# Physics Advanced Laboratory FOURIER ANALYSIS – GRAVITATIONAL WAVES

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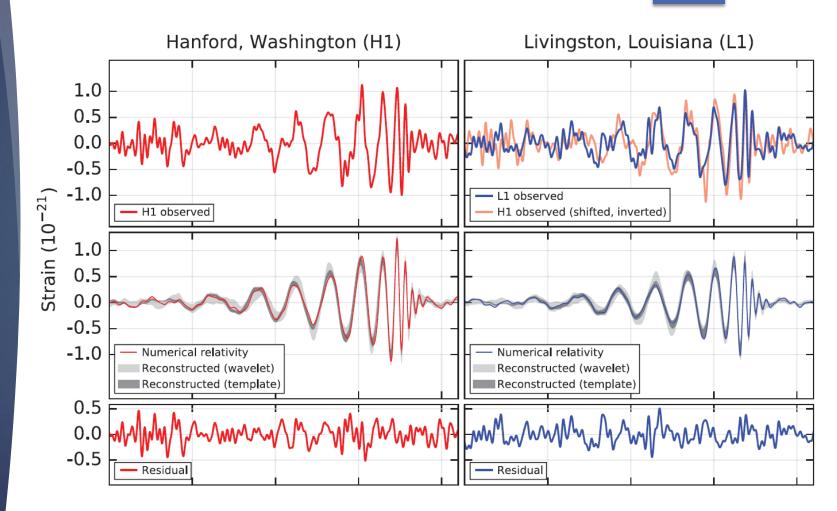
# Outline

- Description of the experiment;
- Fourier Analysis: main tool to analyze data from the experiment;
- Identifying and removing noise;
- Correlation analysis;
- Analysis in a selected window;
- Conclusions.

#### Gravitational Waves

 Disturbances in the spacetime curvature;

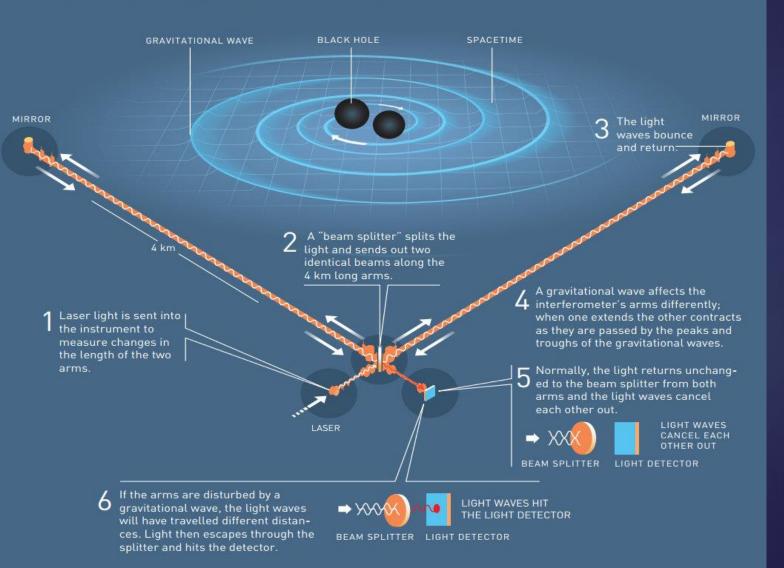
- Generated by the merger of two black holes;
- ► Strain: ~10<sup>-21</sup>



### Gravitational Waves:

#### How can we "see" them?

#### LIGO - A GIGANTIC INTERFEROMETER



## Fourier Analysis

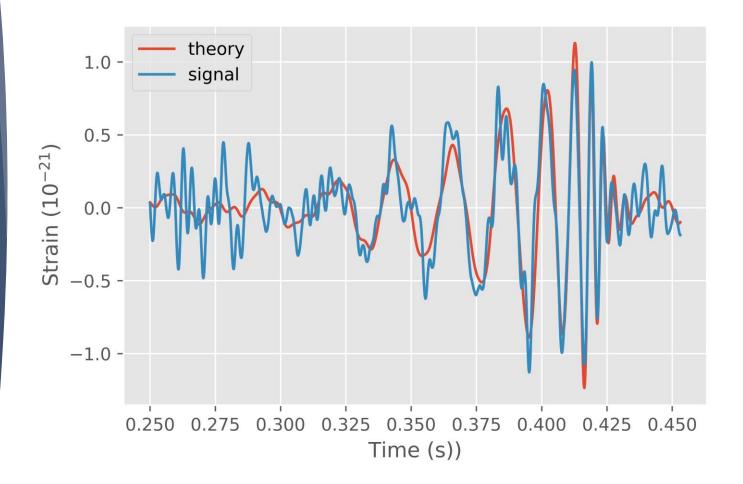
- Tool for decomposing a signal into periodic components (sines and cosines) and recover it from those components.
- Allow us to decompose signal in components characterized by a different choice of frequencies, ex. noise in high frequencies.
- Signal x[n] is discrete, given only in N sampling times n∆t, over a period T: need to use Discrete Fourier Transform

$$x(t) = \sum_{k=-\infty}^{\infty} y_k e^{i\frac{2\pi kt}{T}} \longrightarrow x[n] = \frac{1}{N} \sum_{k=0}^{N-1} e^{2\pi i\frac{kn}{N}} y[k]$$
$$y_k = \frac{1}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} e^{-i\frac{2\pi kt}{T}} x(t) dt \longrightarrow y[k] = \sum_{n=0}^{N-1} e^{-2\pi i\frac{kn}{N}} x[n]$$

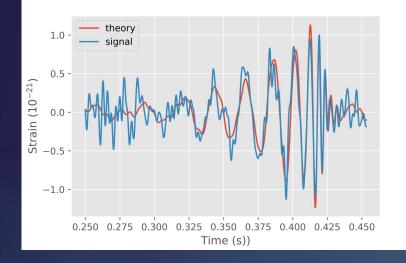
$$i\frac{2\pi}{T}kt \to i\frac{2\pi}{N\Delta t}kn\Delta t \to 2\pi i\frac{kn}{N}$$

#### Gravitational Waves: Theory vs Experiment

- Analysis of the correlation between theory (expected signal) and observation (actual signal)
- Analysis of the correlation between residue and observation

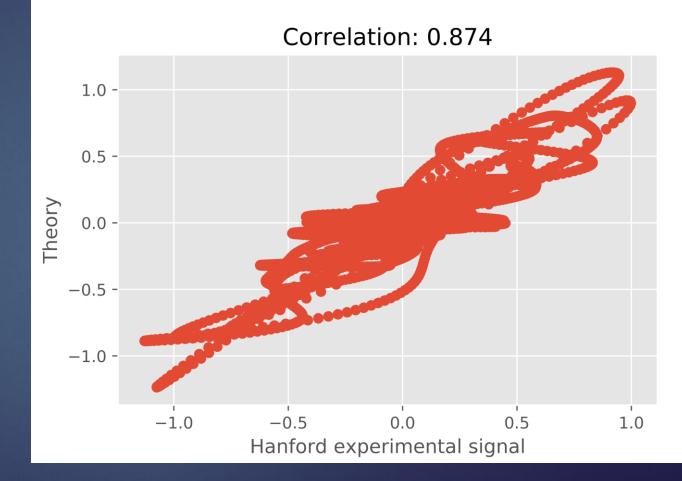


# Gravitational Waves theory vs observation

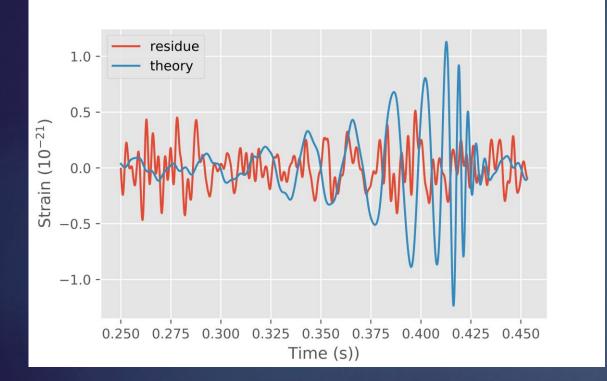


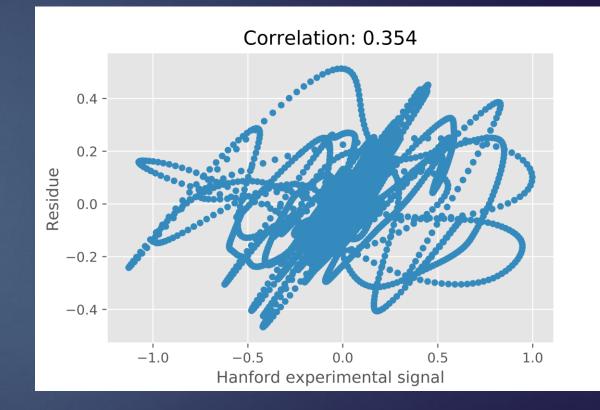
# Pearson correlation coefficient

$$r_{xy} = rac{\sum_{i=1}^n (x_i - ar{x})(y_i - ar{y})}{\sqrt{\sum_{i=1}^n (x_i - ar{x})^2} \sqrt{\sum_{i=1}^n (y_i - ar{y})^2}}$$



#### **Residues:** difference between signal and theory





Residue varies too much in amplitude (contains more than noise)

Some correlation between residue and signal: where signal is not expected, actual signal could pure noise (ex: high frequencies)

## Fourier Transforms: finding noise (raw windows)

Amplitude Spectral Density residue 250 theory signal 200 -150 -100 -50 -0 1 1 100 200 300 0 400 500 freqs (Hz)

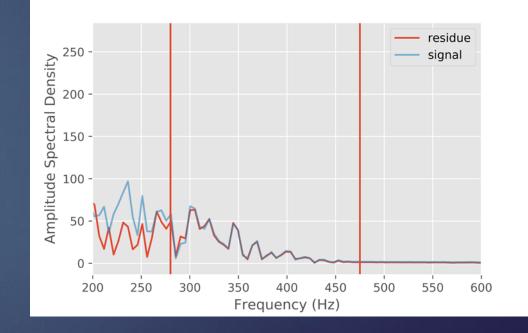
residue Amplitude Spectral Density 100 -100 -200 -100 -100 -200 -10 signal 0 200 250 300 350 450 500 550 600 400 residue 250 signal 200 -150 -100 . 50 -0 20 40 60 80 100 0

Frequency (Hz)

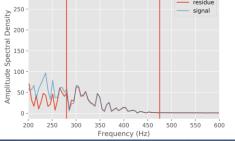
# Optimizing the extraction of noise

#### Procedure

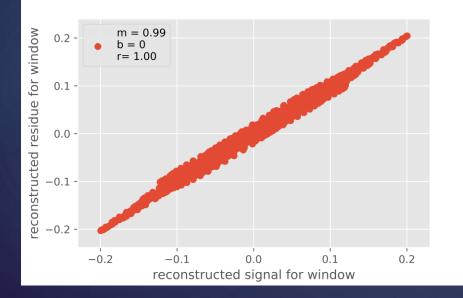
- Departing from crude estimates for the limits of frequency window, search methodically for values that give the best fit for Residue = Signal
- For the cases where the fit is almost perfect (slope > 0.99; intercept ~ 0), find the larger window.



# Optimizing the extraction of noise results



High frequencies window : 286 Hz - 398 Hz



not filtered, r = 0.874filtered, r = 0.8911.0 1.0 0.5 0.5 theory 0.0 0.0 -0.5-0.5 -1.0 -1.0-1.0-0.5 0.0 0.5 1.0 -1.0-0.5 0.5 1.0 0.0 raw observation observation - noise

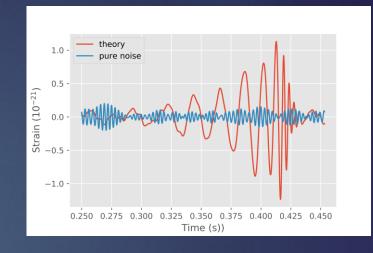
0.874

11

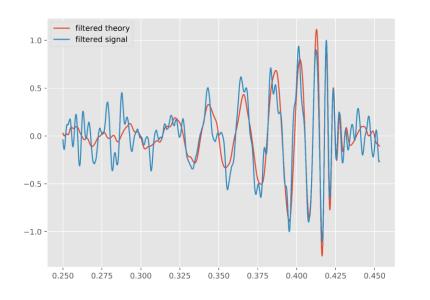
0.891

# Before and After

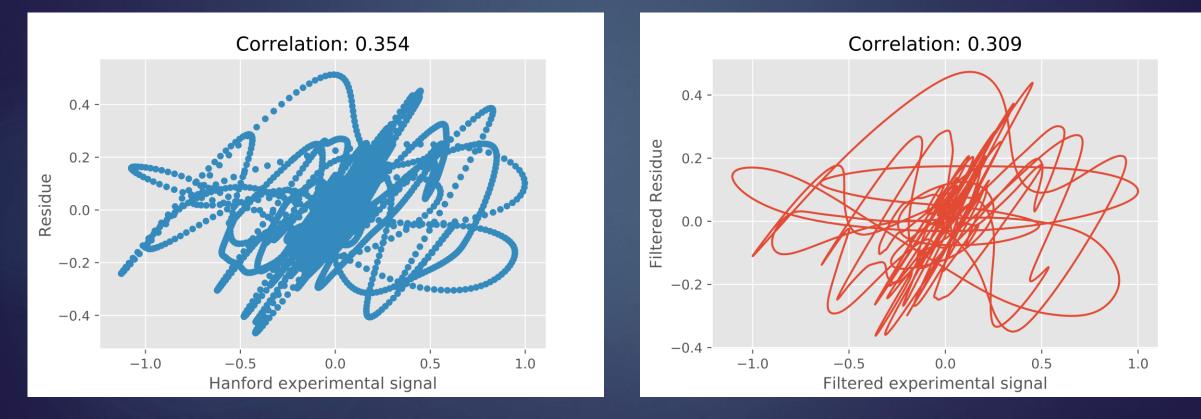
Mostly attenuation of spiky behavior in the signal



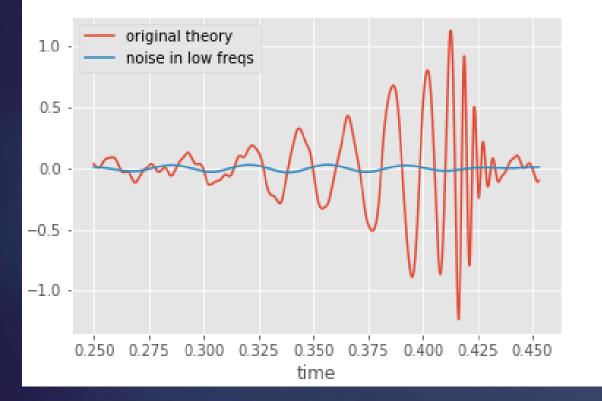
theory raw signal 1.0 0.5 0.0 -0.5-1.00.400 0.425 0.450 0.250 0.275 0.300 0.325 0.350 0.375



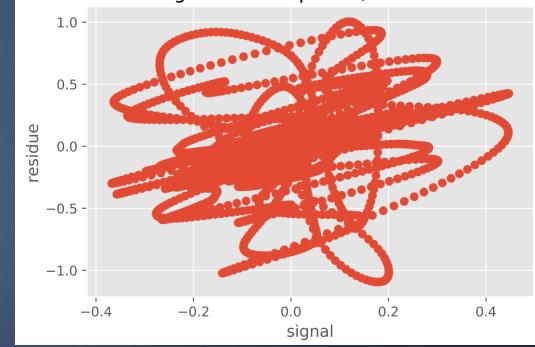
## Residue vs Signal Before and after removing high frequency noise



## Low frequencies noise correlation residue/signal actually increases

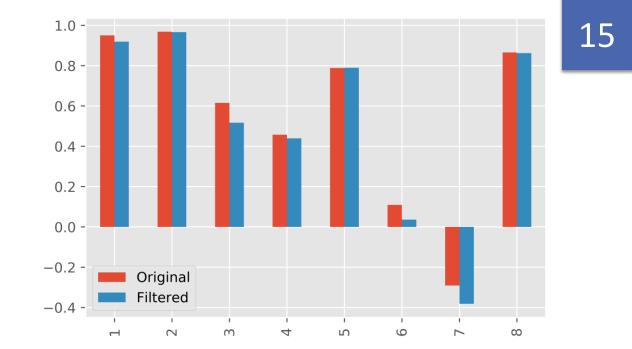


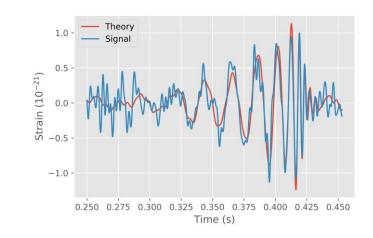
high + low freq. filter, r = 0.310



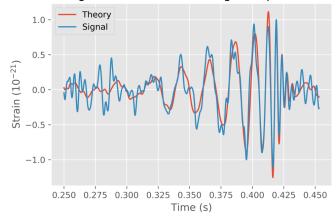
#### Correlation Analysis

- 1, 2 & 8 low amplitudes, theory is almost zero
- ► 3 & 4 correlation decreases
- $\blacktriangleright$  5 high correlation
- 7 negative correlation;
  filtered has bigger correlation





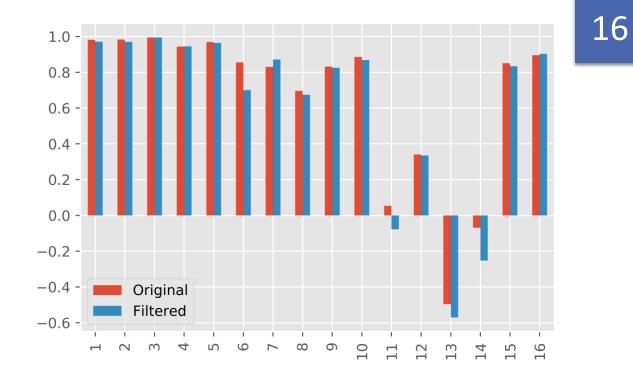
#### Signal Without Noise in High Frequencies

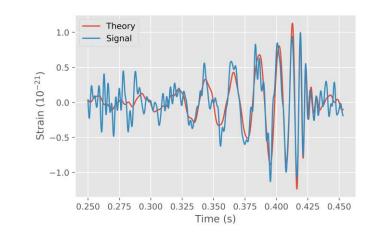


#### Correlation Analysis

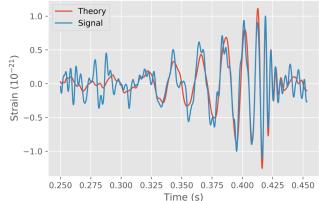
- Sometimes filtered data has bigger correlation than original data;
- Different behaviors for different windows;

We could analyze the data in different windows...

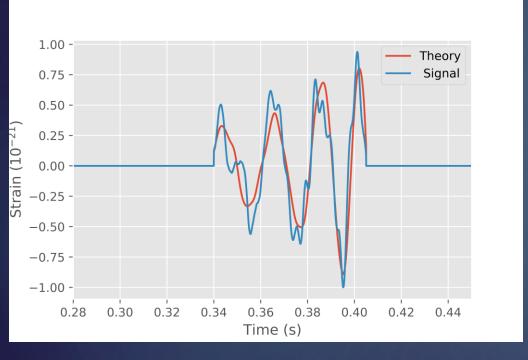


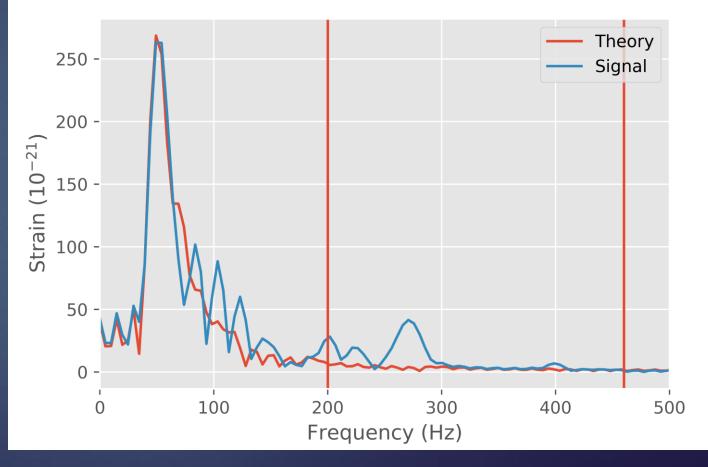


#### Signal Without Noise in High Frequencies

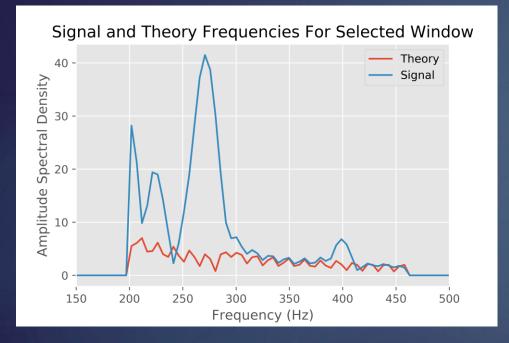


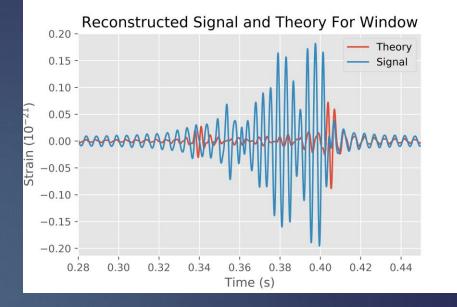
# Inspiral Fase Analysis

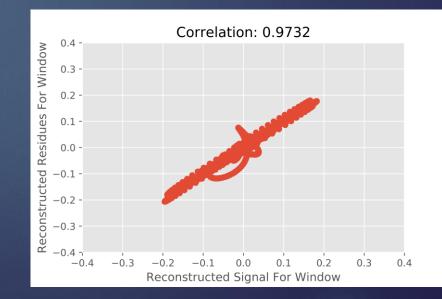




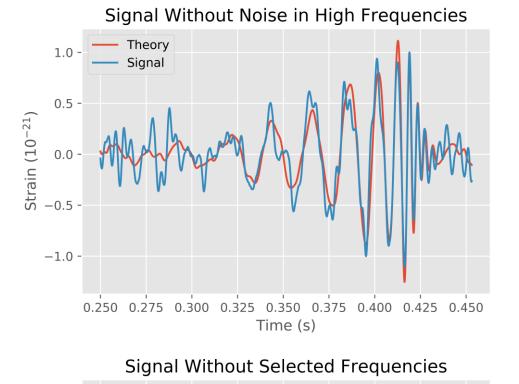
# More Noise?

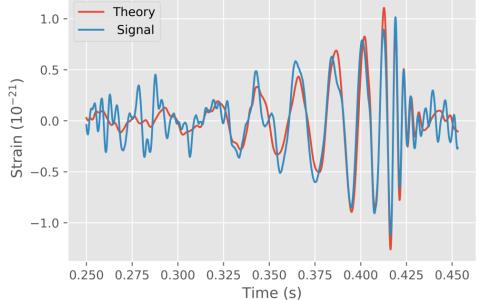




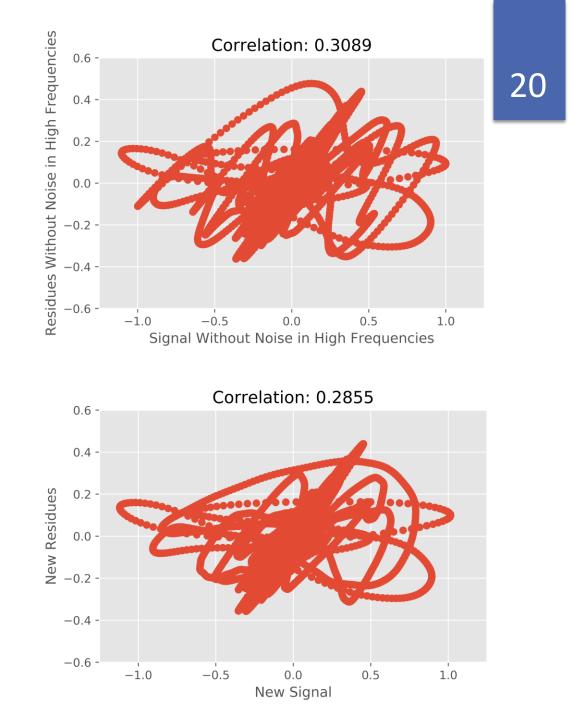


#### Data Evolution

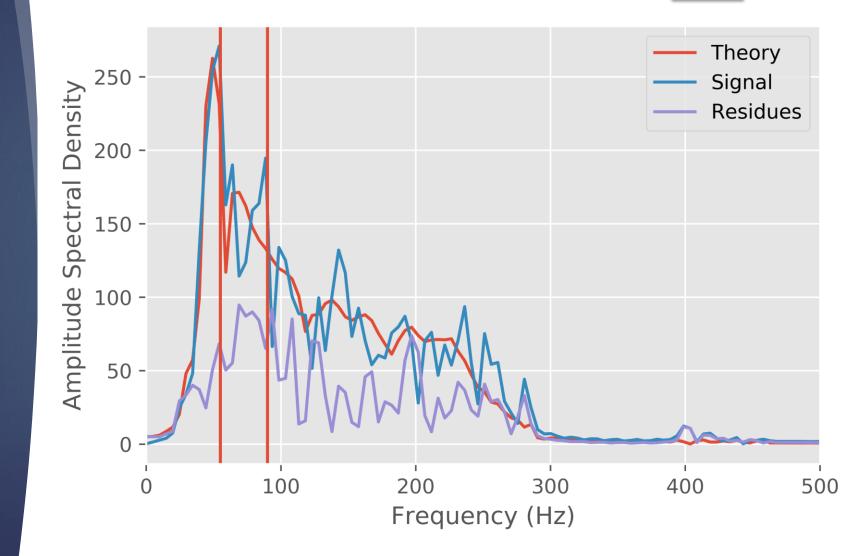




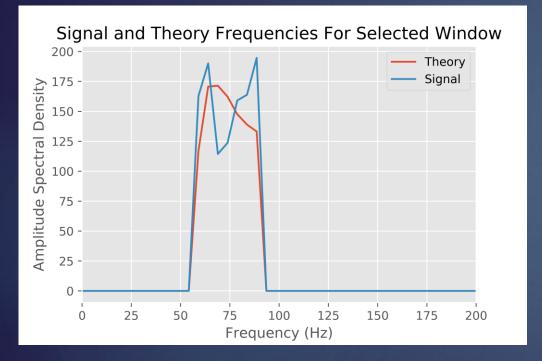
# Correlation Evolution

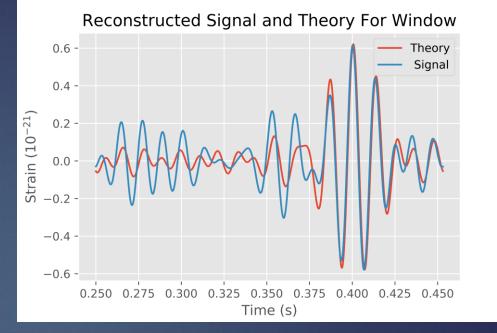


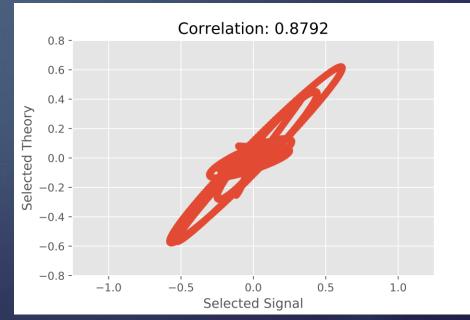
#### A New look at the Frequency Domain



# High Residues



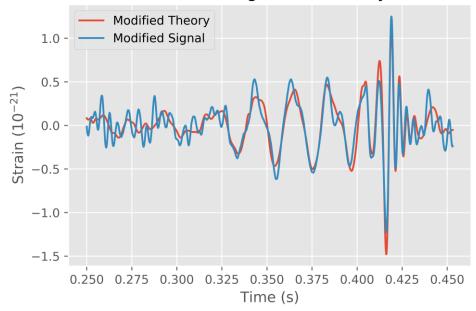




#### Data Evolution

Theory 1.0 Signal 0.5 Strain (10<sup>-21</sup>) -1.0 -0.250 0.275 0.300 0.325 0.350 0.375 0.400 0.425 0.450 Time (s)

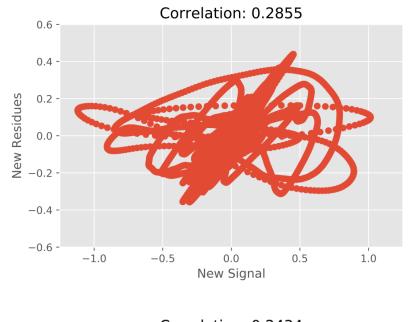
Modified Signal and Theory

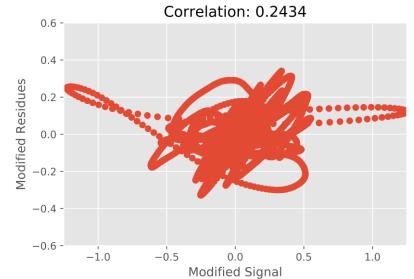


Signal Without Selected Frequencies

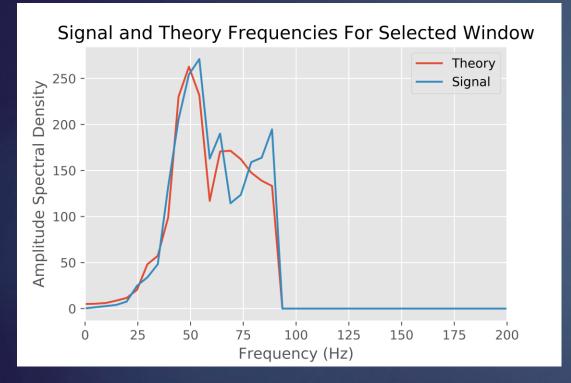
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#### Correlation Evolution

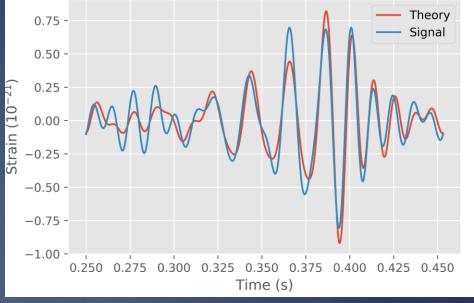


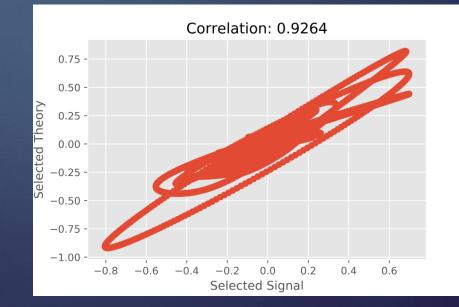


# Lower Frequencies

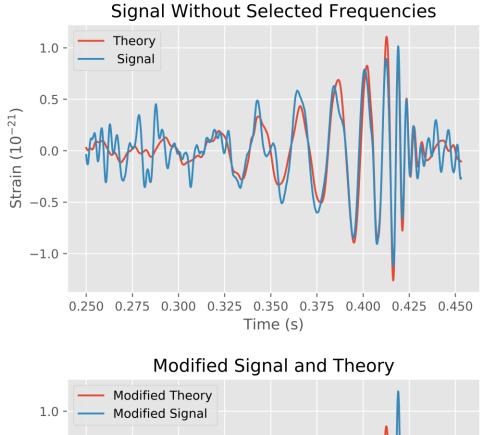


Reconstructed Signal and Theory For Window





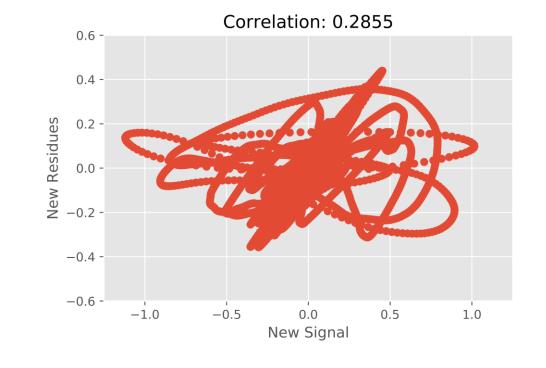
#### Data Evolution



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#### 1.0 - Modified Theory 1.0 - Modified Signal 0.5 -0.0 - MMMMMMMMMMMMMM -0.5 --1.0 -0.250 0.275 0.300 0.325 0.350 0.375 0.400 0.425 0.450 Time (s)

#### Correlation Evolution





# Conclusions

▶ We were able to find a window of noise in the high frequency region.

In low frequencies, we didn't consider any noise because what we isolated was weak and by using that filter the correlation between residue and signal increased.

- Looking at the correlation plots we see that, by clearing the noise, not all zones got better correlations, some even got a little worse.
- This can happen because this noise has some modulation and does not behave the same way along the signal.
- By looking at different regions of the signal, we can find new noise or regions were our model seems to not describe the signal very well.