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%Authors: Wendy Mendoza & Shahrear Khan Faisal
%Due Date: October 18, 2021
%Mid-term Project PHYS 5387
addpath 'https://drive.matlab.com/files/'
%(i):Loading Data for LIGO S5 strain
load S5 878486500 878486600 DARM.mat
% data1 = load('dat1.mat');
data1 = dat1;
fs = 16384; %Sampling frequency
t = length(dat1)/fs; %Duration in seconds
len = t*fs; %total length of the data
timeVec = 0:1/fs:(len-1)/fs; %time vector
%(ii):Estimation of the power spectrum of dat1 using pwelch
methods.
%Power spectrum density measure the signal power versus
frequency (width).
%PDS shows the strong and weak frequencies
variations (energy).
%They are known to become auto PSD Pxx(f)=Pxx(exp[ij2pif])
and cross PSD Pxy(f) = Pxy(exp[j2pif])
%Auto-PSD describes the frequency of the power of x[n] and
real and
%nonnegative. Cross-PSD describes the frequency component
in x[n] are
%associated with large or small amplitude same frequency
v[n].
%For the PSD we use the terms of Welch method to approach a
spectral density estimation.
% Using PWELCH methods to calculate the spectral density.
[pxx, f] = pwelch(data1, [], [], [], fs);
% (iii):Plotting the time series and the power of spectrum.
figure(1)
subplot(2,1,1);
plot(timeVec, dat1);
xlabel('Time(sec)');
ylabel('Amplitude');
title('Time Series');
subplot(2,1,2);
plot(f, 10*log10(pxx));
xlabel('Frequency(Hz)');
```

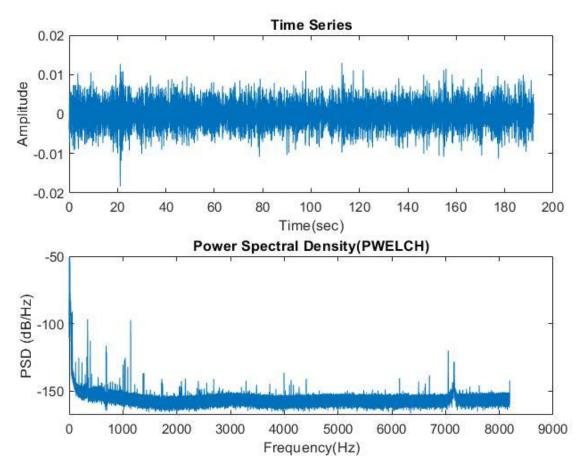


Figure 1: Time Series at the top, with strain in the Y-axis and time in the X-axis. The signal duration is 192 seconds. Power Spectral Density of the Original Data at the bottom, with PSD in the Y-axis and frequency in the X-axis.

Frequency ranges from 0 to more than 8 kHz.

```
dat low pass = filter(b,a,data1);
[pxx 2, f2] = pwelch(dat low pass,[],[],[],fs);
figure(2)
subplot(2,1,1);
plot(timeVec, dat low pass);
xlabel('Time(seconds)');
ylabel('Amplitude');
title('Filtered Time Series');
subplot(2,1,2);
plot(f2, 10*log10(pxx 2))
xlabel('Frequency (Hz)');
ylabel('PSD (dB/Hz)');
title('Power Spectral Density(PWELCH) low pass filter');
%The low pass filter retains frequencies below a given cut-
off which means
%it eliminate the higher frequencies to allow the lower
frequencies to pass
%through.
```

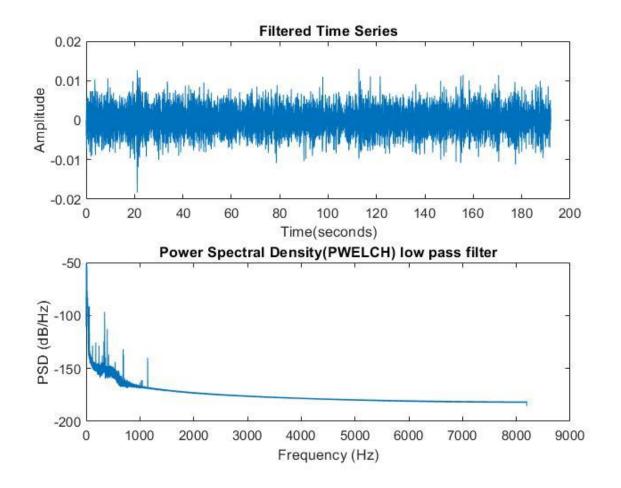


Figure 2: At the top there is time series of the data after low pass filtering at 2048 Hz. Again, time is in the X-axis and amplitude is in the Y-axis. Power Spectral Density after low pass is at the bottom. The peaks of the frequency components after 2048 Hz have disappeared as expected.

```
%(v):Resample the data to a lower sampling frequency
p = 1; %Resampling factors
q = 4; %Resampling factors
dat_low_pass2=resample(dat_low_pass,1,4); %change sampling
rate
%New sample frequency will be 4096Hz
fs_new = fs * p/q;
t2 = length(dat_low_pass2)/fs_new; %duration in seconds
len2 = t2*fs_new; %lenght of the data
timeVec2 = 0:1/fs_new:(len2-1)/fs_new; %time vector
```

```
% (vi) pwelch the new sampling rate
[pxx 3, f3] = pwelch(dat low pass2,[],[],[],fs new);
figure(3);
plot(f3, 10*log10(pxx 3));
xlabel('Frequency(Hz)');
ylabel('PSD (dB/Hz)');
title('Power Spectral Density with Resample Data');
% Comparing the power spectrum of the data before and after
low pass
% filering, we can see that before the data had frequency
components above
% 8 kHz, but after applying the filter the frequency
components above 2048
% Hz have been discarded, which is expected, as out cut of
frequency of the
% butterworth filter was 2048 Hz.
```

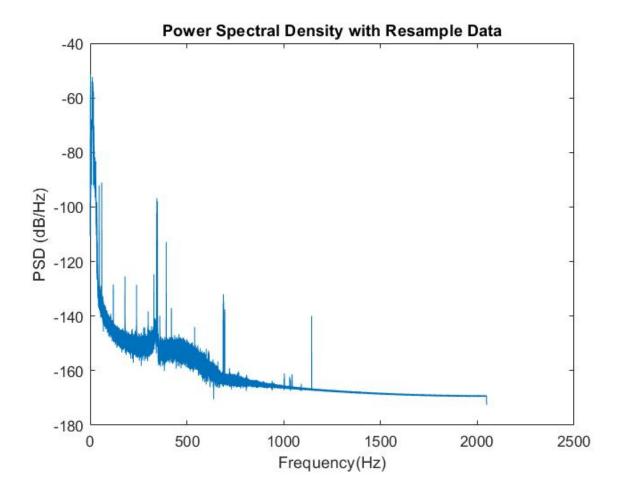


Figure 3: Power Spectral Density of the resampled data, with frequency in the X-axis and PSD in the Y-axis. The frequency ranges from 0 to 2048 Hz

%(vii):Whiten the data

```
pxx_median_est = rngmed2(pxx,256); % Median is the middle number of a Gaussian distribution in an order set of data values % symmetrically around mean=median=mode. % Inverse the median freq=0:0.125/(fs*4):1; % This function design filters with frequency inverse and the magnitude characterisitics of % the white noise which is the containing vectors desires in the frequency. N = 500;
```

```
bfilt=fir2(500, freq', 1./sqrt(pxx median est)); %N is number
of order, frequency, & magnitude
%fir2 is a digital filter order specified as an integer
scalar.
%This return the 500 orders FIR filter with frequency
magnitude charateristics.
%The filter coefficients are obtained by linearly
interpolating the required
%frequency response onto a dense grid and then using the
inverse Fourier transform and a Hamming window.
Dat whitened=fftfilt(bfilt,dat low pass2);
[pxx 4, f4] = pwelch(dat whitened, [], [], [], fs new);
figure (4);
subplot(2,1,1)
plot(timeVec2, dat whitened);
xlabel('Time(sec)');
ylabel('Amplitude');
title('Time Series of Whitened Data');
subplot(2,1,2);
plot(f4, 10*log10(pxx 4));
xlabel('Frequency(Hz)');
ylabel('PSD (dB/Hz)');
title('Power of Spectral Density of Whiten Data')
```

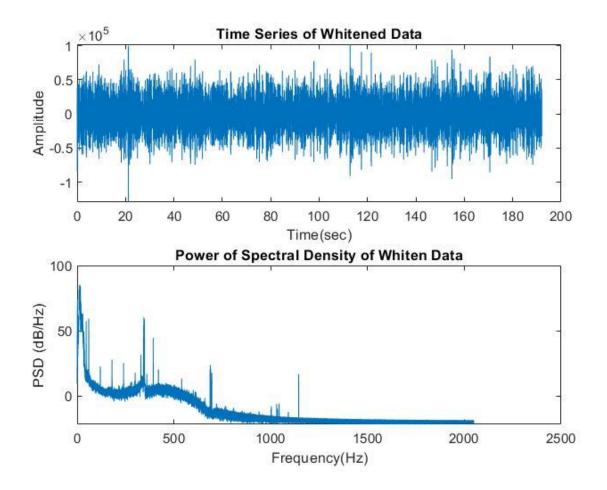


Figure 4. The first graph is the Time Series of Whitened Data the x-axis represents the Time in seconds and the y-axis is the Amplitude. The second graph is the Power Spectral Density of Whiten Data with the x-axis of frequency ranges between 0-2300 Hz and peaks as values on the y-axis PSD.

```
%(viii) Applying the Bandpass filter to the whitened data
[b,a]=butter(6,[100 1024]/2048);
dat_whitened2=filtfilt(b,a,dat_whitened);
[pxx_5, f5] = pwelch(dat_whitened2,[],[],[],fs_new);
figure(5);
subplot(2,1,1)
plot(timeVec2,dat_whitened2);
xlabel('Time(sec)');
ylabel('Amplitude');
title('Time Series of Bandpass Whiten Data');
```

```
subplot(2,1,2);
plot(f5, 10*log10(pxx_5));
xlabel('Frequency(Hz)');
ylabel('PSD (dB/Hz)');
title('Power of Spectral Density Bandpass Whiten Data');
%The bandpassfilter retains frequencies between a given
lower cut-off and a
%higher cut-off. % Why we bandpassed the whitened data -
Because LIGO is most sensitive
%between 100 Hz and 1000 Hz. Outside this frequency range
signals other
%than our desired signals (noise) prevail, which makes it
nearly
%impossible to detect gravitational wave in those frequency
ranges.
```

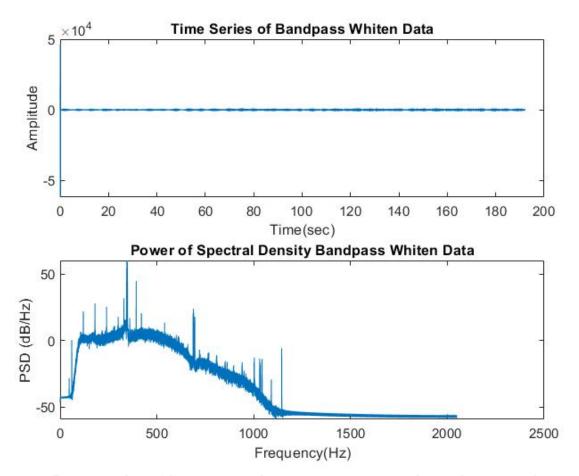


Figure 5. In the first graph, we can see the Time series of the Bandpass Whiten Data in the x-axis of Time in seconds from 0-195 seconds and the y-axis of amplitude. The second graph below is the Power of Spectral Density Bandpass Whiten Data the x-axis frequency (Hz) ranges between 0-2200 Hz.

```
%(ix):Resample the whitened data to 2018 HZ
dat_low_pass2=resample(dat_low_pass,1,4);
nfft = 2018 * 4;
[pxx_6, f6] = pwelch(dat_low_pass2,[],[],nfft,fs_new);
figure(6);
plot(f6, 10*log10(pxx_6));
xlabel('Frequency(Hz)');
ylabel('PSD (dB/Hz)');
title('Power of Spectral Density Resample Data to 2018 Hz');
```

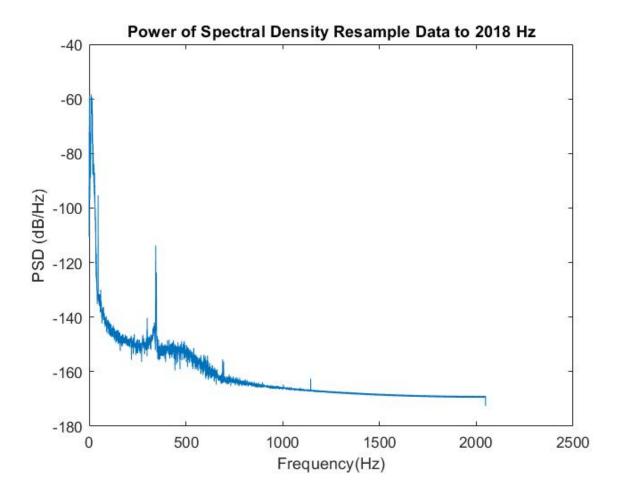


Figure 6. This figure shows the Power of Spectral Density Resample Data to 2018 the x-axis is the frequency from ranges of 0-2100 Hz and the y axis PSD(dB/Hz) peaking its values.

%(x): In the Power of Spectral Density Resample Data of 2018 Hz we can see

%several narrowband noises. These narrowband noises could be produce by

%electrical and mechanical resonances. On our loop below we created the

%list of the highest peak frequencies that detect our narrowband frequency

%ranges and the half of the maximum amplitude. Running our loop it detected

\$11 lines that could be the narrowband noises. We were able to identify

```
%some source between the ranges of OHz-1200Hz due to
measured noise, expected noise
%& thermal noise.
% Finding peaks in the power spectral density plot.
[pks, locs, w, p] = findpeaks(10*log10(pxx 6));
pks = pks';
locs = locs';
w = w';
p = p';
p sorted = sort(p, 'descend');
indx = zeros(1,11);
% Get the indices of the strongest narrowband lines.
for i = 1:11
    for j = 1:length(pks)
        if p sorted(i) == p(j)
            indx(i) = locs(j);
        end
    end
end
freq list = f6(indx);
w1 = w(w indx);
%Full Width Half Maximum.
w1 = w1'/2;
%2X2 matrix with narrowband frequencies and full width half
maximum.
narrowband = [freq list w1];
```