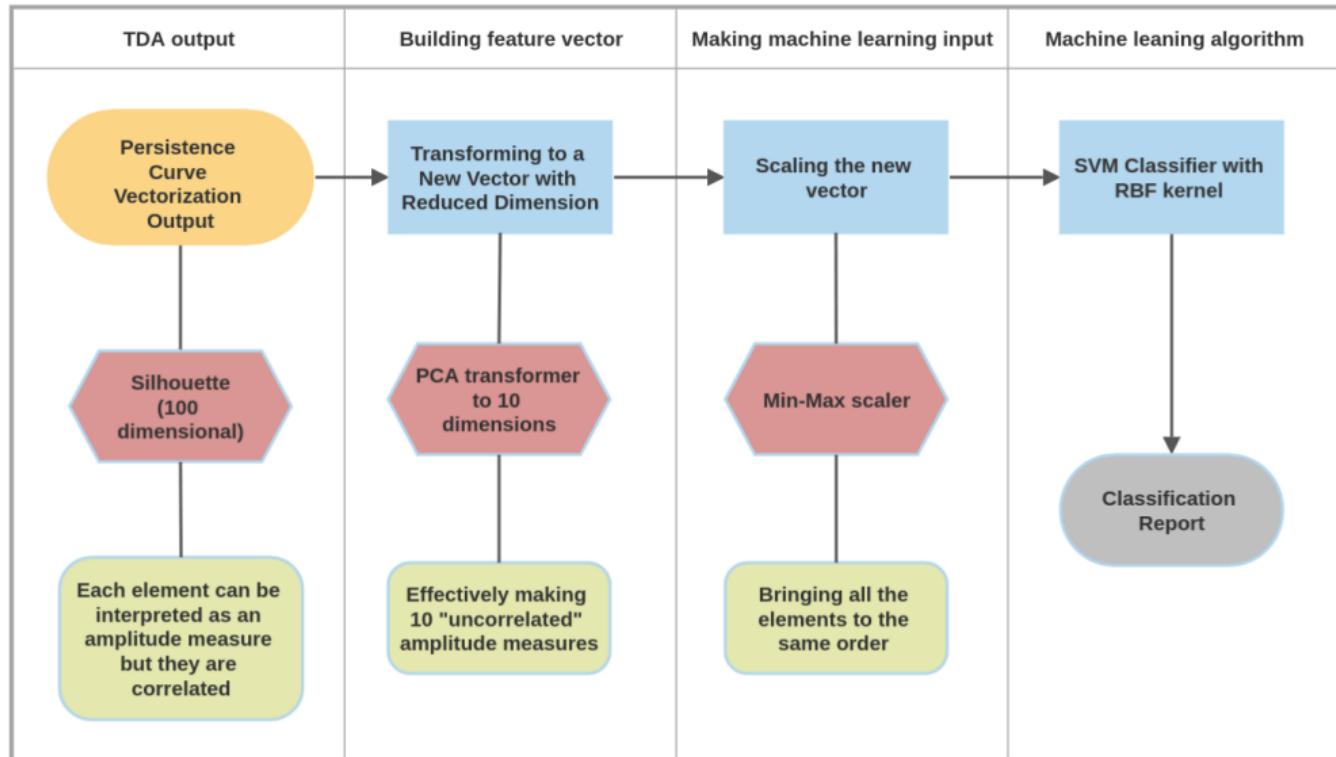
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Hence to build a classification pipeline, we need to use more complicated topological measures.

# A classification pipeline



# Classification report

Physical parameter	precision (%)	recall (%)	f1-score	support
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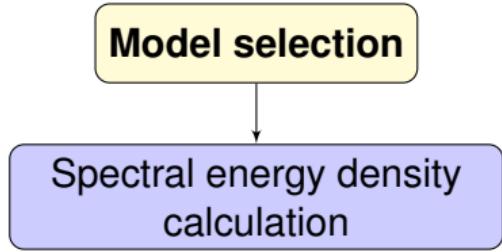
## Section 6

# **Conclusion and Outlook**

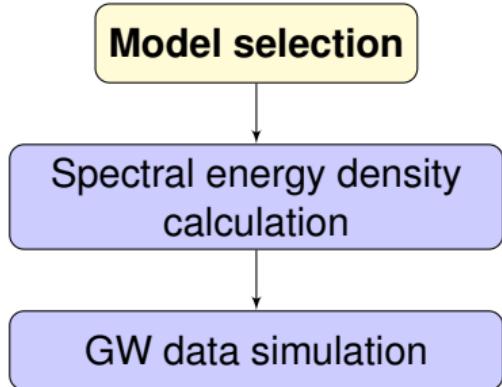
# Our work in perspective

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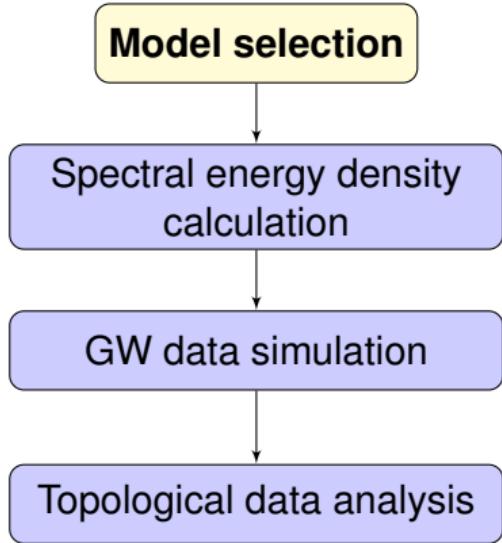
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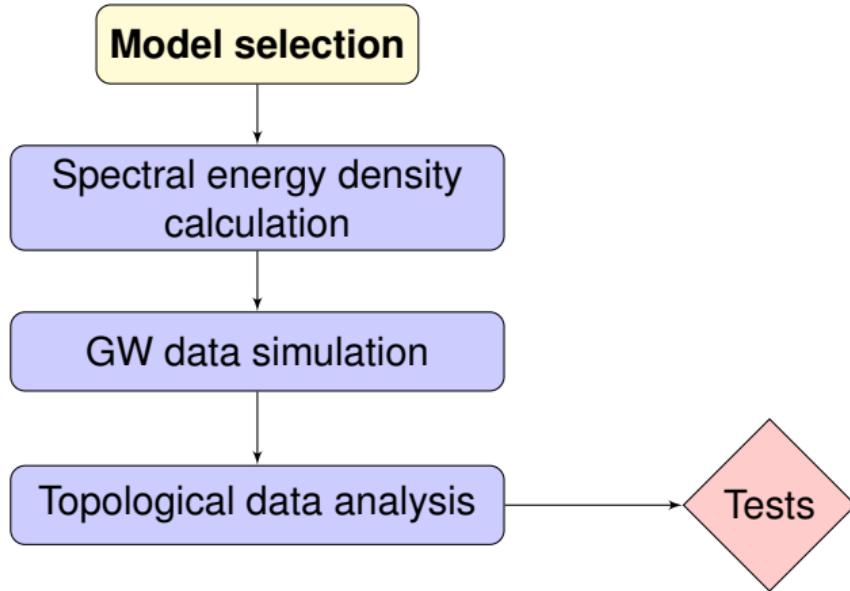
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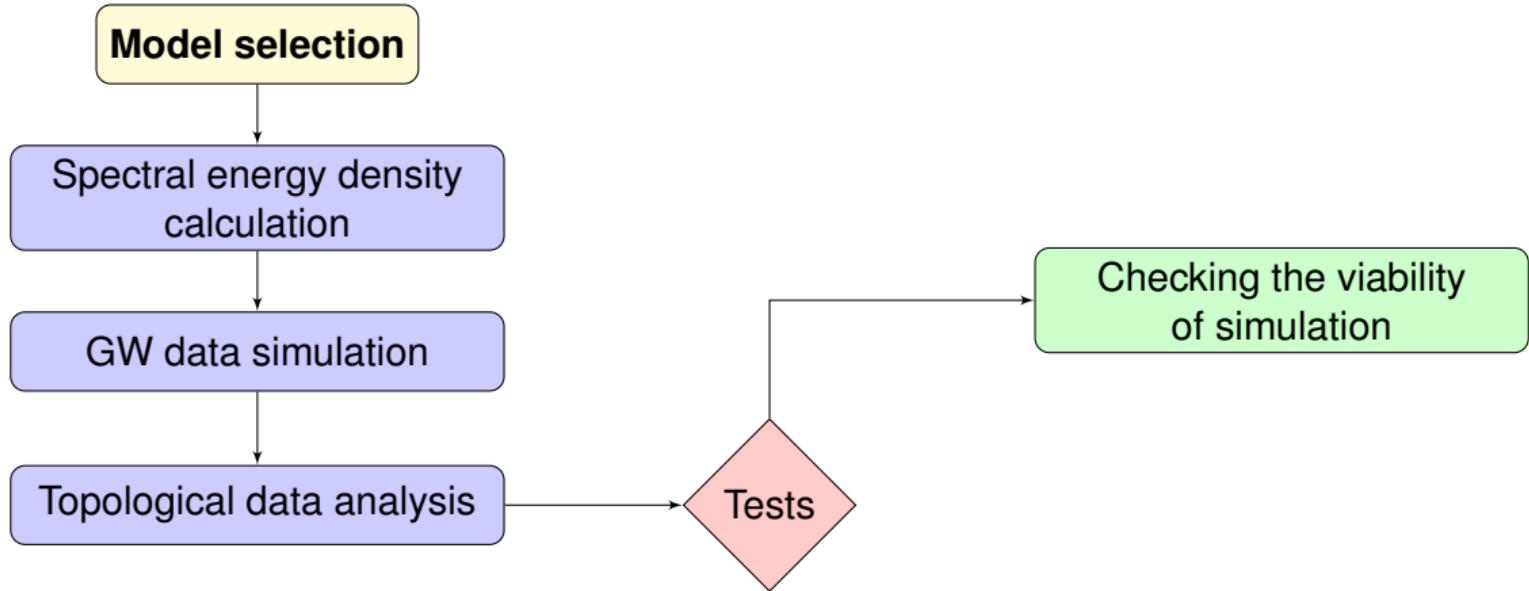
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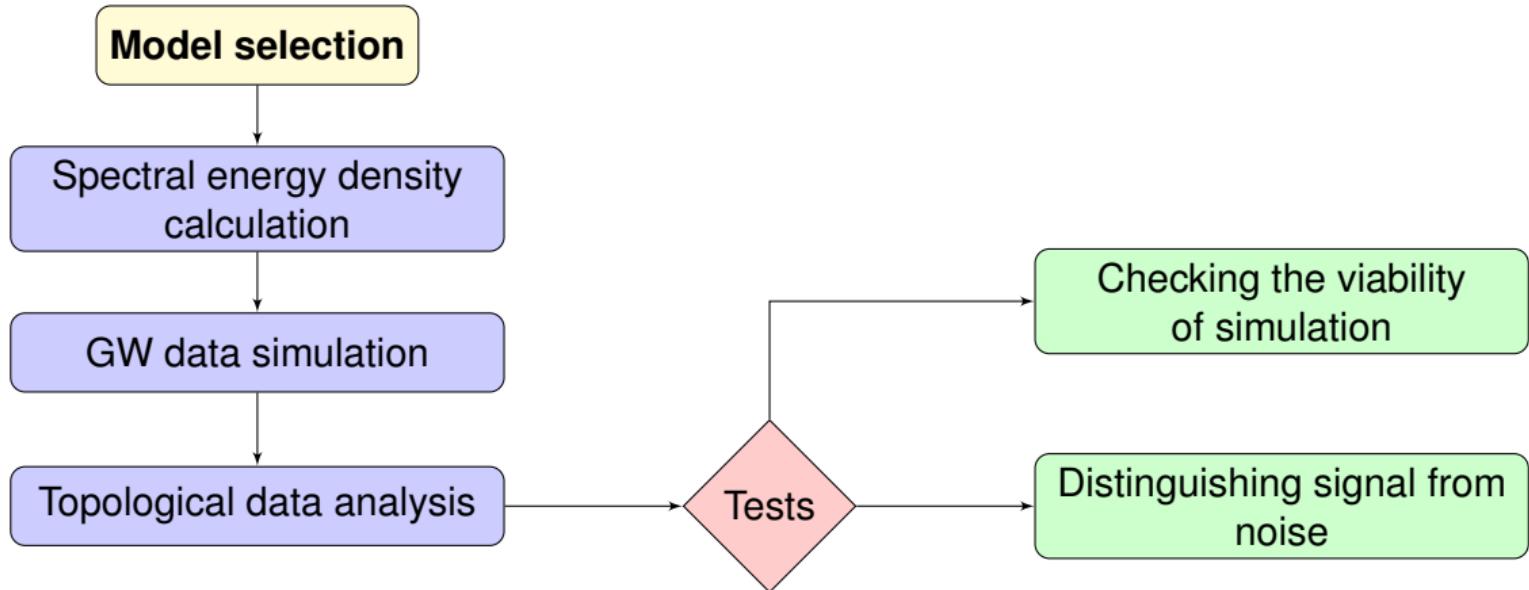
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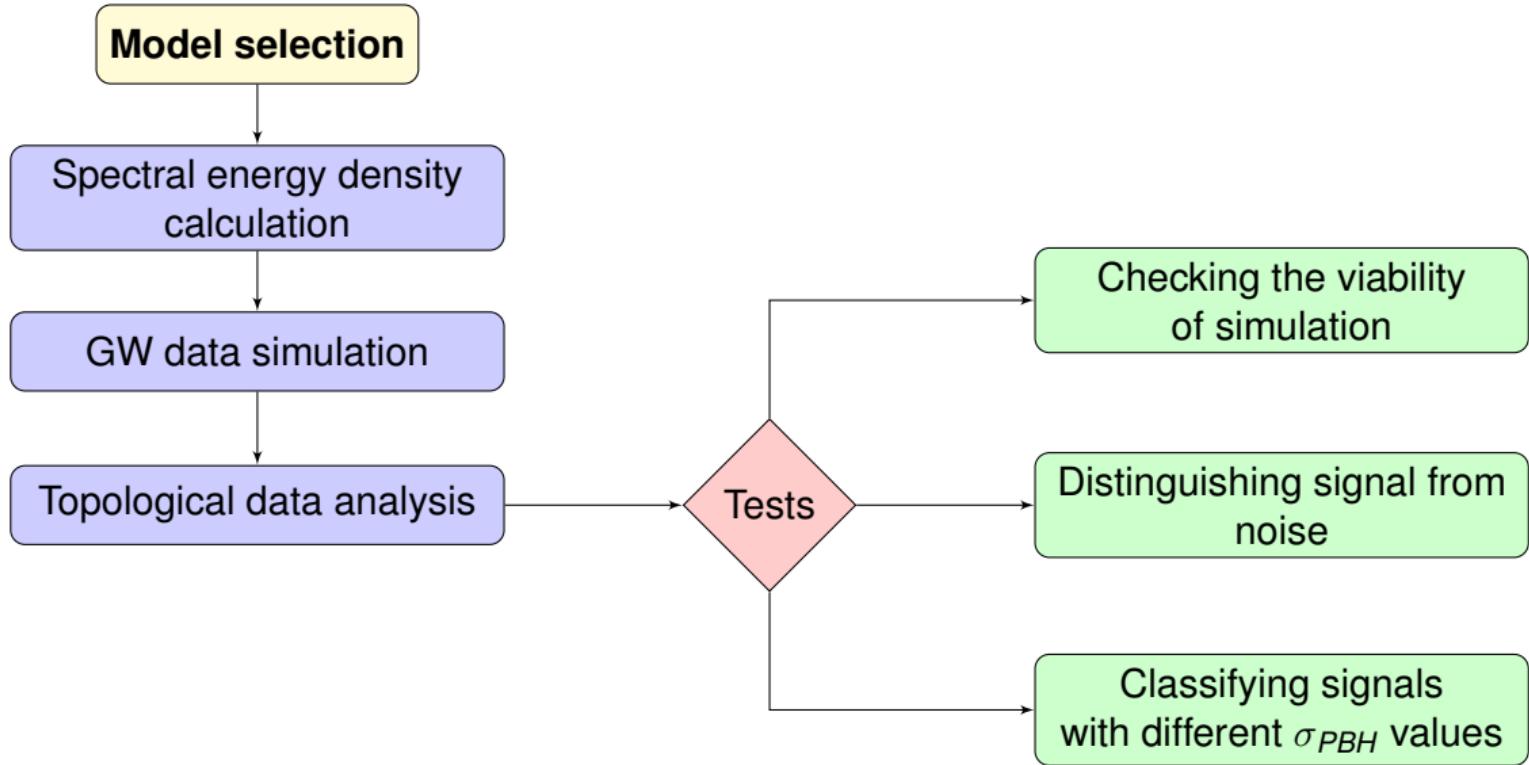
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- ▶ Changing various astrophysical and cosmological parameters to investigate our models ability to constrain those parameters.

# Accessibility

In the end, if you are interested in my research, you can download my thesis and see my contact information on my page at physics department's computational cosmology group website:

[ccg.sbu.ac.ir/people/ali-salehi](http://ccg.sbu.ac.ir/people/ali-salehi)

The background features a dynamic, swirling pattern of blue and black lines that resemble a celestial body like a planet or star. Two bright yellow spheres, possibly representing planets or stars, are positioned within this swirling mass. One sphere is larger and more prominent, while the other is smaller and located slightly below and to the right of the first. The overall composition has a dreamlike, astronomical feel.

Thank you for your attention!