

In Class Example: Quick and Dirty Gradient Descent and Newton rhapson tutorial

April 16, 2019

Sinusoid Composite Hypothesis Examples

- Section 7.6 in Kay's detection theory
- $H_0 : \mathbf{x}[n] = \mathbf{w}[n]$
- $H_1 : \mathbf{x}[n] = A \cos(2\pi f_0(n - n_0) + \phi) + \mathbf{w}[n]$
- We want to form the hypothesis test when:
 - ▶ A is unknown
 - ▶ A and ϕ are unknown
 - ▶ A , ϕ , and f_0 are unknown
 - ▶ All parameters unknown

Unknown Amplitude

- The generalized likelihood ratio test for the unknown amplitude is:

$$\frac{\exp(\frac{-1}{2\sigma^2} \sum_{n=0}^{N-1} (x[n]^2 - 2x[n]\hat{A}\cos(2\pi f_0(n-n_0) + \phi) + \hat{A}^2 \cos(2\pi f_0(n-n_0) + \phi)^2))}{\exp(\frac{-1}{2\sigma^2} \sum_{n=0}^{N-1} x[n]^2)}$$

- \hat{A} is the maximum likelihood estimate of A based on x :

$$\hat{A} = \arg \max_A \sum_{n=0}^{N-1} (x[n]^2 - x[n]A\cos(2\pi f_0(n-n_0) + \phi) + A^2 \cos(2\pi f_0(n-n_0) + \phi)^2)$$

- Take the derivative with respect to A :

$$\sum_{n=0}^{N-1} -2x[n] \cos(2\pi f_0(n-n_0) + \phi) + 2A \cos(2\pi f_0(n-n_0) + \phi) = 0$$

- This means that $\hat{A} = \frac{\sum_{n=0}^{N-1} x[n] \cos(2\pi f_0(n-n_0) + \phi)}{\sum_{n=0}^{N-1} \cos(2\pi f_0(n-n_0) + \phi)^2}$

Unknown Amplitude Cont.

- Using the fact that

$\sum_{n=0}^{N-1} x[n] \cos(2\pi f_0(n - n_0) + \phi) = \hat{A} \sum_{n=0}^{N-1} \cos(2\pi f_0(n - n_0) + \phi)^2$ we can rewrite the likelihood ratio:

$$-\sum_{n=0}^{N-1} -2\hat{A}^2 \cos(2\pi f_0(n - n_0) + \phi)^2 + \hat{A}^2 \cos(2\pi f_0(n - n_0) + \phi)^2 \stackrel{H_1}{\underset{H_0}{\gtrless}} \log \lambda'$$

$$\hat{A}^2 \stackrel{H_1}{\underset{H_0}{\gtrless}} \frac{\lambda'}{\sum_{n=0}^{N-1} \cos(2\pi f_0(n - n_0) + \phi)^2}$$

- This means that our final test statistic is:

$$(\sum_{n=0}^{N-1} x[n] \cos(2\pi f_0(n - n_0) + \phi))^2 \underset{H_0}{\gtrless} \lambda'$$

Performance of Only Amplitude Detector

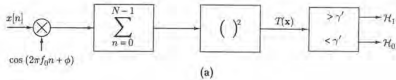


Figure 1: Block Diagram of Detector when Only amplitude is unknown

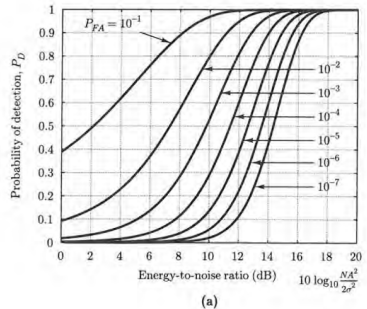


Figure 2: Performance of Detector when Only amplitude is unknown

Amplitude and Phase Unknown

- When Amplitude and Phase are unknown, we have to reduce to the case where either $A > 0$ or $A < 0$, otherwise shifting the phase by π means we have two functions that can produce the same signal.
- Assume $A > 0$, then find \hat{A} and $\hat{\phi}$ from:

$$\arg \max_{A, \phi} \sum_{n=0}^{N-1} (x[n]^2 - x[n]A \cos(2\pi f_0(n - n_0) + \phi) + A^2 \cos(2\pi f_0(n - n_0) + \phi)^2)$$

- Now we have to set the gradient equal to $(0, 0)^T$:

$$\nabla \log(p(x; A, \phi)) =$$

$$\begin{bmatrix} \sum_{n=0}^{N-1} -2x[n] \cos(2\pi f_0(n - n_0) + \phi) + 2A \cos(2\pi f_0(n - n_0) + \phi) \\ \sum_{n=0}^{N-1} 2x[n]A \sin(2\pi f_0(n - n_0) + \phi) - A^2 \sin(2\pi f_0(n - n_0) + \phi) \end{bmatrix}$$

- Using some trig identities we can approximate the solutions with :

$$\hat{A} = \sqrt{\left(\frac{2}{N} \sum_{n=0}^{N-1} x[n] \cos(2\pi f_0 n)\right)^2 + \left(\frac{2}{N} \sum_{n=0}^{N-1} x[n] \sin(2\pi f_0 n)\right)^2}$$

$$\hat{\phi} = \arctan\left(\frac{\frac{2}{N} \sum_{n=0}^{N-1} x[n] \cos(2\pi f_0 n)}{\frac{2}{N} \sum_{n=0}^{N-1} x[n] \sin(2\pi f_0 n)}\right)$$

Amplitude and Phase Unknown

- Now the Likelihood ratio is:

$$-1/2\sigma^2 \sum_{n=0}^{N-1} -2\hat{A}^2 \cos(2\pi f_0(n-n_0) + \hat{\phi})^2 + \hat{A}^2 \cos(2\pi f_0(n-n_0) + \hat{\phi})^2 \underset{H_0}{\geq} \log \lambda$$

- Using the substitution $\hat{\alpha}_1 = \hat{A} \cos(\hat{\phi})$ and $\hat{\alpha}_2 = -\hat{A} \sin(\hat{\phi})$ and some more trigonometry we can simplify this expression into:

$$\frac{N}{4\sigma^2} (\hat{\alpha}_1^2 + \hat{\alpha}_2^2) \underset{H_0}{\geq} \log \lambda$$

- But it turns out that

$\hat{\alpha}_1^2 + \hat{\alpha}_2^2 = \frac{2^2}{N^2\sigma^2} ((\sum_{n=0}^{N-1} x[n] \cos(2\pi f_0 n))^2 + (\sum_{n=0}^{N-1} x[n] \sin(2\pi f_0 n))^2)$, so the test becomes:

$$\frac{1}{N\sigma^2} ((\sum_{n=0}^{N-1} x[n] \cos(2\pi f_0 n))^2 + (\sum_{n=0}^{N-1} x[n] \sin(2\pi f_0 n))^2) = \frac{I(f_0)}{\sigma^2} \underset{H_0}{\geq} \log \lambda$$

- This is either the sum of two correlators similar to the unknown amplitude case, or something called the [periodogram](#), which estimates $|X(f)|^2$, where $X(f)$ is the fourier transform of $x[n]$.

Amplitude and Phase Unknown

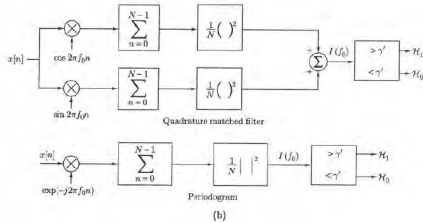


Figure 3: Block Diagram of Detector when Amplitude and Phase are unknown, this is sometimes called an incoherent detector or quadrature matched filter.

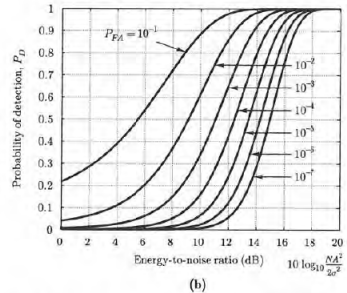


Figure 4: Performance of Detector when Amplitude and Phase are unknown

Performance Comparison

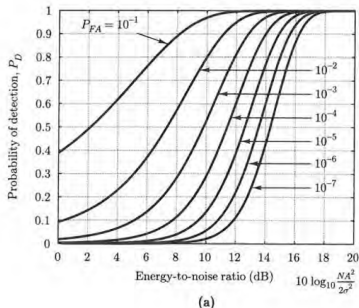


Figure 5: Performance of Detector when Only amplitude is unknown

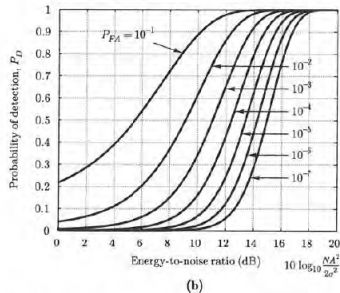


Figure 6: Performance of Detector when Amplitude and Phase are unknown - note that the PD is generally worse than the corresponding points in the first graph.

Amplitude, Phase, and Frequency Unknown

- When the Amplitude, Phase, and Frequency are unknown we need to find the maximum of:

$$\arg \max_{A, \phi} \sum_{n=0}^{N-1} (x[n]^2 - x[n] A \cos(2\pi f_0(n - n_0) + \phi) + A^2 \cos(2\pi f_0(n - n_0) + \phi)^2))$$

across three variables.

- To do this, define $I(f_0) = \frac{2}{N} ((\sum_{n=0}^{N-1} x[n] \cos(2\pi f_0 n))^2 + (\sum_{n=0}^{N-1} x[n] \sin(2\pi f_0 n))^2)$ as the periodogram of $x[n]$ as a function of f_0 .
- To perform the maximization, note that if we know the best \hat{f}_0 we can compute \hat{A} and $\hat{\phi}$ the same way as the previous case - this is an example of concentrated likelihood.
- Also, The likelihood under H_0 is not a function of any of the variables so:

$$\begin{aligned} \max_{A, f_0, \phi} p(x; A, f_0, \phi) &= \frac{\max_{A, f_0, \phi} p(x; A, f_0, \phi)}{p(x; H_0)} = \max_{f_0} \frac{p(x; \hat{A}, \hat{\phi}, f_0)}{p(x; H_0)} = \\ &= \max_{f_0} \log \frac{p(x; \hat{A}, \hat{\phi}, f_0)}{p(x; H_0)} = \max_{f_0} \frac{I(f_0)}{\sigma_2^2} \end{aligned}$$

Amplitude, Phase, and Frequency Unknown cont.

- So the likelihood ratio test becomes:

$$\max_{f_0} \frac{I(f_0)}{\sigma^2} \underset{H_0}{\geq} \log \lambda$$

- This is the same as finding the maximum of the FFT output in matlab and comparing it to your threshold.
- In general the performance gets worse as the frequency increases.

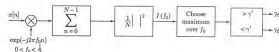


Figure 7: Block Diagram of Detector when Amplitude, Phase, and Frequency are unknown

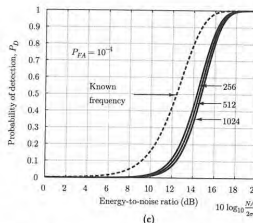


Figure 8: Performance of Detector when Amplitude, Phase, and Frequency are unknown

Unknown Amplitude, Phase, Frequency, and Arrival Time

- The parameter n_0 is the delay of the signal, also called the "Arrival Time". Assume we have a "long" set of data.
- Estimating the arrival time is finding the exact time window $[n_0, n_0 + N - 1]$ steps that the signal is active.
- First, we have to modify our \hat{A} and $\hat{\phi}$ expressions from the previous cases
- Let $\hat{\alpha}_1 = \frac{2}{N} \sum_{n=n_0}^{n_0+N-1} x[n] \cos(2\pi \hat{f}_0(n - n_0))$ and $\hat{\alpha}_2 = \frac{2}{N} \sum_{n=n_0}^{n_0+N-1} x[n] \sin(2\pi \hat{f}_0(n - n_0))$
- Then, given a frequency and arrival time:

$$\hat{A} = \sqrt{\hat{\alpha}_1^2 + \hat{\alpha}_2^2}$$

$$\hat{\phi} = \arctan\left(\frac{-\hat{\alpha}_1}{\hat{\alpha}_2}\right)$$

- Our likelihood ratio is still the periodogram, but for a window $[n_0, n_0 + N - 1]$

Unkown Amplitude, Phase, Frequency, and Arrival Time

- If we know the arrival time, we can use the same maximum as in the three paramter case.
- So, starting at $n_0 = 0$ we have to perform a frequency analysis and find the maximum frequency. Then set $n_0 = 1$ and find the same thing. Plot this for every n_0 and find the maximum frequency. This is called the **short time periodogram** (or **short time FFT**).
- This can be computed with the **spectrogram** command in matlab, and is widely used.

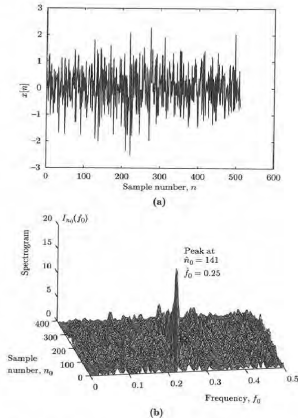


Figure 9: Example of signal and its short term FFT

Comments

- The generalized likelihood ratio test is a combination of MLE and the NP test for simple hypotheses.
- As more parameters are unknown, our detection performance generally goes down - same thing happened in estimation.
- As more parameters are unknown our MLE is more complicated.
- This example featured concentrated likelihood - i.e. we could express \hat{A} and $\hat{\phi}$ in terms of \hat{f}_0 and \hat{n}_0 .
- The Short Time Fourier Transform is commonly used in signal processing, and particularly in audio engineering.