Lab Topic 4

SNR of LR test

Joint pdf of noise:

$$p_{\bar{E}}(\bar{y}) = \frac{1}{(2\pi)^{N/2} |\mathbf{C}|^{1/2}} \exp\left(-\frac{1}{2} ||\bar{y}||^2\right)$$
$$||\bar{y}||^2 = \bar{y} \mathbf{C}^{-1} \bar{y}^T$$

- $\blacktriangleright \quad \text{Let } \langle \bar{z}, \bar{y} \rangle = \bar{z} \mathbf{C}^{-1} \bar{y}^T$
- LR statistic: $\Lambda = \frac{1}{2} ||\bar{y}||^2 \frac{1}{2} ||\bar{y} \bar{s}||^2 = \langle \bar{y}, \bar{s} \rangle \frac{1}{2} \langle \bar{s}, \bar{s} \rangle \to \Lambda = \langle \bar{y}, \bar{s} \rangle$
- ► SNR: ||*s*||

Inner product for stationary noise

White Gaussian Noise

$$\langle \bar{x}, \bar{y} \rangle \to \sum_{k=0}^{N-1} x_k y_k$$

Stationary Gaussian noise with Power Spectral Density (PSD) $S_n(f)$

$$\langle \bar{x}, \bar{y} \rangle \to \frac{\Delta}{N} \tilde{x} (\tilde{y}^{\dagger} . / \bar{S}_{n}^{T}) = \frac{1}{T} \tilde{x} (\tilde{y}^{\dagger} . / \bar{S}_{n}^{T})$$

Where $\tilde{x} = F\bar{x}$ is the DFT

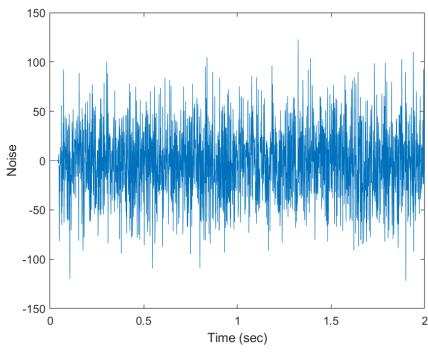
T is the length of the data $(\frac{\Delta}{N} = \frac{1}{N f_s} = \frac{1}{T})$

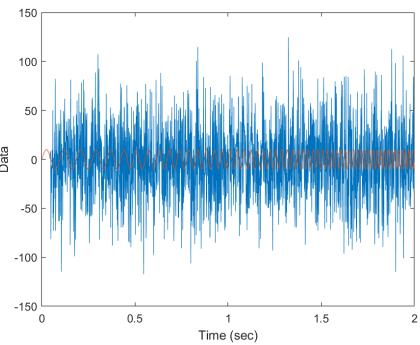
 \bar{S}_n is $S_n(f)$ evaluated at the DFT frequencies

Normalization of signal amplitude: SNR

Objective:

- 1. Generate an *N* sample realization of colored gaussian noise with some specified Power Spectral Density (PSD)
- 2. Normalize a given signal vector of *N* samples such that the Likelihood Ratio test for noise with the above PSD has a specified Signal to Noise Ratio (SNR)





Codes

- Generating stationary Gaussian noise: NOISE/statgaussnoisegen.m
 - Script NOISE/colGaussNoiseDemo.m has been modified to use this function
 - ► The function supplies the right scale factor for the generated noise (earlier we only wanted to get the shape of the PSD right, not its overall scale)
- ► Taking an inner product: **DETEST/innerprodpsd.m**
- ▶ Demo for normalization of signal: **DETEST/SNRcalc.m**

- ► Follow the script **DETEST/SNRcalc.m** and write a function that calculates the normalization factor for a given signal and noise PSD
 - ► Inputs:
 - ► Signal vector
 - ► Sampling frequency
 - ▶ PSD at positive DFT frequencies
 - ► SNR
 - Outputs:
 - ► Normalized signal vector
 - ► Normalization factor
- ► Code should do a sanity check: Number of PSD values should match the number of positive DFT frequencies based on the length of the signal vector

- ► Generate a realization of initial LIGO noise using NOISE/statgaussnoisegen.m
- ▶ Add the signal your team was assigned with an SNR=10 to the noise realization
- ► Time domain: Plot the data (signal+noise) realization and the signal
- Frequency domain: Plot the periodogram of the noise and data together
- Time-frequency domain: Plot the spectrogram of the data
- Can you see the signal in any of the three domains? Namely, the time, or the Fourier, or the time-frequency domains

- Follow **DETEST/SNRcalc.m** to write a function to calculate the Likelihood Ratio for a given data vector
 - ▶ What are the inputs needed?
- Run this function over M realizations of data under H_0 and M realizations of data under $H_1 \rightarrow$ Estimate the SNR of the LR test as shown in SNRcalc.m
- Does the estimated SNR come close to the one you had normalized the signal with? (It should!)

Test significance

- For any detection statistic $\Gamma(\bar{y})$, a function of the data \bar{y} , the **significance** of detection made with the statistic is defined as :
 - "The probability of observing a value of $\Gamma \geq$ its observed value for given data if the null hypothesis is true"

$$\Pr\{\Gamma \ge \Gamma^{(observed)} | H_0\} = \int_{\Gamma^{(observed)}}^{\infty} \underbrace{p_{\Gamma}(x | H_0)}_{Estimated} dx$$

 $\Gamma^{(observed)} = \Gamma(\bar{y})$: The value of Γ obtained for the given data \bar{y}

Estimating probability

$$\Pr\{\Gamma \ge \Gamma^{(observed)} | H_0\} = \int_{\Gamma^{(observed)}}^{\infty} \underbrace{p_{\Gamma}(x | H_0)}_{Estimated} dx$$

- Frequentist probability: Obtain a large number of trial values of Γ
 - ▶ Trial value \Rightarrow generate a data realization and compute the Γ value for it
 - ► Then,

$$\Pr\{\Gamma \geq x\} \approx \frac{Number\ of\ times\ \Gamma \geq x}{Total\ number\ of\ data\ realizations}$$

The approximation above becomes better as the total number of trials increases

Exercise #4: Part 1

Write a function to calculate the GLRT for the unknown amplitude case

$$L_G = \langle \bar{y}, \bar{q} \rangle^2$$
 where $\bar{q} = \frac{\bar{s}}{\|\bar{s}\|}$

- ► Inputs:
 - ightharpoonup Data vector \bar{y}
 - ► Noise PSD at positive DFT frequencies
- Output:
 - ► GLRT value
- This function is similar to the one you wrote earlier in Exercise #3 except that the data inner product is computed with the template vector \bar{q} , which is the signal vector normalized to have SNR=1
 - Note: A new function DETEST/normsig4psd.m is available for normalizing a signal (it is the solution to Lab Topic 4→Exercise #1)

Exercise #4: Part 2

- You have been provided 3 data realizations named **DETEST/data<n>.mat** where n=0,1,3
- You are told that the signal in each data realization is a quadratic chirp but the amplitude is unknown (maybe zero also!)
- ▶ Signal parameters: a1=10; a2=3; a3=3;
- You will calculate the GLRT for this signal, using the function coded in Exercise #4: Part1, for each of the 3 data realizations
 - ► The quadratic chirp signal is the one coded in DSP/crcbgenqcsig.m
 - and the noise PSD used is the one in DETEST/SNRCalc.m noisePSD = @(f) (f>=100 & f<=300).*(f-100).*(300-f)/10000 + 1;</p>
- Use simulations to estimate the significance of the GLRT values for the 3 data realizations
- Finally, git push a text files into DETEST called Team<n>_testSig.txt containing the significances you found for the 3 data realizations (each value on a new line starting from n=1). In the last line of the file, write the number of data realizations your simulated to estimate the significance

- Extend the function in Exercise #4 Part 1 to create a new function:
 - ► Inputs:
 - ▶ Data vector
 - \triangleright Parameter values a_1 , a_2 , a_3 for a quadratic chirp
 - ► Noise PSD vector at positive DFT frequencies
 - Output:
 - value of Likelihood ratio maximized over amplitude: $\langle \bar{y}, \bar{q}(\Theta) \rangle^2$ where Θ is the set of input parameters a_1 , a_2 , a_3
- See the script **DETEST/LRcalc.m** for an example of the steps involved for a specific set of values for Θ
- ► Test: Repeat Exercise #4 Part 2 using this function: You should get identical values for the significance

Conclusion: Lab Topic 4

The function created in Exercise #5 will serve as the fitness function that will be optimized by PSO to obtain the GLRT

$$L_G = \max_{\Theta} \langle \bar{y}, \bar{q}(\Theta) \rangle^2$$

- ► This optimization problem and the results associated with it are fully discussed in the textbook
- ► The main difference is that we have generalized the textbook problem to the case of colored Gaussian noise