



Matched Filtering Technique

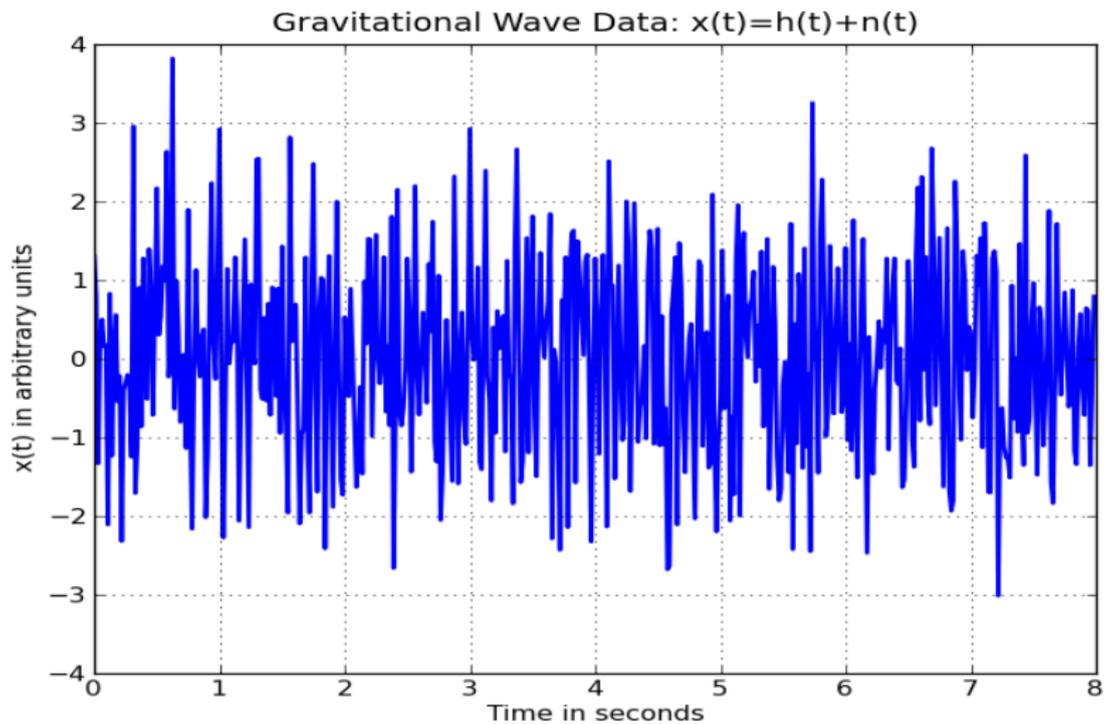
Archana Pai

Indian Institute of Science Education and Research
Trivandrum

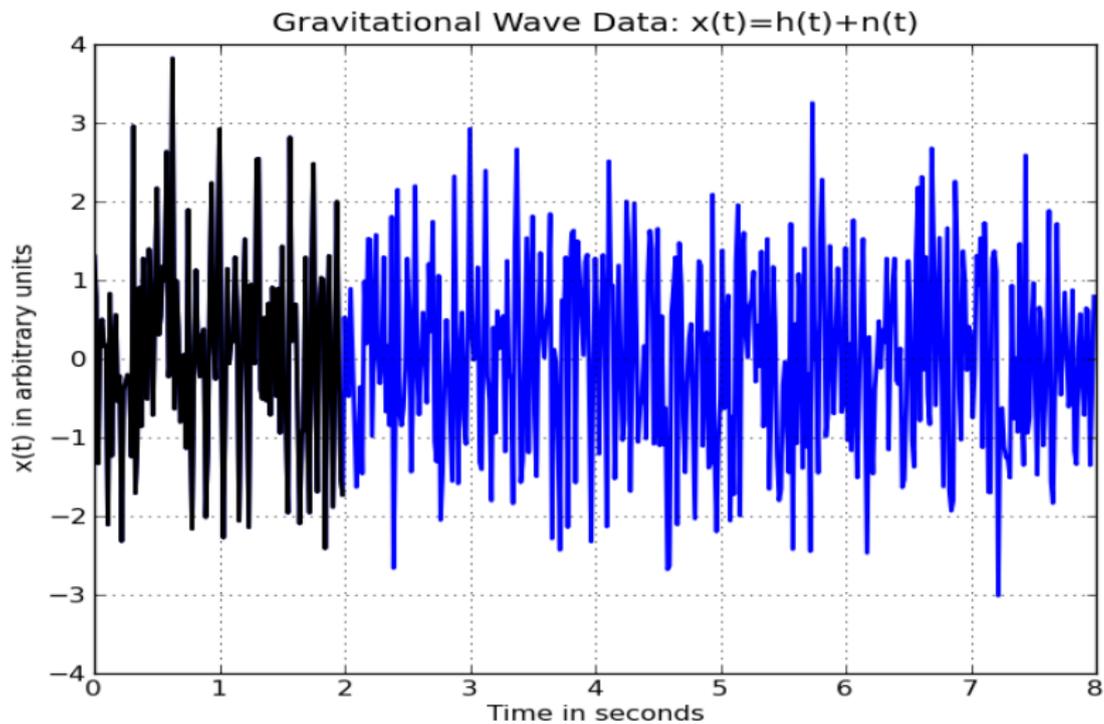


ISGWA2010, 16 December 2010

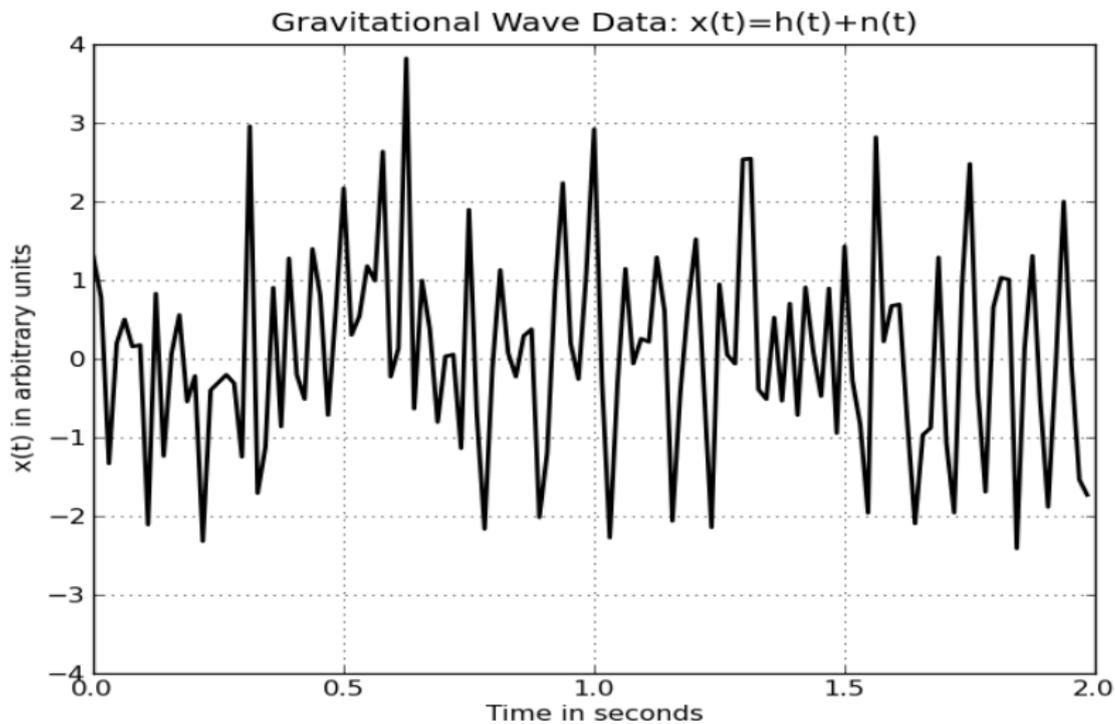
Linear Filtering of Data



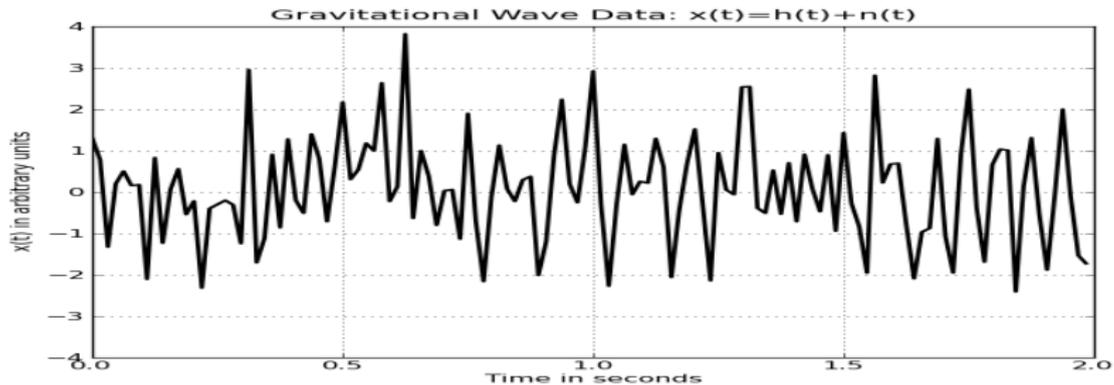
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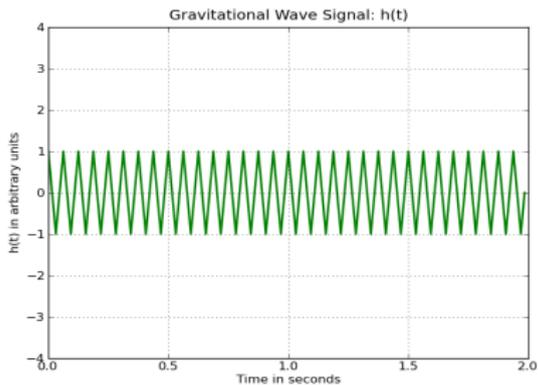
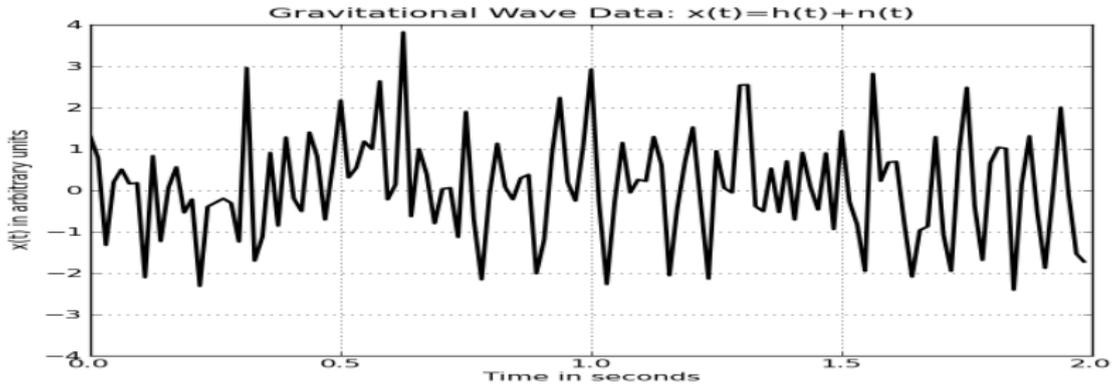
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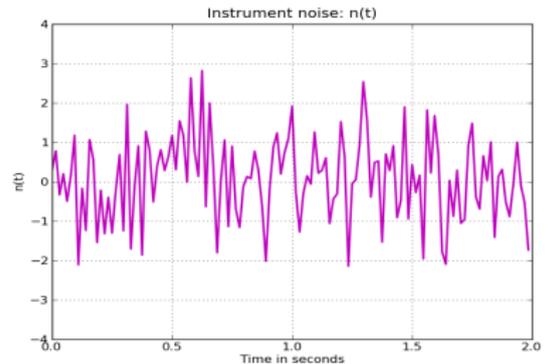
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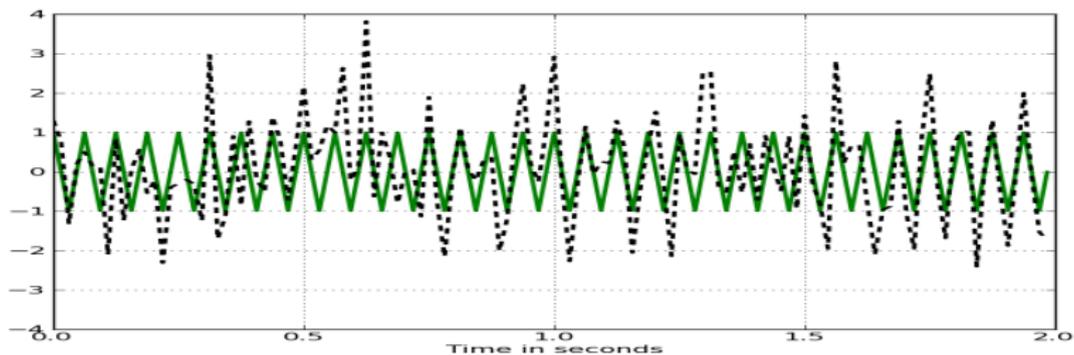
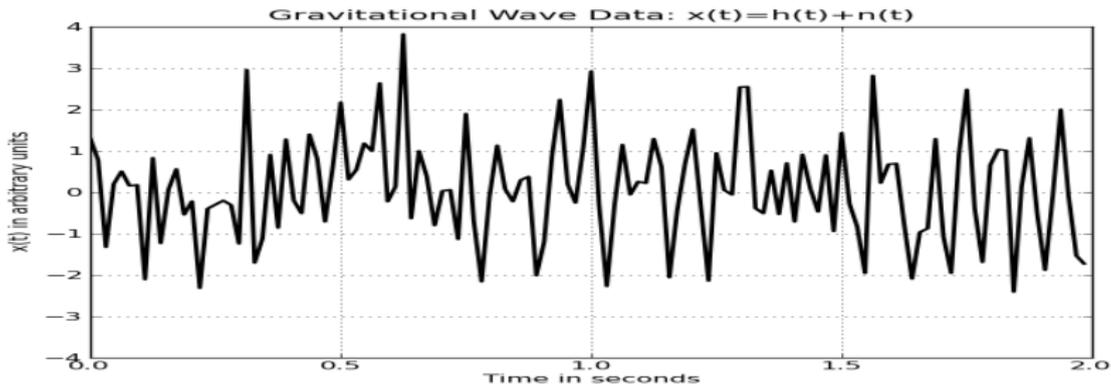
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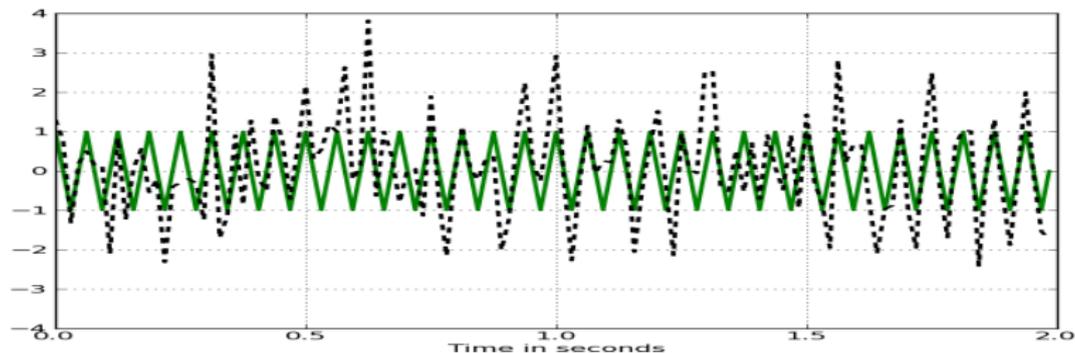
+



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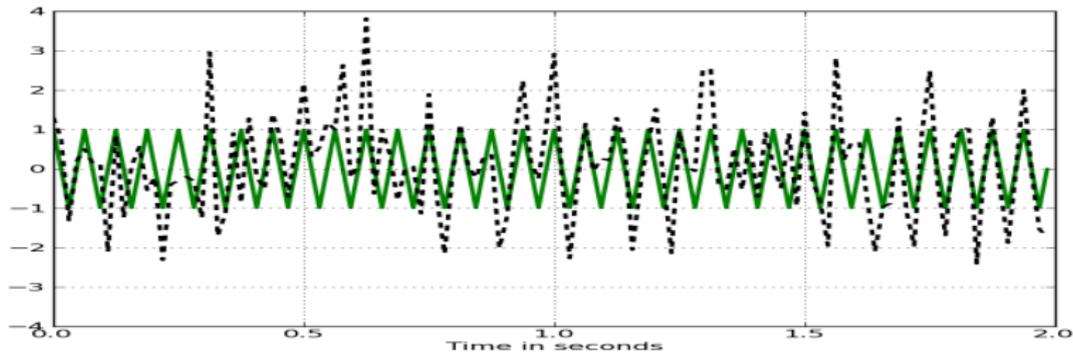
Linear Filtering of Data



Questions to address:

Is there any Gravity-Wave signal in the data?

Linear Filtering of Data



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Is there any Gravity-Wave signal in the data?

What kind of source and its astrophysical parameters?

Linear Filtering of Data

Conv of $x(t)$ and $q(t)$: $X \otimes Q$

$$y(T) = \int x(t) q(T-t) dt$$



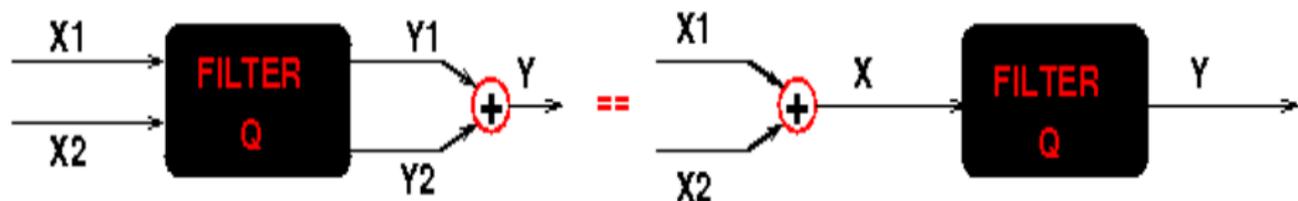
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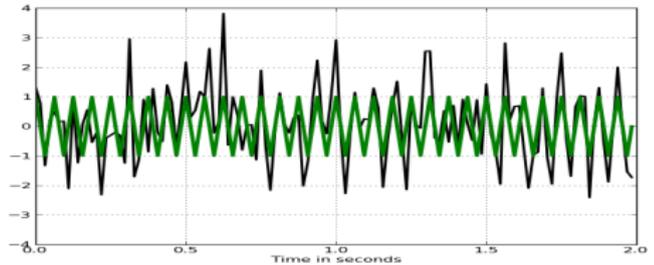
Linearity of a filter:



Linear Filtering of Data

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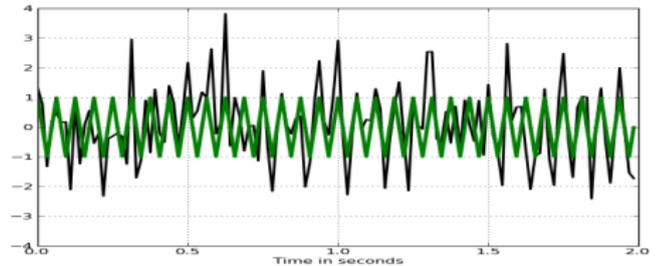
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Linear Filtering of Data

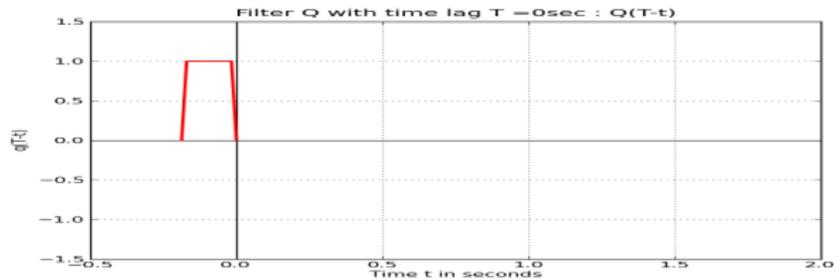
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Filter function

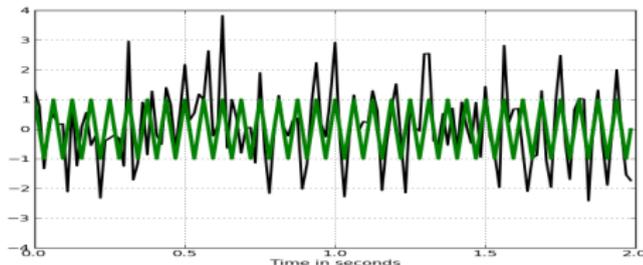
$$q(T - t)$$



Linear Filtering of Data

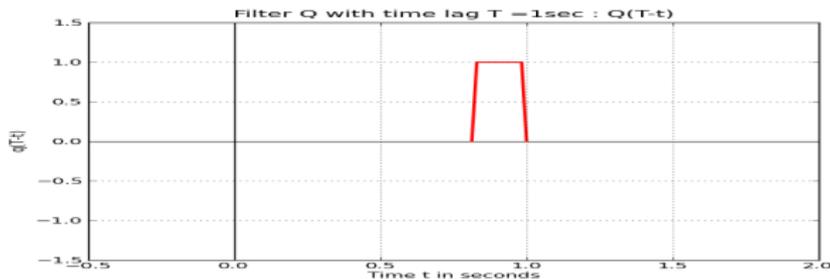
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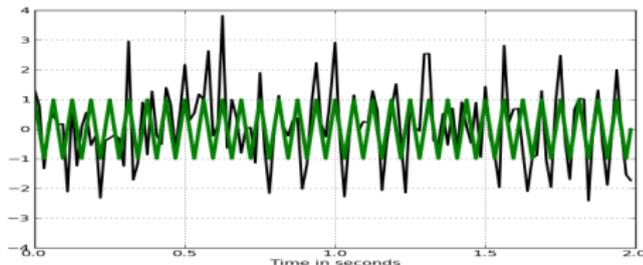
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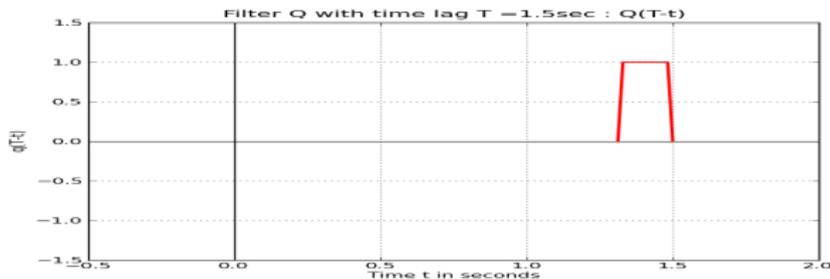
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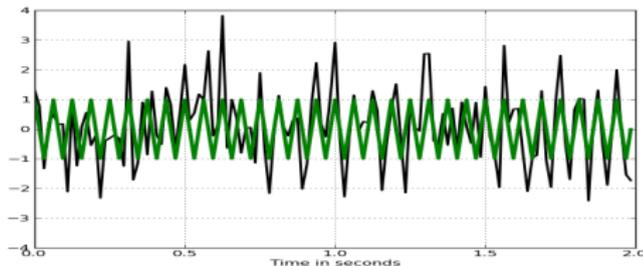
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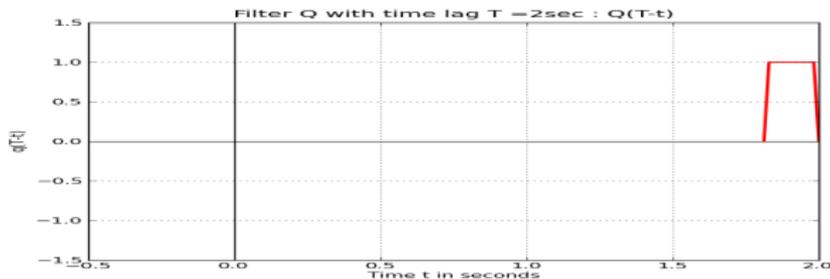
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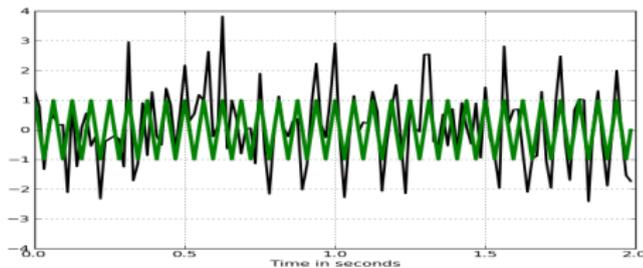
$$q(T - t)$$



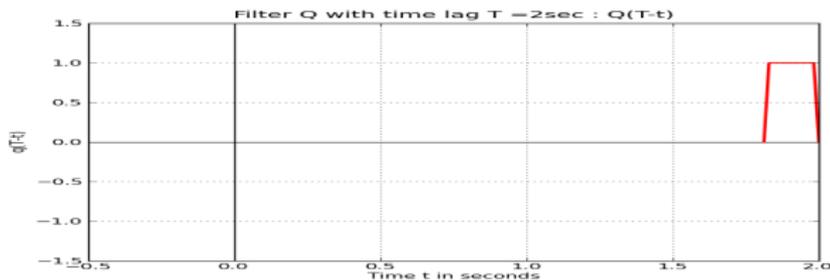
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Filter function
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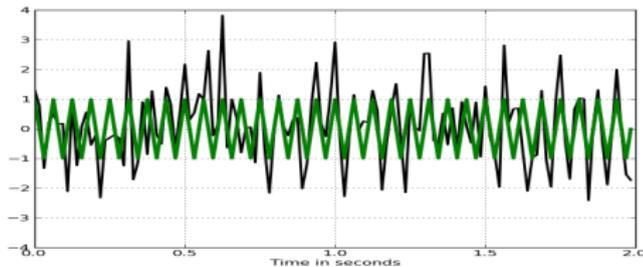
Filter Output $y(T)$



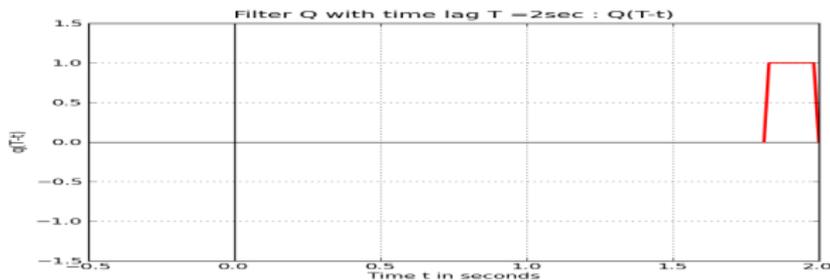
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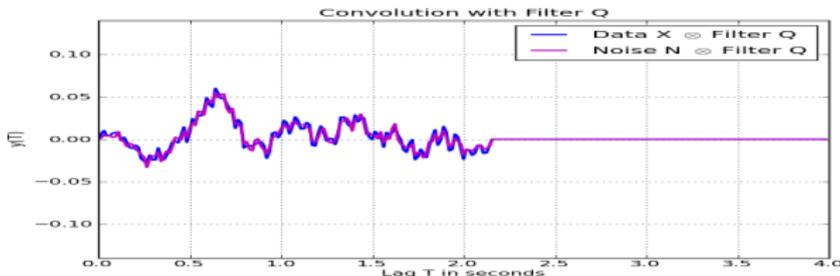
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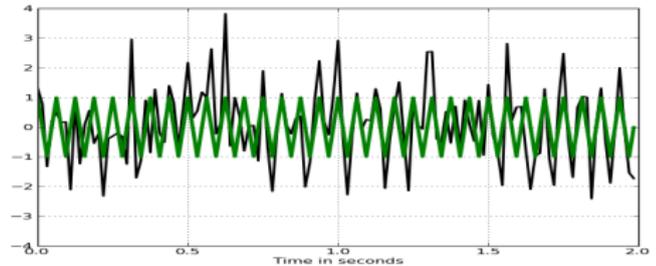
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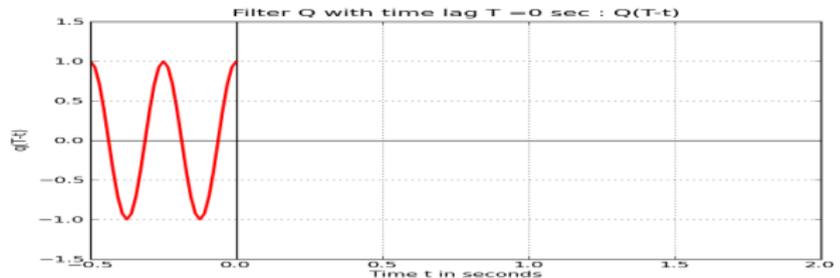
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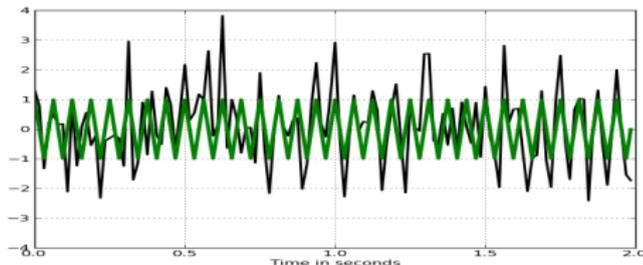
Cos Filter $q(T - t)$
Freq = 4Hz



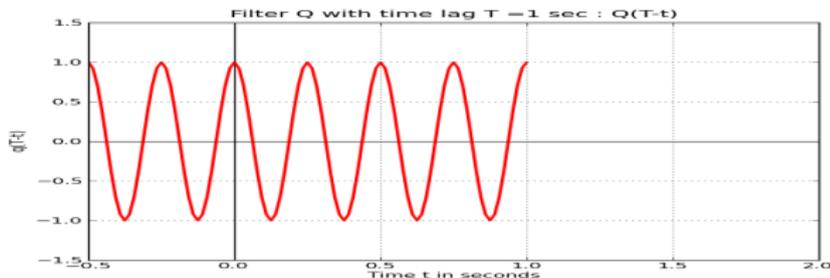
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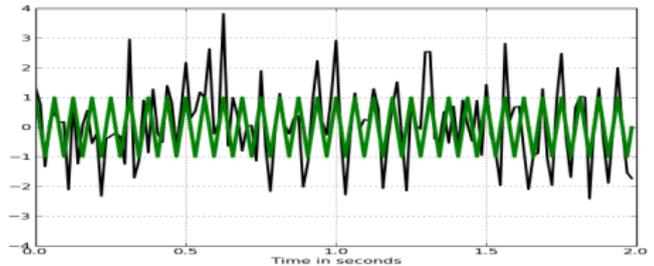
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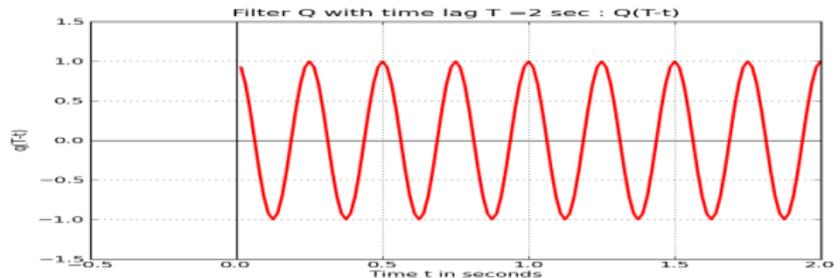
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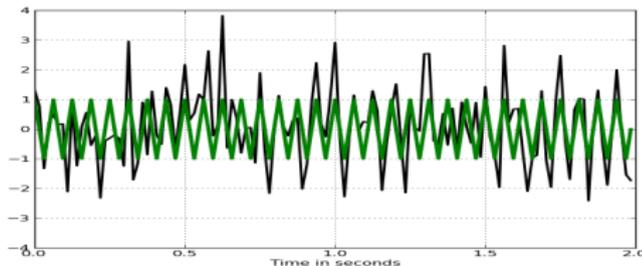
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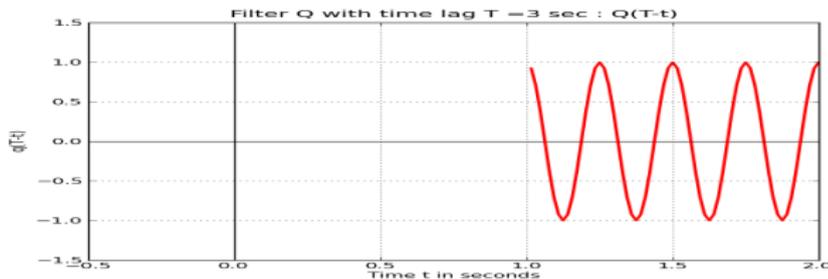
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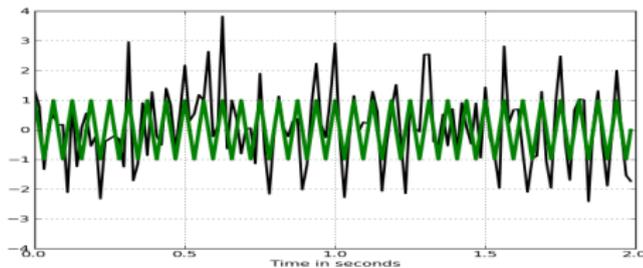
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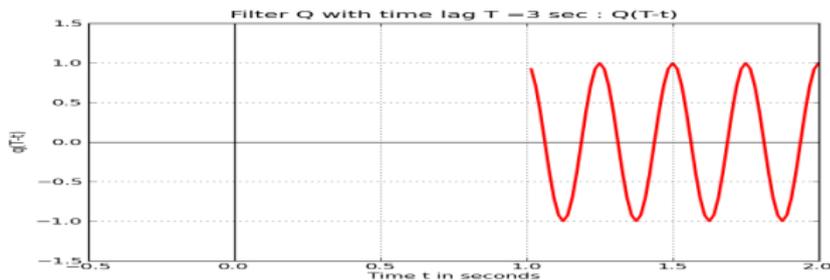
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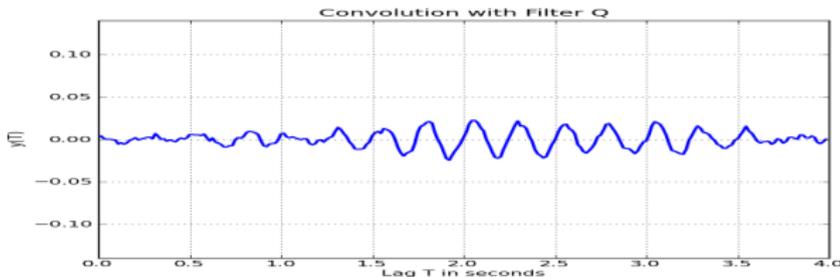
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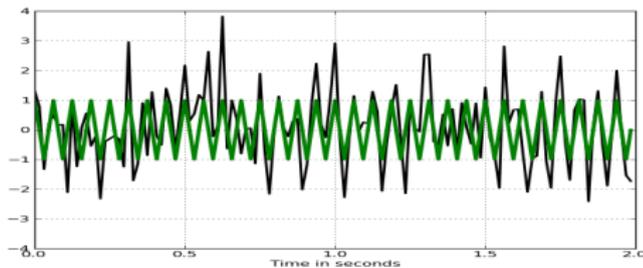
Filtered Output
 $y(T)$



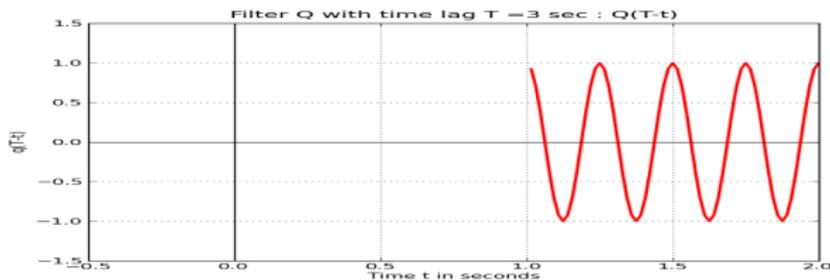
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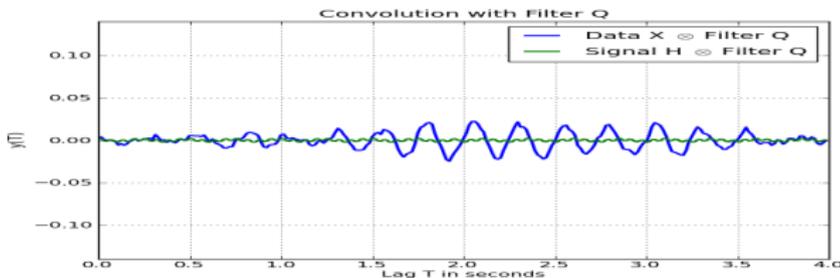
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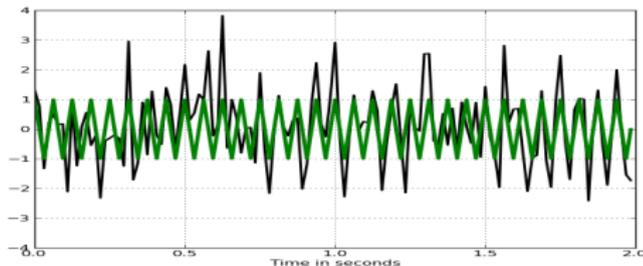
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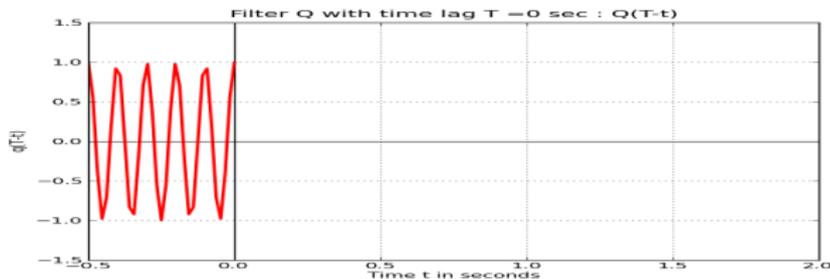
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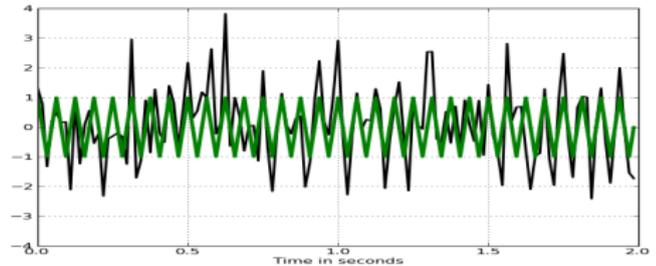
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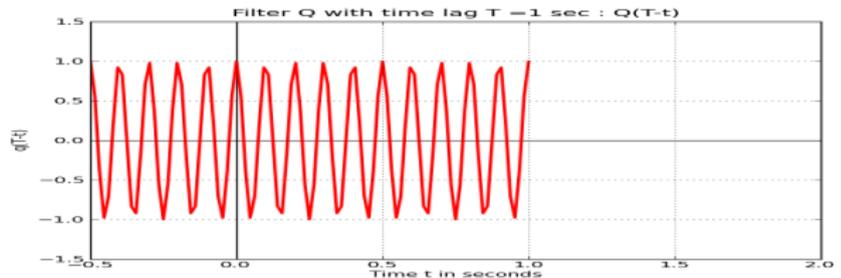
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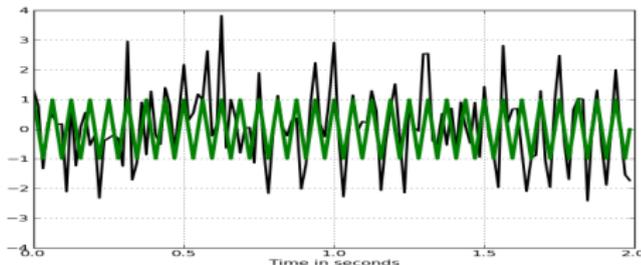
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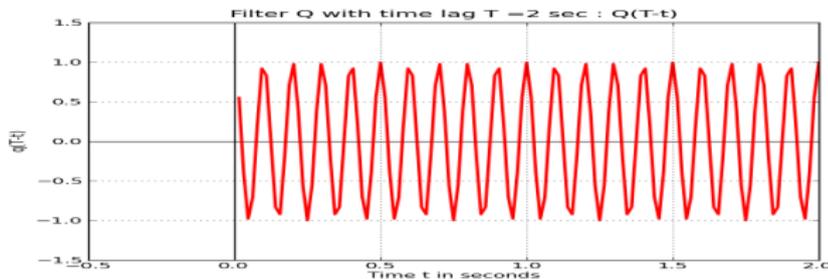
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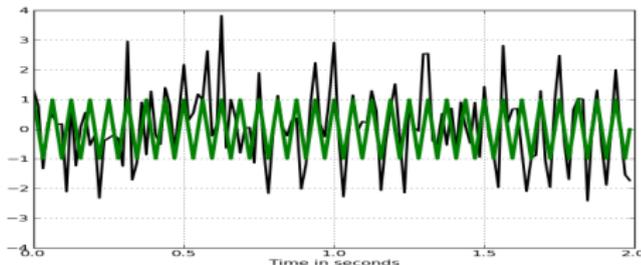
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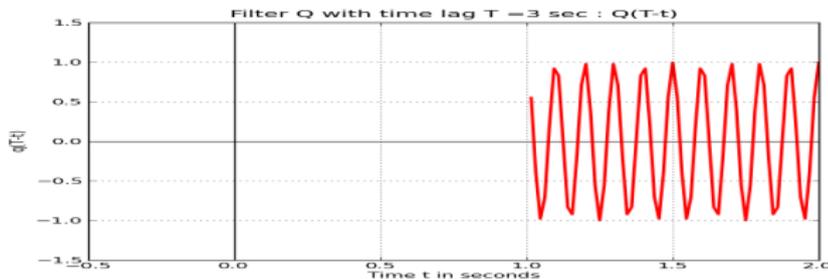
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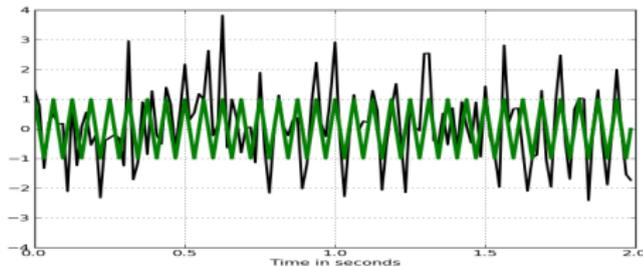
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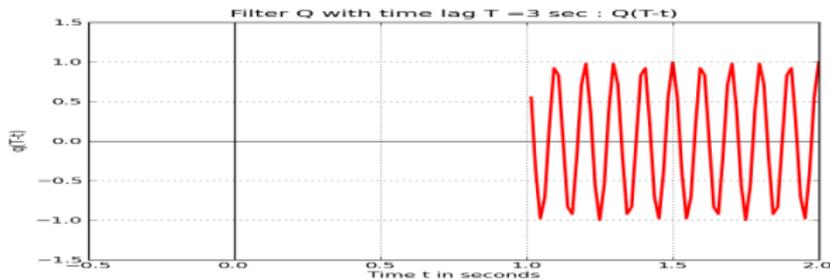
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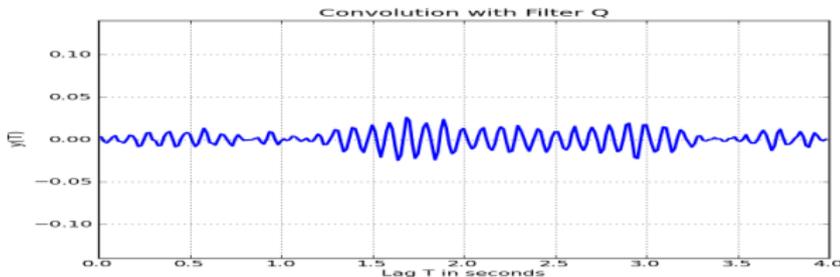
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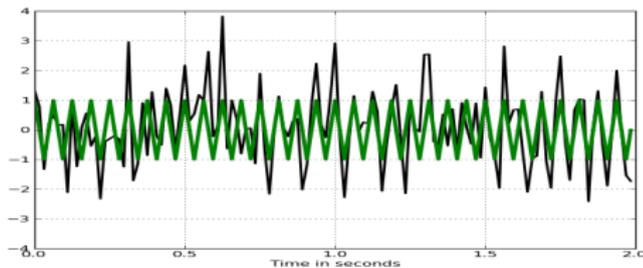
Filtered Output
 $y(T)$



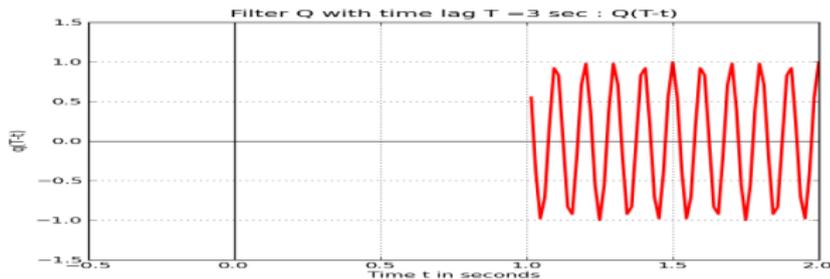
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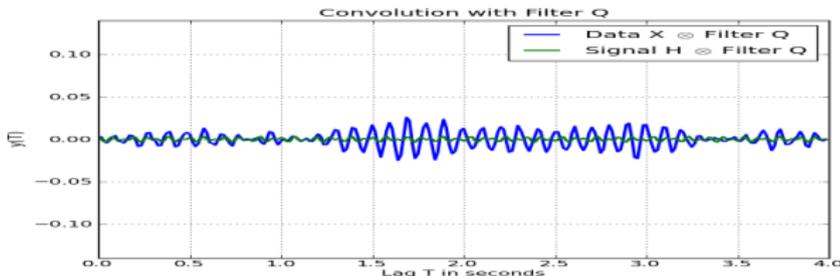
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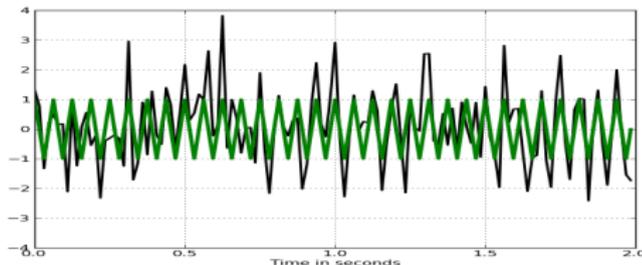
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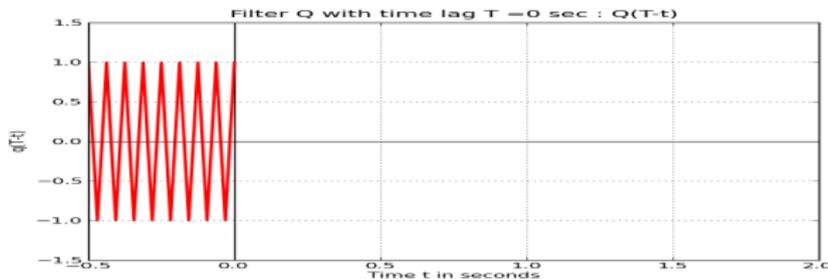
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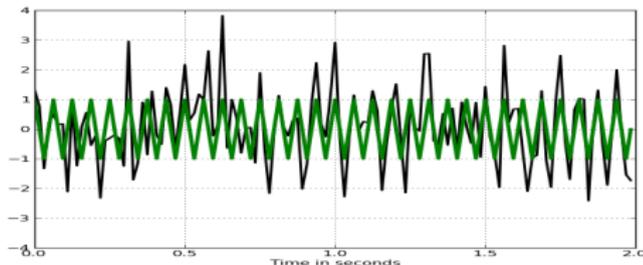
Cos Filter $q(T - t)$
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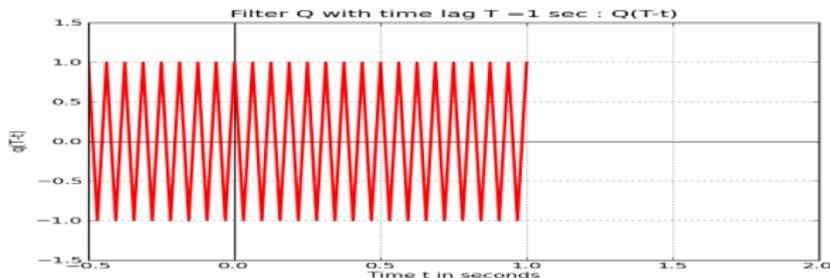
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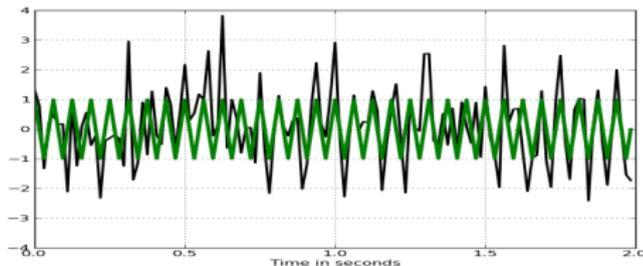
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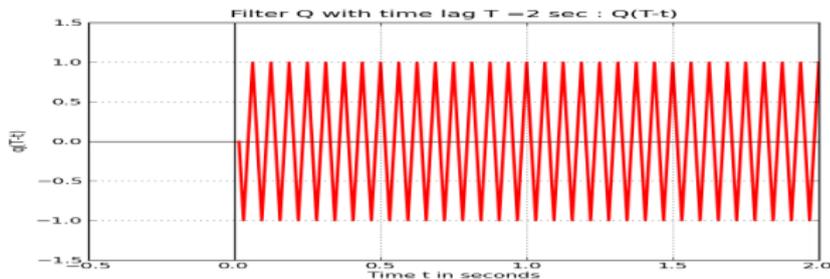
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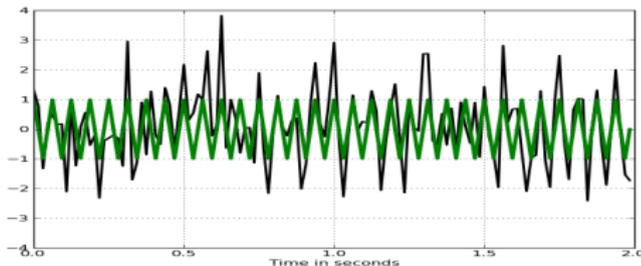
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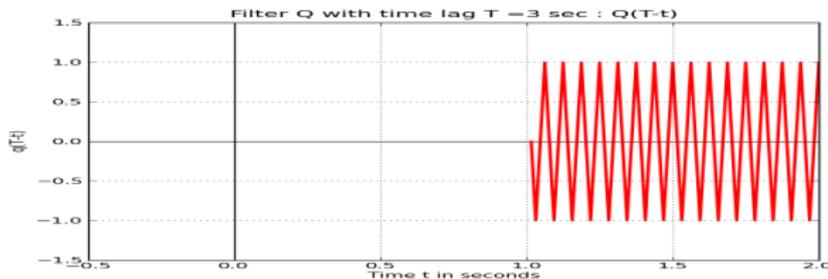
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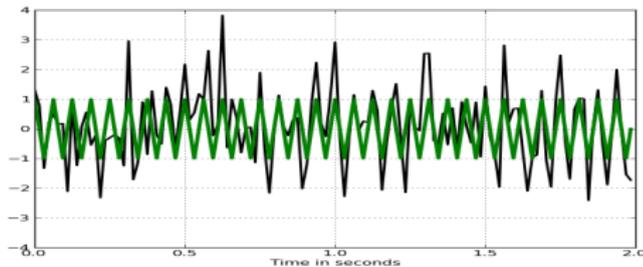
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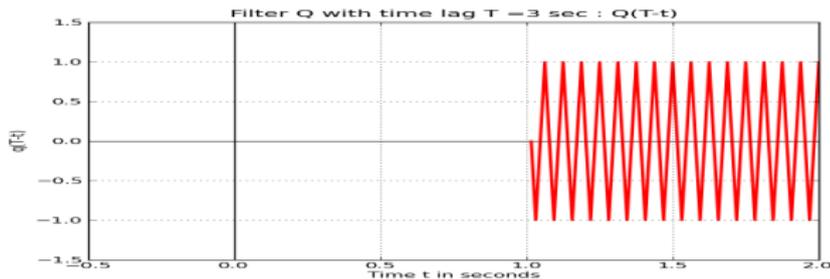
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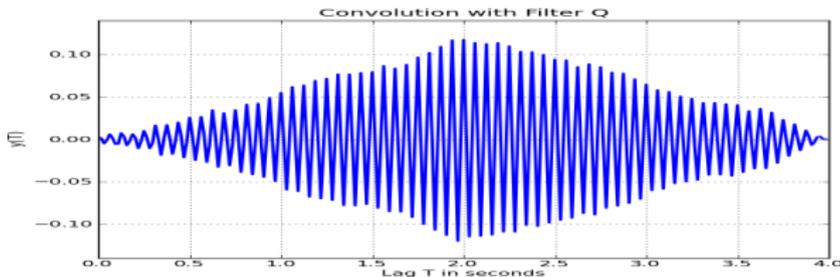
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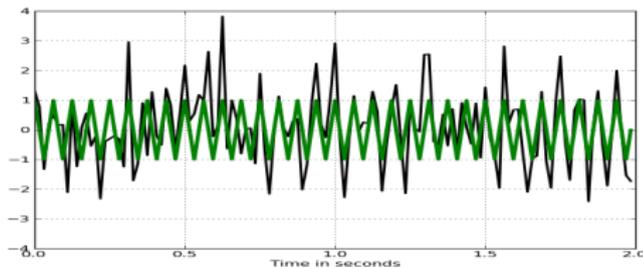
Filtered Output
 $y(T)$



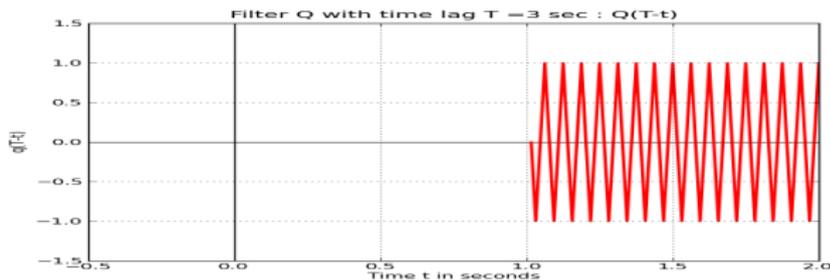
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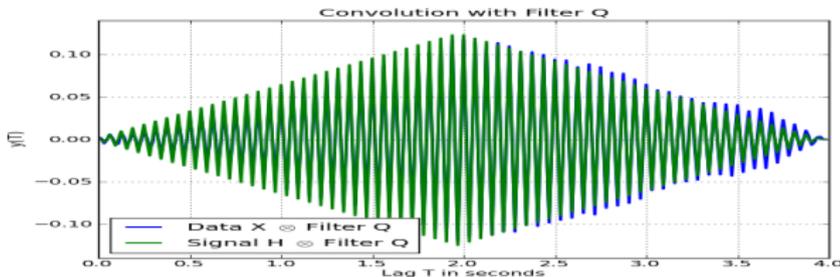
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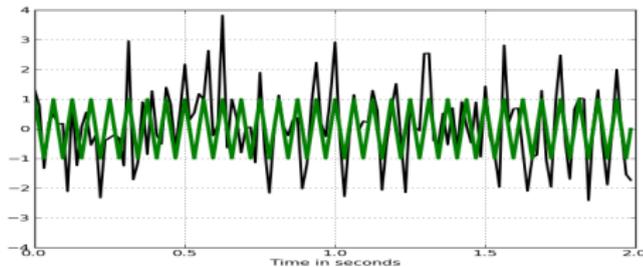
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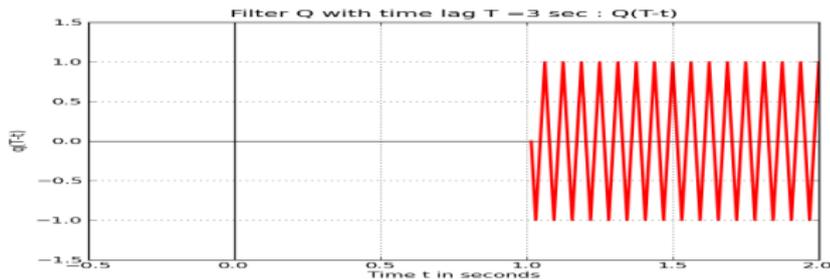
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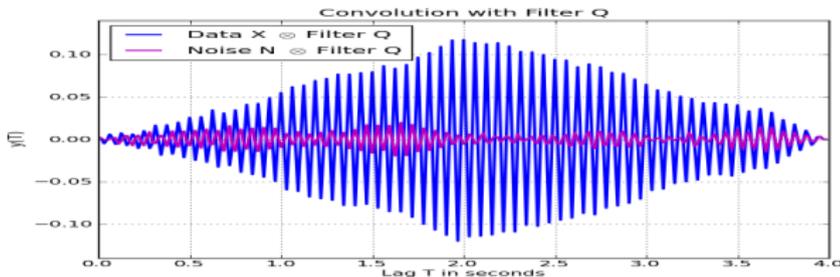
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Filtered Output
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Important questions in GW data analysis

Assume: GW signal is hidden in the noise

Important questions in GW data analysis

Assume: GW signal is hidden in the noise

Take home lesson –

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Best filter depends on signal shape

Important questions in GW data analysis

Assume: GW signal is hidden in the noise

Take home lesson –

Best filter depends on signal shape

Phase matching is crucial for signal detection

Important questions in GW data analysis

Assume: GW signal is hidden in the noise

Several Questions

Important questions in GW data analysis

Assume: GW signal is hidden in the noise

Several Questions

GW signal – Known/Unknown shape?

Important questions in GW data analysis

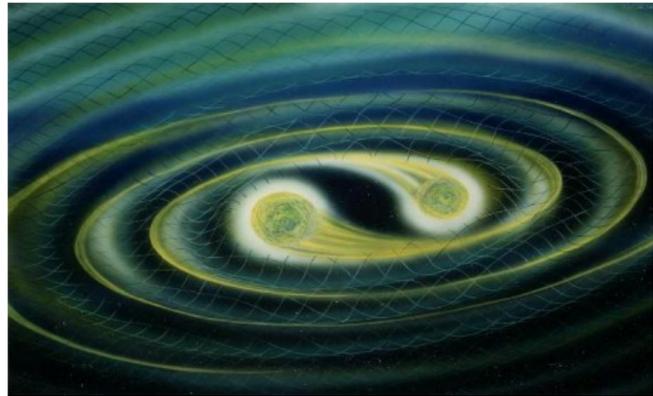
Assume: GW signal is hidden in the noise

Several Questions

GW signal – Known/Unknown shape?

Known signal

- Compact Binary Stars
- Neutron Star – Black Hole
- Neutron Star – Neutron Star
- Black hole – Black hole



Important questions in GW data analysis

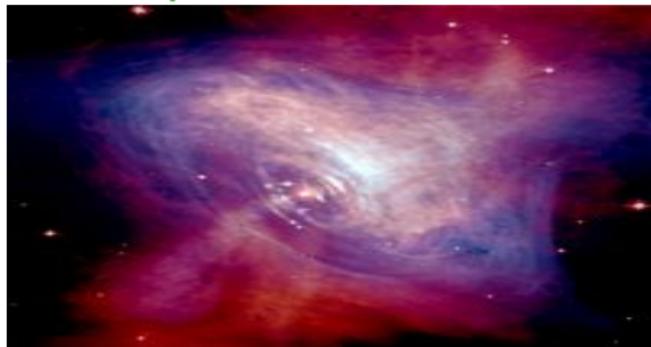
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Pulsars – Pulsating Neutron Stars



Hester et al, 2003; CXC, HST, NASA

Important questions in GW data analysis

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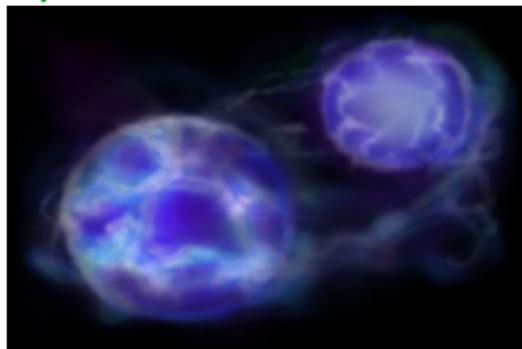
Several Questions

GW signal – Known/Unknown shape?

Unknown/Unmodeled signal

Supernova event *[DA IV]*

Accreting systems *[DA IV]*



Important questions in GW data analysis

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Signal of known shape

Important questions in GW data analysis

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What is the best filter in presence of noise? *[DA III]*

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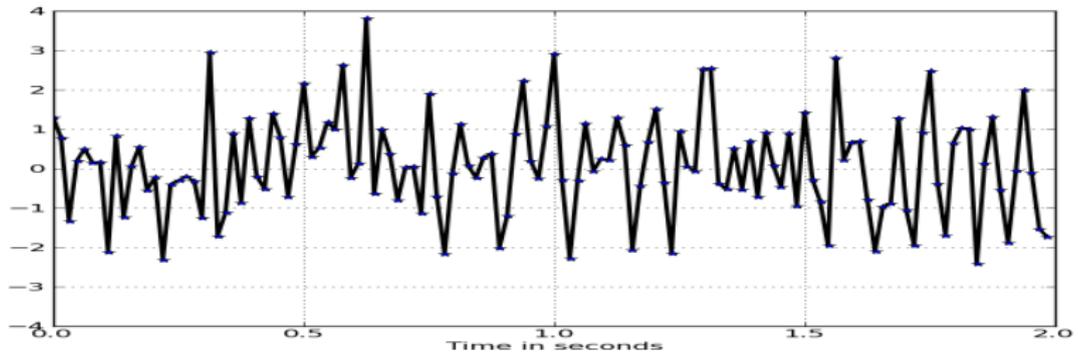
What is the best filter in presence of noise? [DA III]

Parameters affect the signal phase? One/Many ?

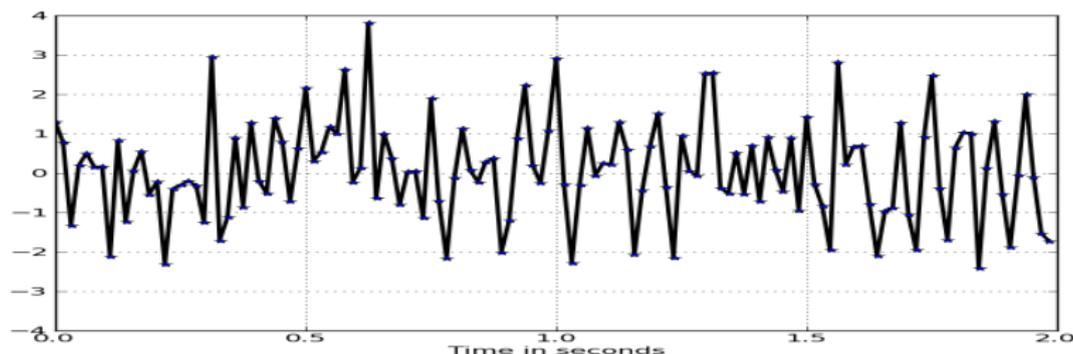
How filters should be spaced?

Phase mismatch between filter and signal? — [DA V]

Revise: Continuous and Discrete Fourier Transform



Revise: Continuous and Discrete Fourier Transform



Continuous Domain

Time series:

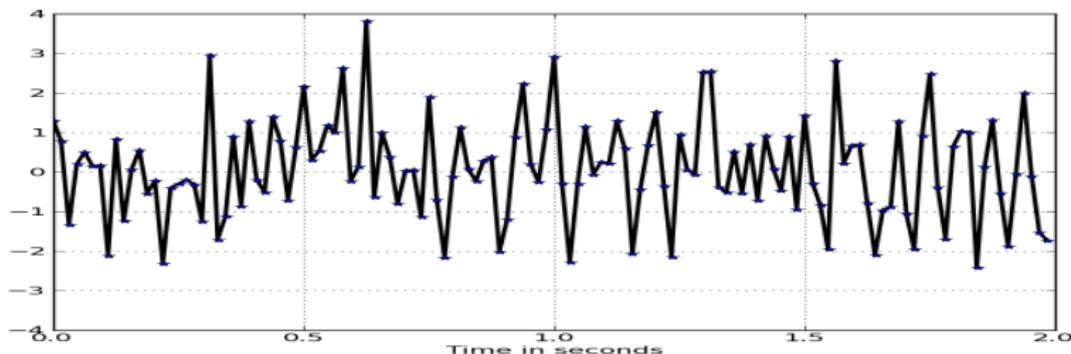
$$x(t) \quad 0 < t < T$$

Discrete Domain

Sampled Data at $f_{\text{samp}} = \Delta^{-1}$,

$$x_j = x(t_j) \quad t_j = 0, \Delta, \dots, (N-1)\Delta$$

Revise: Continuous and Discrete Fourier Transform



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Parseval Theorem:

$$\int x^2(t) dt = \int |\tilde{x}(f)|^2 df$$

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$$N \sum_0^{N-1} y_j z_j = \sum_0^{N-1} \tilde{y}_k \tilde{z}_k^*$$

Revise: Continuous and Discrete Fourier Transform

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$$\text{IFT: } x(t) = \int_{-\infty}^{\infty} \tilde{x}(f) e^{2\pi i f t} df$$

Convolution $Y = X \otimes Q$:

$$y(T) = \int x(t) q(T - t) dt$$

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Noise: White and Colored

What is the color of the noise?

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White noise == White light

Noise: White and Colored

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White light == All colors in equal proportions

White noise == All frequencies in equal proportions

Frequency spectrum is flat

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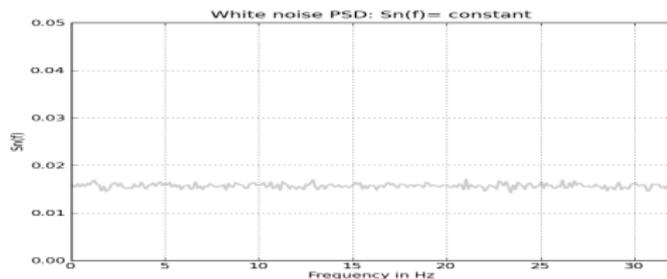
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Noise Power Spectral
Density:

$$S_n(f) = \frac{\langle |\tilde{n}(f)|^2 \rangle}{T}$$



Application : Electronic music, Audio testing etc

Noise: White and Colored

What is the color of the noise?

What is the colored noise?

Noise: White and Colored

What is the color of the noise?

What is the colored noise?

How to obtain blue color from white color??

Noise: White and Colored

What is the color of the noise?

What is the colored noise?

Colored noise - Pass white noise through a
band-pass filter

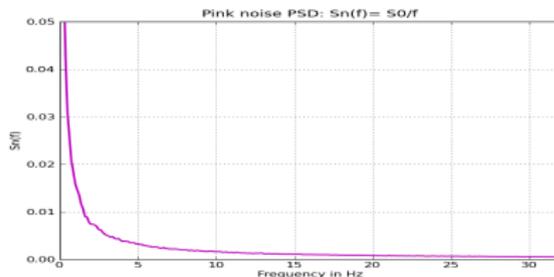
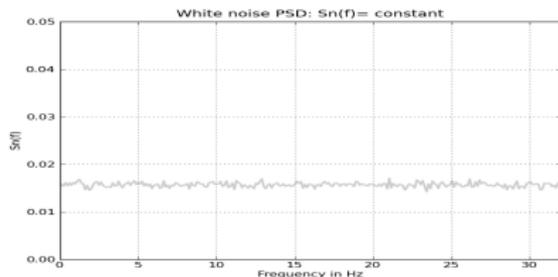
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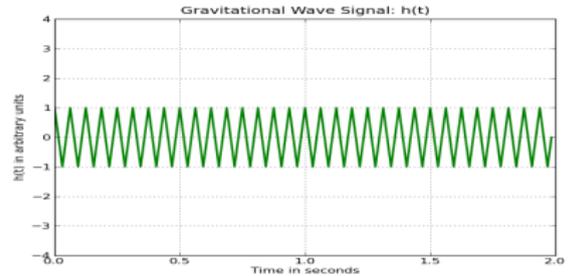
Example: Power law noise Noise power spectrum $1/f^\alpha$

$\alpha = 0, 1, 2, \dots \Rightarrow$ white, pink, brown, ... noise



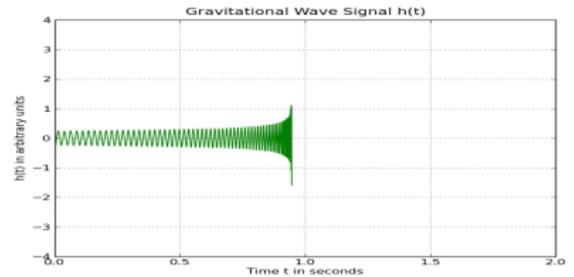
What is the best filter?

Signal – Known shape



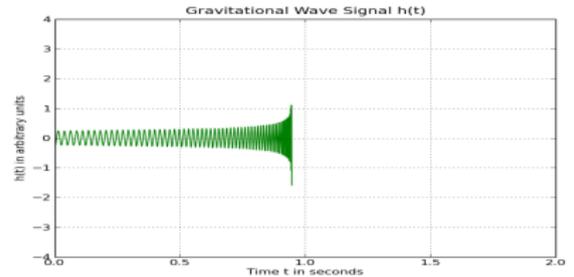
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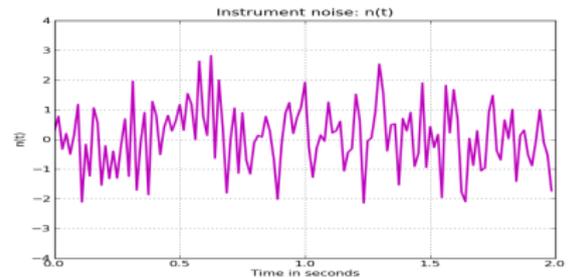


What is the best filter?

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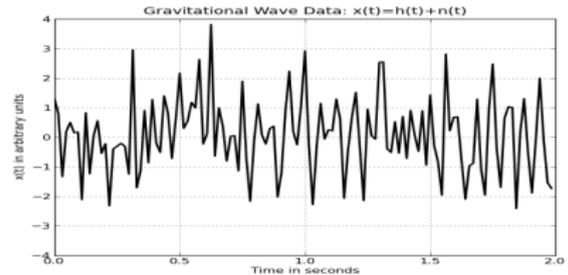


Noise – White



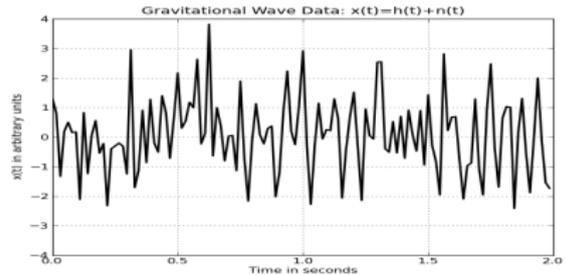
What is the best filter?

Data $x(t) = h(t) + n(t)$
Known signal in Stationary
noise



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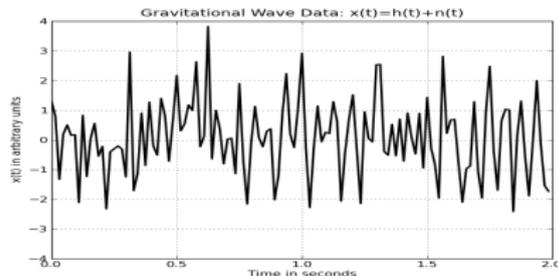


$$Y \equiv X \otimes Q = H \otimes Q + N \otimes Q$$

Signal Noise

What is the best filter?

Data $x(t) = h(t) + n(t)$
Known signal in Stationary noise



$$Y \equiv X \otimes Q = H \otimes Q + N \otimes Q$$

Signal Noise

Signal-To-Noise Ratio:

$$SNR = \frac{\langle X \otimes Q \rangle}{\sigma(X \otimes Q)} = \frac{H \otimes Q}{\sigma(N \otimes Q)}$$

What is the best filter?

Signal-To-Noise Ratio:

$$SNR = \frac{\langle X \otimes Q \rangle}{\sigma(X \otimes Q)} = \frac{H \otimes Q}{\sigma(N \otimes Q)}$$

Which filter function Q optimises the SNR?

What is the best filter?

Signal-To-Noise Ratio:

$$SNR = \frac{\langle X \otimes Q \rangle}{\sigma(X \otimes Q)} = \frac{H \otimes Q}{\sigma(N \otimes Q)}$$

$$\text{Recall — } y_j = \sum_{l=0}^{M-1} x_l q_{-l+j}$$

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$$\begin{aligned} \text{Recall — } y_j &= \sum_{l=0}^{M-1} x_l q_{-l+j} \\ &= \sum_{l=0}^{M-1} x_l q_{l+j}^r \end{aligned}$$

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What is the best filter?

Signal-To-Noise Ratio:

$$SNR = \frac{\langle X \otimes Q \rangle}{\sigma(X \otimes Q)} = \frac{H \otimes Q}{\sigma(N \otimes Q)}$$

$$Y = X \otimes Q = \frac{1}{M}(\tilde{X} \cdot \tilde{Q}^s)$$

$$SNR^2 = \frac{(\tilde{H} \cdot \tilde{Q}^s)^2}{\langle (\tilde{N} \cdot \tilde{Q}^s)^2 \rangle}$$

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Use: Stationary $\langle \tilde{n}_k \tilde{n}'_k^* \rangle = S_k \delta(k, k')$

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$$SNR^2 = \left(\sum_{k=0}^{M-1} \frac{\tilde{h}_k}{\sqrt{S_k}} \tilde{q}_k^* \right)^2 = (A \cdot B)^2$$

What is the best filter?

Signal-To-Noise Ratio:

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$$SNR_{max}^2 = \sum_{k=0}^{M-1} \frac{|\tilde{h}_k|^2}{S_k}$$

Matched filter : $-\tilde{q}_k \propto \frac{\tilde{h}_k}{S_k}$

What is the best filter?

$$SNR_{max}^2 = \sum_{k=0}^{M-1} \frac{|\tilde{h}_k|^2}{S_k}$$

Matched filter : $\tilde{q}_k = M \frac{\tilde{h}_k}{S_k}$

$$Y = X \otimes Q \implies y_j = \sum_{k=0}^{M-1} \tilde{x}_k \frac{\tilde{h}_k^*}{S_k} e^{2\pi jk/M}$$

Matched filter SNR

$$SNR_{MF}^2 = \sum_{k=0}^{M-1} \frac{|\tilde{h}_k|^2}{S_k}$$

Matched filter : $\tilde{q}_k = M \frac{\tilde{h}_k}{S_k}$

- Noise PSD $S_k = E(|\tilde{x}_k|^2)$ — Measure of spectral noise variance
- **White**: All frequencies are present ($S_k = S_0$); $q_k \propto h_k$.
- **Colored**: Noise variance is a function of frequency.

Small $S_k \Rightarrow$ small variance.

More SNR contribution with small S_k frequencies.
Demonstrates how noise distribution limits the sensitivity.

- **Band-pass operation**: Allows data to pass through a frequency band given by the instrument (S_k).

Matched filter SNR and Signal duration

$$SNR_{MF}^2 = \sum_{k=0}^{M-1} \frac{|\tilde{h}_k|^2}{S_k} = \int_{-\infty}^{\infty} \frac{|\tilde{h}(f)|^2}{S(f)} df \quad S(f) = \frac{T}{M^2} S_k$$

Matched filter SNR and Signal duration

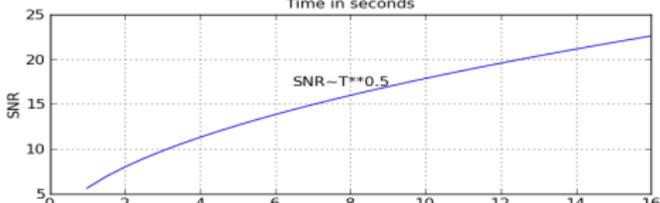
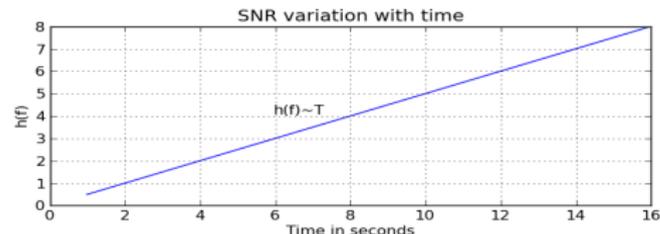
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$$\text{Sinusiod signal } h(t) = A \cos(2 \pi f_0 t) \quad 0 < t < T$$

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Sinusiod signal $h(t) = A \cos(2 \pi f_0 t) \quad 0 < t < T$



$$|h(f)|^2 \sim A^2 T^2 \quad df = T^{-1}$$

$$|h(f)|^2 df \sim A^2 T$$

$$SNR_{MF} \propto \sqrt{T}$$

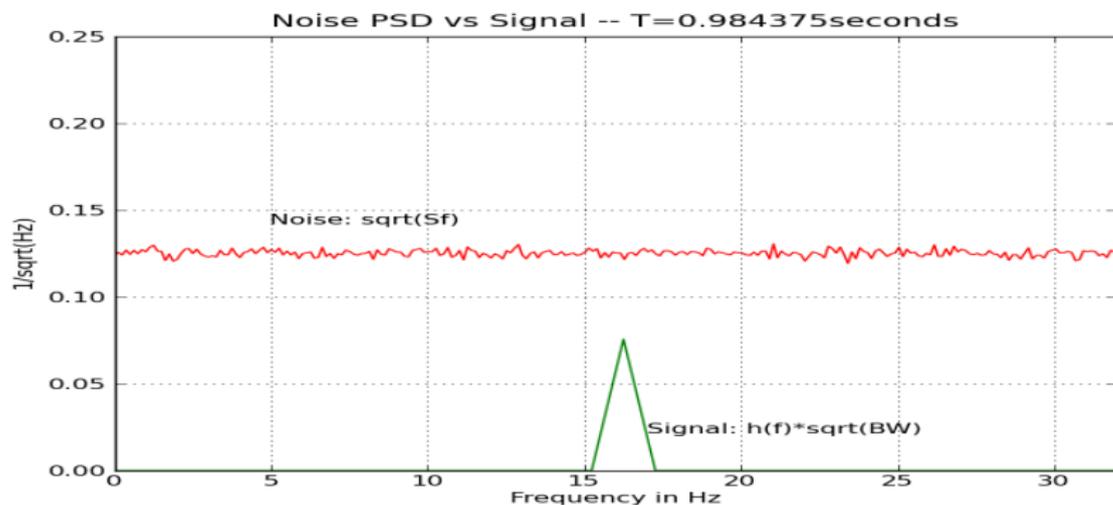
$$\propto \sqrt{\# \text{ cycles}}$$

$$A\sqrt{T} \text{ vs } \sqrt{S(f)}$$

Matched filter SNR and Signal duration

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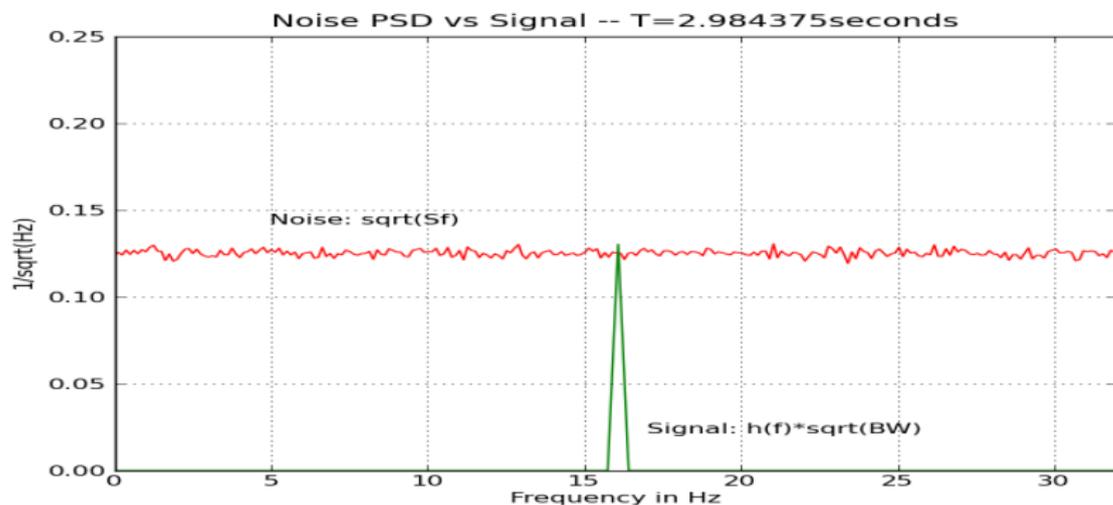
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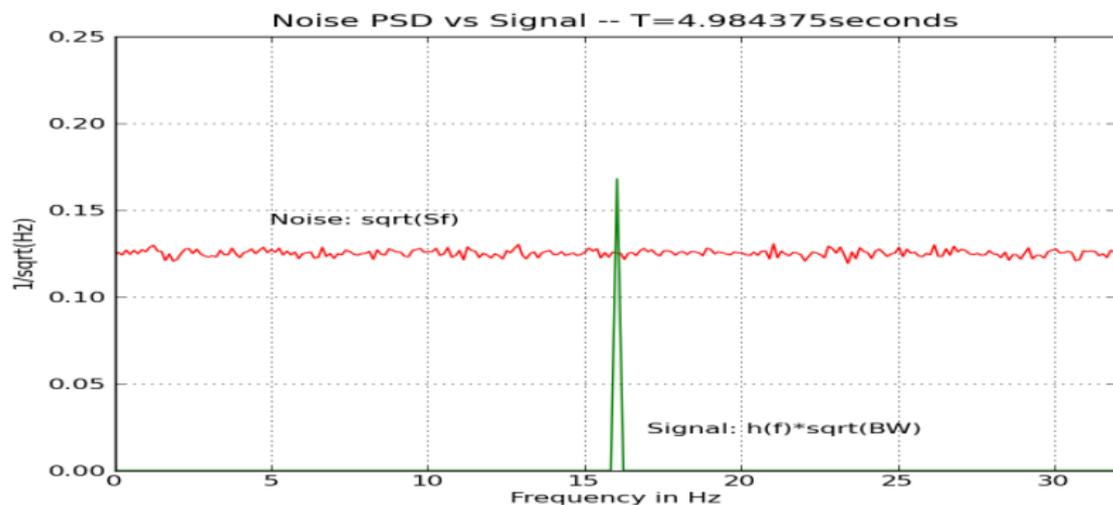
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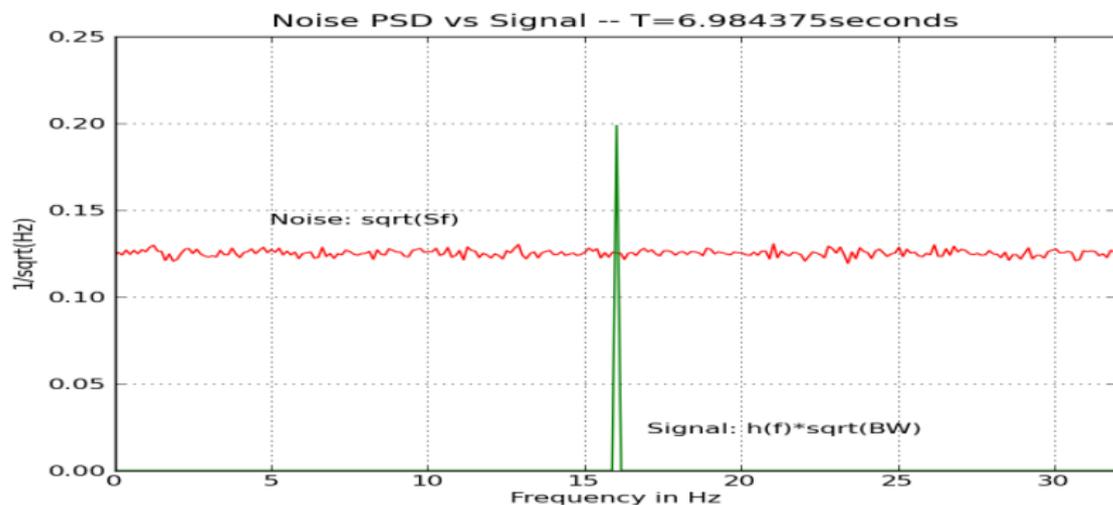
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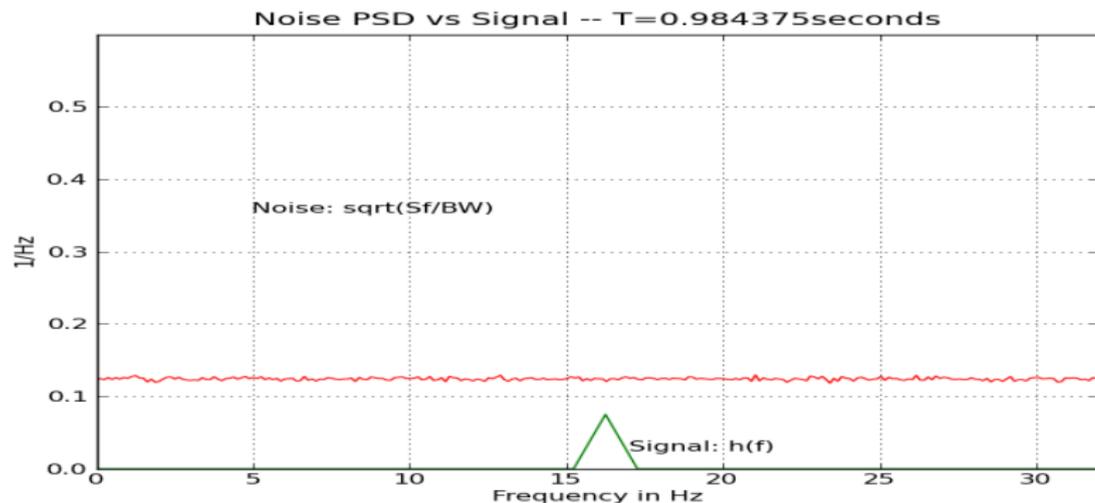
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Matched filter SNR and Signal duration

Sinusiod signal $h(t) = A \cos(2 \pi f_0 t)$ $0 < t < T$

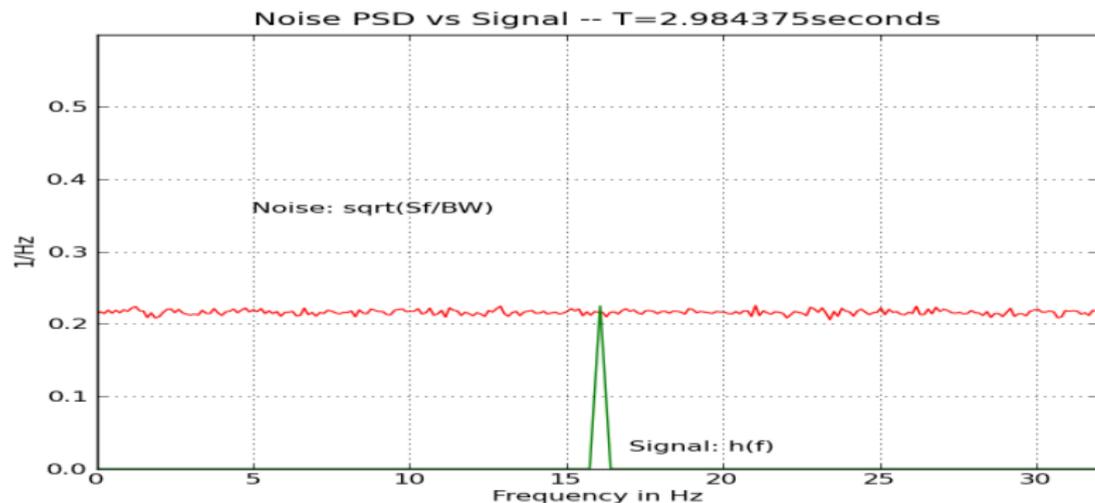


Signal is coherent – $\tilde{h}(f) \propto T$

Noise is incoherent (random walk) – $\tilde{n}(f) \propto \sqrt{T}$

Matched filter SNR and Signal duration

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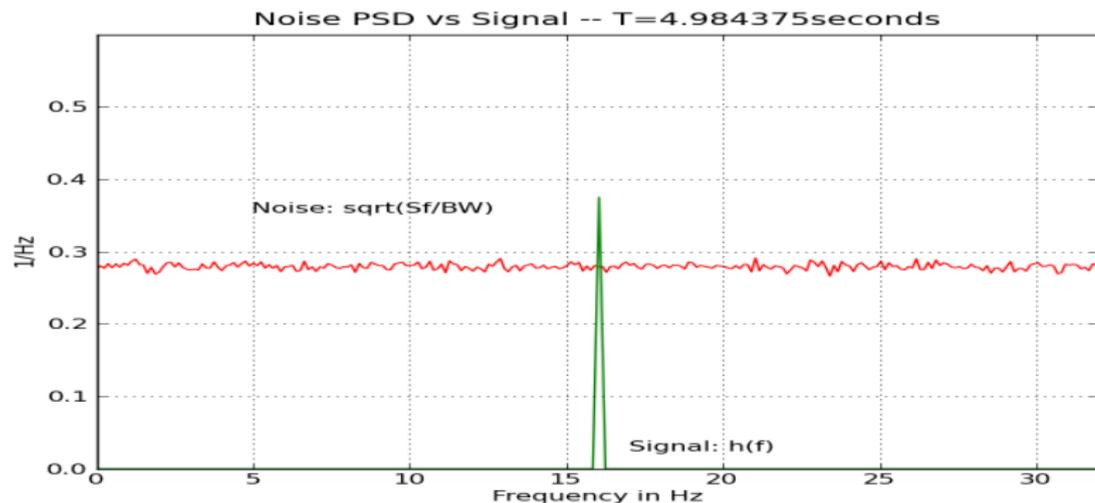


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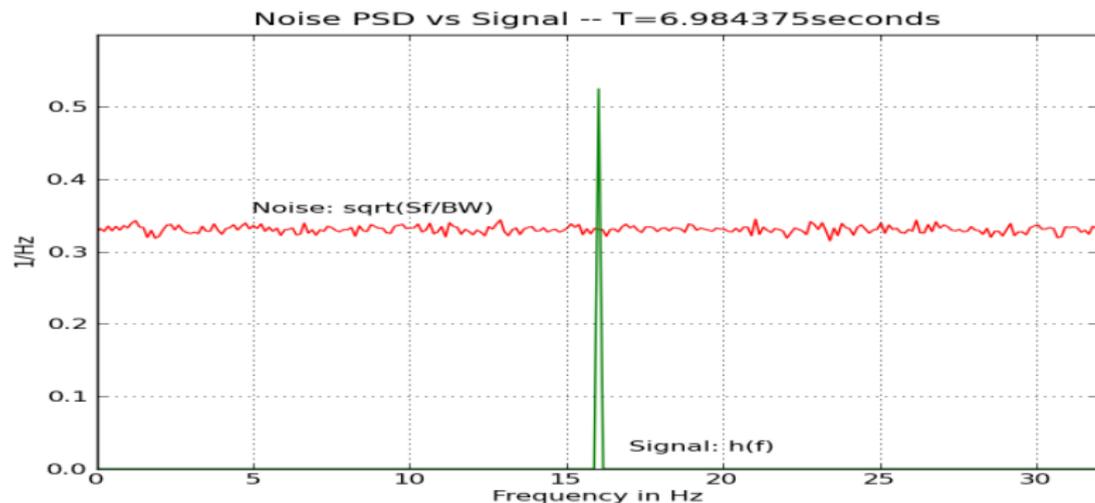


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Matched filtering: Inspiral waveforms

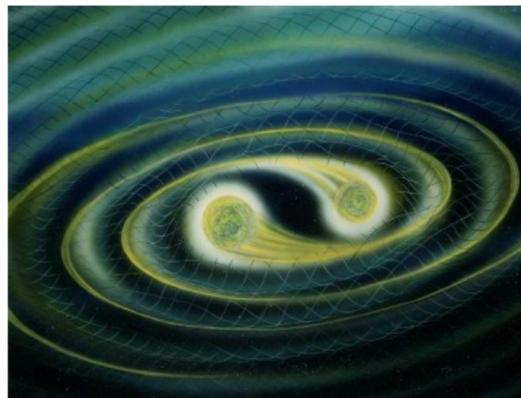
Compact binaries with NS, BH

$$h_+ \propto \frac{\mathcal{M}^{5/3}}{r} f^{2/3} \cos(2\pi \int f(t) dt)$$

$$h_\times \propto \frac{\mathcal{M}^{5/3}}{r} f^{2/3} \sin(2\pi \int f(t) dt)$$

$$\text{Chirp mass } \mathcal{M} = [\mu^3 M^2]^{1/5}$$

$$\text{Freq. } f \propto \mathcal{M}^{-5/8} (t_{\text{coal}} - t)^{-3/8}$$



Matched filtering: Inspirational waveforms

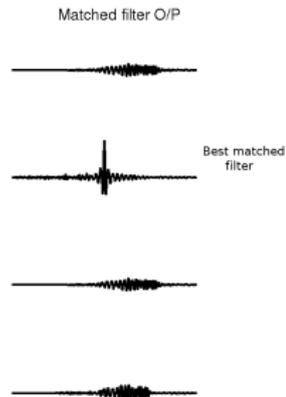
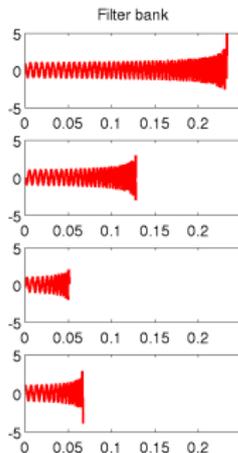
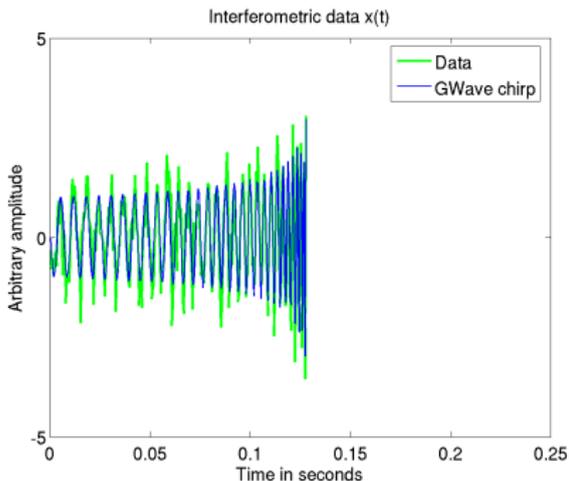
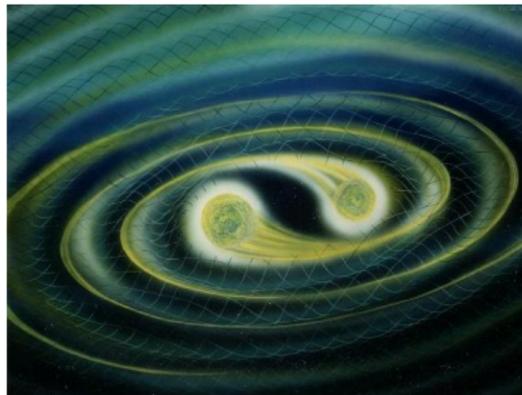
Compact binaries with NS, BH

$$h_+ \propto \frac{M^{5/3}}{r} f^{2/3} \cos(2\pi \int f(t) dt)$$

$$h_\times \propto \frac{M^{5/3}}{r} f^{2/3} \sin(2\pi \int f(t) dt)$$

Chirp mass $\mathcal{M} = [\mu^3 M^2]^{1/5}$

Freq. $f \propto \mathcal{M}^{-5/8} (t_{\text{coal}} - t)^{-3/8}$



Signal identification issues

Illustration:

Pick 10 random digits between 0-9

S1 – 3 4 6 2 1 0 9 8 2 6

S2 – 1 4 5 8 2 1 0 3 7 2

S3 – 1 2 3 4 0 2 4 7 8 3 .

S98 – 3 2 6 2 1 0 7 8 3 6

S99 – 1 3 5 8 1 1 0 5 7 8

S100 – 0 3 5 2 1 2 3 4 0 2

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Prob (1 2 3 4)? == 7 in 10 thousand

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2 in 100 instead of 7 in 10000 – ??

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Prob (1 0)? == 0.09

Signal identification issues

Strategy –

- Decide which signals to look for? Which signal?
- Compute the false alarm probability
- Assess detection significance by fixing the threshold

Signal identification issues

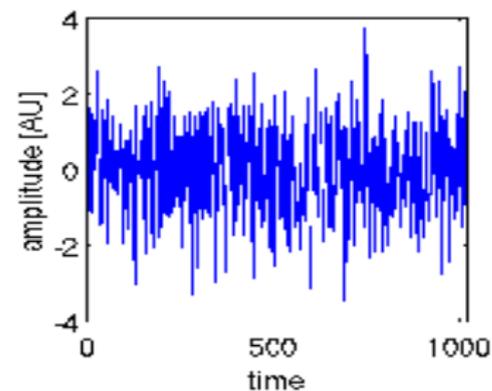
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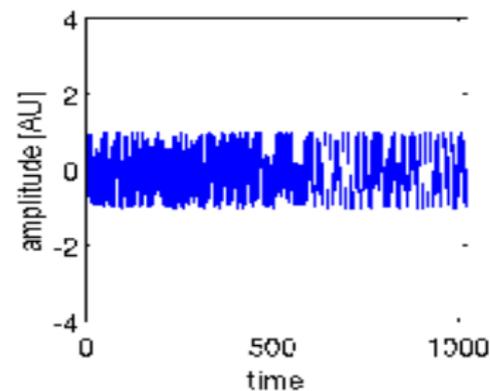
Searching multi-parameter signal increases the false alarm rate

Time-Frequency analysis

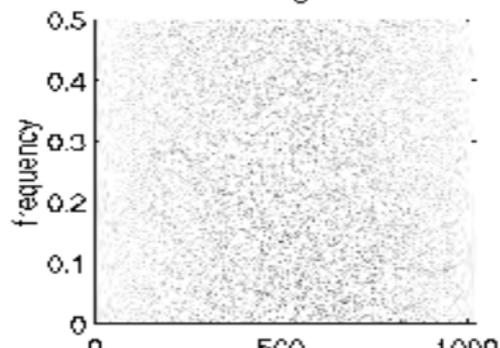
random CC in Gaussian noise, SNR=20



noise free random CC



discrete Wigner-Ville



random chirp [solid/green] best CC [dashed/red]

