

# Signal and Image Processing

## Lab session 4

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### 1 Signal windowing

A signal of finite length  $N$  corresponds to the multiplication of the signal with a rectangular window of length  $N$ , denoted by  $w_r$ . In the Fourier domain, this corresponds to the convolution with the Fourier transform of the rectangular window. Alternatively, we consider the Hamming window of length  $N$ .

$$w_h(n) = \begin{cases} 0.54 - 0.46 \cos\left(\frac{2\pi n}{N}\right) & \text{if } n \in \{0, \dots, N-1\} \\ 0 & \text{otherwise} \end{cases}$$

As the Dirac distribution is the neutral operator for the convolution, smaller distortions are obtained when the Fourier transform of the window is closer to the Dirac distribution.

- 1.1 Assuming the number of frequency bins is  $L = 1024$  and considering  $N = 10, N = 100, N = 200$ , write a program that plots the modulus of the Fourier transform of the windows  $w_r$  and  $w_h$  at frequency locations  $\frac{k}{L}$  for  $-\frac{L}{2} \leq k \leq \frac{L}{2} - 1$ .

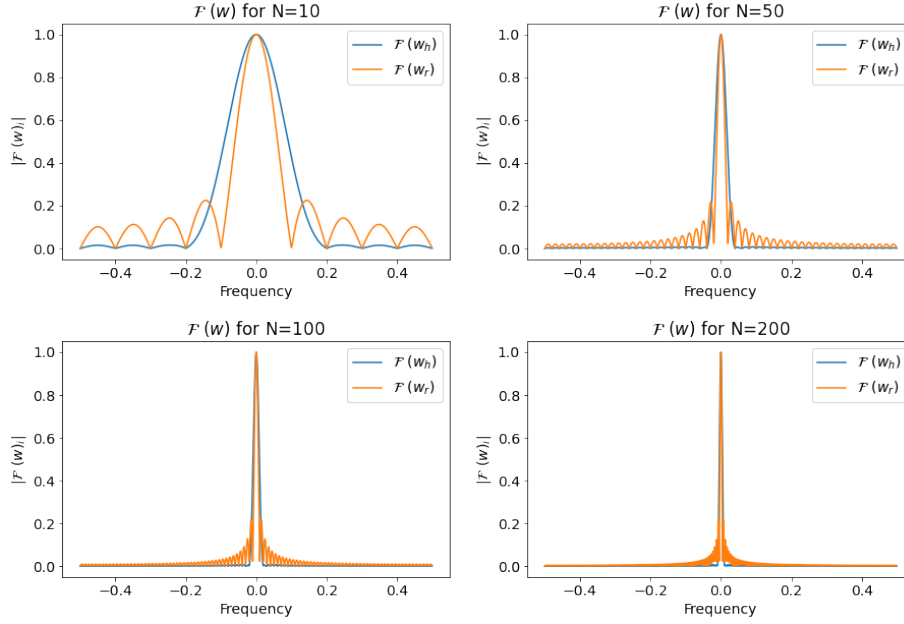


Figure 1: Comparison of the modulus of Fourier transform of  $w_r$  and  $w_h$  for  $N = 10, 50, 100, 200$ .  $N=50$  is included because a 2x2 grid of figures looks better.

## 1.2 What is the effect of $N$ on the frequency representation?

Both transforms converge to the Dirac distribution. In all figures,  $\hat{w}_r$  goes to zero faster than  $\hat{w}_h$ .  $\hat{w}_h$  does not have the large oscillations that  $\hat{w}_r$  does. As  $N$  increases, the oscillations of  $\hat{w}_r$  do not go away but instead move closer to 0. Their amplitude remains large even for large  $N$ . By analogy, this is similar to an under-damped harmonic oscillator versus a critically damped oscillator. The critically damped oscillator behaves like  $\hat{w}_h$  as it goes away from frequency 0 in that it goes to the baseline much slower than the underdamped system, but its oscillations are much lower. The underdamped oscillator goes to zero quickly but overshoots and takes a long time to settle to 0.

## 2 Low Pass signal filtering

In this exercise, we would like to build a finite impulse response filter  $h$  whose Fourier transform approximates  $\mathbf{1}_{[-f_0, f_0]}$ , the indicator function of interval  $[-f_0, f_0]$  for  $f_0 < 1/2$ .

**2.1 Assuming  $\mathbf{1}_{[-f_0, f_0]}$  is periodized into a 1-periodic function, compute the Fourier coefficients of the function, denoted by  $h_n$ .**

$$h(\lambda) = \sum_{n \in \mathbb{Z}} h_n e^{2i\pi n \lambda} \approx \sum_{m=-\frac{N-1}{2}}^{\frac{N-1}{2}} h_n e^{2i\pi m \lambda}$$

$$\begin{aligned} h_n &= \int_{-\frac{1}{2}}^{\frac{1}{2}} \mathbf{1}_{[-f_0, f_0]}(\lambda) e^{-2i\pi n \lambda} d\lambda = \int_{-f_0}^{f_0} e^{-2i\pi n \lambda} d\lambda = -\frac{1}{2i\pi n} (e^{-2i\pi n f_0} - e^{2i\pi n f_0}) = \\ &= \frac{\sin(2\pi n f_0)}{\pi n} = 2f_0 \text{sinc}(2\pi n f_0) \end{aligned}$$

**2.2 As in the previous lab session, we truncate the  $h_n$  and keep only  $N$  coefficients, and we adopt the same rule as in the previous lab session to obtain  $\tilde{h}_n$  when  $n$  is even.**

$$\begin{aligned} h(\lambda) &\approx \sum_{-N/2+1}^{N/2} h_n e^{2i\pi n \lambda} \approx \sum_{-N/2+1}^{N/2} \tilde{h}_n e^{2i\pi(n-1/2)\lambda} \\ \hat{h}_n &= \int_{-\frac{1}{2}}^{\frac{1}{2}} \mathbf{1}_{[-f_0, f_0]}(\lambda) e^{-i\pi(2n-1)\lambda} d\lambda = \int_{-f_0}^{f_0} e^{-i\pi(2n-1)\lambda} d\lambda = \\ &= -\frac{1}{i\pi(2n-1)} (e^{-i\pi(2n-1)f_0} - e^{i\pi(2n-1)f_0}) = \frac{2}{\pi(2n-1)} \sin(\pi(2n-1)f_0) \end{aligned}$$

**2.3 Multiply  $h$  with the Hamming window of length  $N$  to obtain  $g$ , this should be done by writing a function  $g = \text{FIR}(f_0, N)$ .**

$$\begin{aligned} g(n) &= h(n)w_h(n + N/2) \\ \hat{h}_n w_h(n) &= \frac{2}{\pi(2n-1)} \sin(\pi(2n-1)f_0) \left( 0.54 - 0.46 \cos\left(\frac{2\pi(n + \frac{N}{2})}{N}\right) \right) \\ \text{for } -\frac{N}{2} &\leq n \leq \frac{N}{2} - 1 \quad \text{when } N \text{ is even} \end{aligned}$$

$$h_n w_h(n) = \frac{\sin\left(2\pi n f_0\right)}{\pi n} \left(0.54 - 0.46 \cos\left(\frac{2\pi(n + \frac{N-1}{2})}{N}\right)\right)$$

for  $-\frac{N-1}{2} \leq n \leq \frac{N-1}{2}$  when  $N$  is odd

**2.4 Compute the modulus of the DFT of  $h$  and  $g$  and conclude which filter is better, namely  $h$  or  $g$ .**

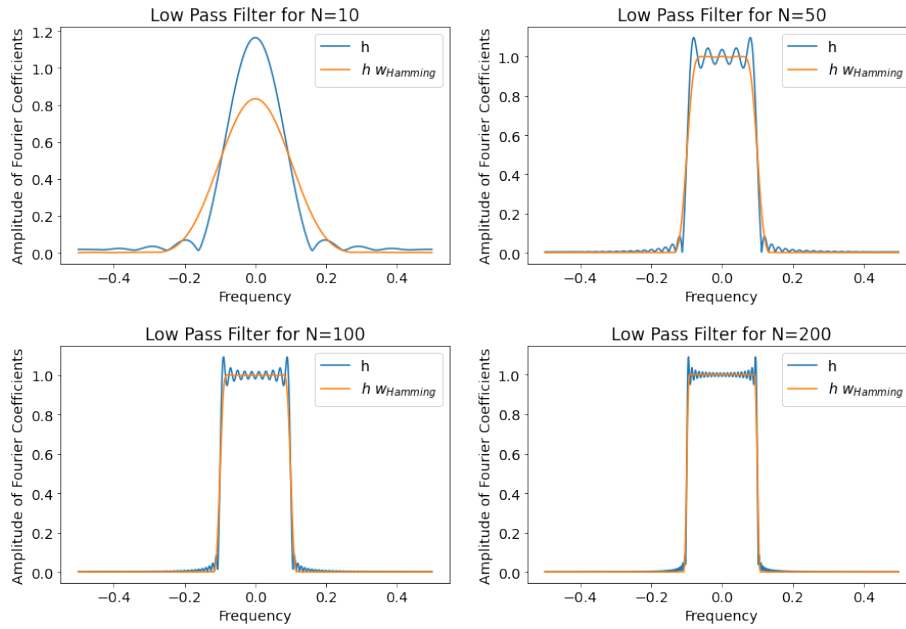


Figure 2: Comparison of DFT of an approximation of a Low pass filter using  $h, g$  for  $N = 10, 50, 100, 200$

### 3 Band Pass signal filtering

In this exercise, we would like to build a finite impulse response filter  $h$  whose Fourier transform approximates  $\mathbf{1}_{[-f_0-f_1, f_0-f_1]} + \mathbf{1}_{[-f_0+f_1, f_0+f_1]}$  for some  $f_0 < 1/2$  and  $f_1 > f_0$  and  $f_1 + f_2 < 1/2$ .

**3.1** If we denote  $H$  by the Fourier transform of a sequence  $h$ , show that the Fourier transform of the sequence  $2h_n \cos(2\pi f_1 n)$  is  $H(\xi - f_1) + H(\xi + f_1)$ .

$$\sum_{n \in \mathbb{Z}} h_n e^{-2i\pi \xi n} = H(\xi)$$

$$\begin{aligned} \sum_{n \in \mathbb{Z}} 2h_n \cos(2\pi f_1 n) e^{-2i\pi \xi n} &= \sum_{n \in \mathbb{Z}} h_n (e^{2i\pi f_1 n} + e^{-2i\pi f_1 n}) e^{-2i\pi \xi n} = \\ &= \sum_{n \in \mathbb{Z}} h_n (e^{-2i\pi(\xi - f_1)n} + e^{-2i\pi(\xi + f_1)n}) = H(\xi - f_1) + H(\xi + f_1) \end{aligned}$$

**3.2** Modify the program of the previous question to obtain filters of size  $N$  with Fourier transform approximating  $1_{[-f_0 - f_1, f_0 - f_1]} + 1_{[-f_0 + f_1, f_0 + f_1]}$ . Deal with the cases  $N$  odd and even, and plot the modulus of the corresponding discrete Fourier transform. Numerical application:  $f_0 = 1/8, f_1 = 1/4$ . What is the role of such a filter?

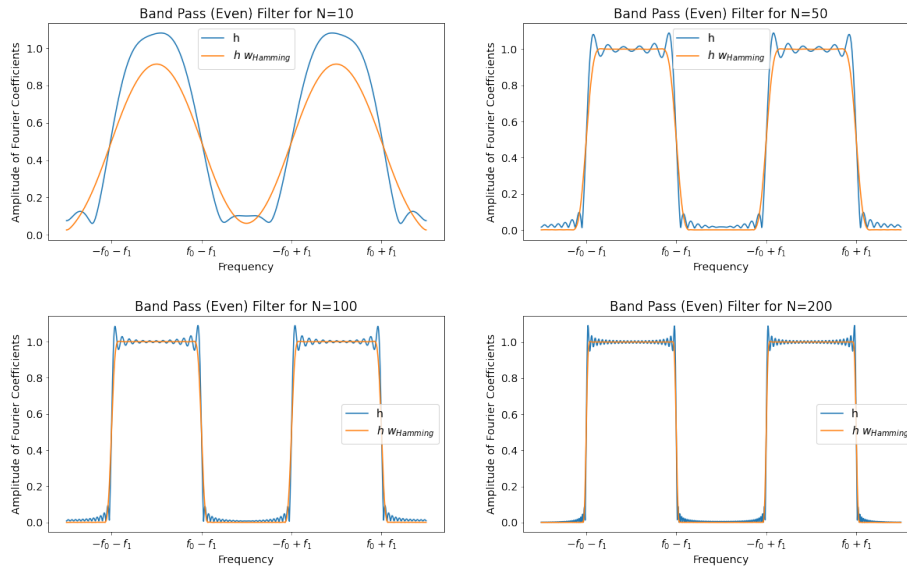


Figure 3: Fourier transform of a approximations of a band pass filter using  $h$  and  $h_{hamming}$  for even values  $N = 10, 50, 100, 200$

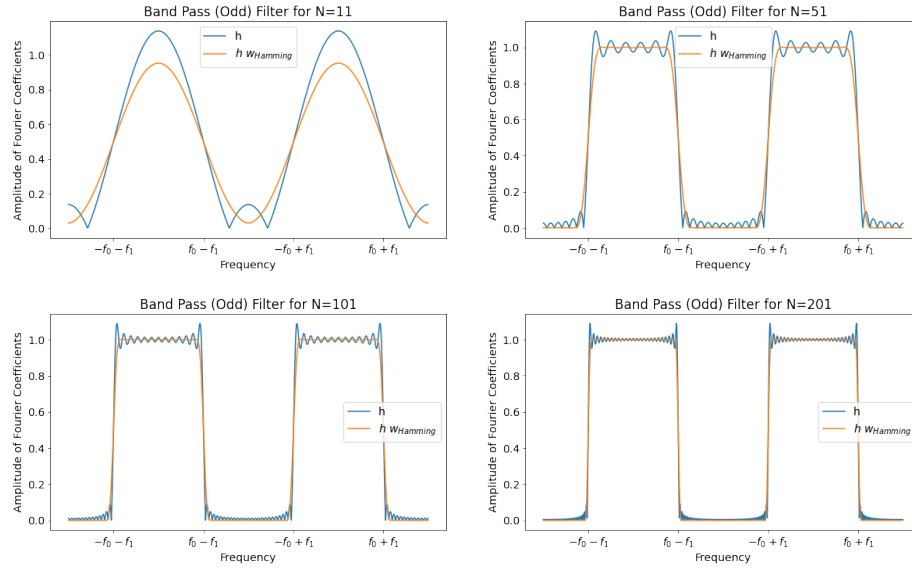


Figure 4: Fourier transform of a approximations of a band pass filter using  $h$  and  $h_{\text{hamming}}$  for odd values  $N = 11, 51, 101, 201$

This filter cancels out too high and too low frequencies, ignoring high and low frequency noise. One example of a use case would be in cleaning a signal that has some long slow oscillation in its baseline, possibly from some physical shaking, and a high frequency noise component. This would depend on the frequency representation of the signal not having large amplitudes in the regions that would be set to zero.