

# Fundamentals of Signal Processing and Data Analysis

## Homework 4

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### 1 Frequency Response of a Rectangular Window Filter

Suppose the impulse response of a digital filter is defined by

$$h(n) = \begin{cases} 1, & 0 \leq n \leq N-1 \\ 0, & \text{otherwise} \end{cases}$$

Find its frequency response:

$$\begin{aligned} H(e^{j\omega}) &= \sum_{n=-\infty}^{\infty} h(n)e^{-j\omega n} = \sum_{n=0}^{N-1} e^{-j\omega n} \\ &= \frac{1 - e^{-j\omega N}}{1 - e^{-j\omega}} \\ &= \frac{e^{-j\omega N/2} (e^{j\omega N/2} - e^{-j\omega N/2})}{e^{-j\omega/2} (e^{j\omega/2} - e^{-j\omega/2})} \\ &= e^{-j\omega \frac{N-1}{2}} \cdot \frac{\sin(\omega N/2)}{\sin(\omega/2)} \end{aligned}$$

Magnitude response:

$$|H(e^{j\omega})| = \frac{\sin(\omega N/2)}{\sin(\omega/2)}$$

### 2 Trigonometric Form of the Fourier Series

Deduce the trigonometric Fourier series of  $x(t)$ :

$$\begin{aligned}
x(t) &= \sum_{k=-\infty}^{\infty} C_k e^{ik\omega_0 t} \\
&= C_0 + \sum_{k=1}^{\infty} (C_k e^{ik\omega_0 t} + C_{-k} e^{-ik\omega_0 t}) \\
&= C_0 + \sum_{k=1}^{\infty} [(C_k + C_{-k}) \cos(k\omega_0 t) + i(C_k - C_{-k}) \sin(k\omega_0 t)]
\end{aligned}$$

Since  $x(t) \in \mathbb{R}$ , it follows:

$$C_k + C_{-k} \in \mathbb{R}, \quad C_k - C_{-k} \in \mathbb{I}, \quad C_{-k} = \overline{C_k}$$

Thus:

$$\begin{aligned}
x(t) &= C_0 + 2 \sum_{k=1}^{\infty} (\operatorname{Re} C_k \cos(k\omega_0 t) + i \operatorname{Im} C_k \sin(k\omega_0 t)) \\
&= C_0 + 2 \sum_{k=1}^{\infty} (\operatorname{Re} C_k \cos(k\omega_0 t) - \operatorname{Im} C_k \sin(k\omega_0 t)) \\
&= C_0 + 2 \sum_{k=1}^{\infty} (|C_k| \cos \angle C_k \cos(k\omega_0 t) - |C_k| \sin \angle C_k \sin(k\omega_0 t)) \\
&= C_0 + 2 \sum_{k=1}^{\infty} |C_k| \cos(k\omega_0 t + \angle C_k)
\end{aligned}$$

### 3 Gibbs Phenomenon Demonstration

#### 3.1 Theoretical Proof of the Gibbs Phenomenon

We consider the function:

$$f(t) = \begin{cases} 1 & -\frac{1}{2} \leq t \leq \frac{1}{2} \\ 0 & \text{otherwise} \end{cases}$$

The Fourier series is:

$$f_N(t) = \frac{1}{2} + \sum_{n=0}^N \frac{2(-1)^n}{(2n+1)\pi} \cos((2n+1)\pi t)$$

Let:

$$f_N(t) = \frac{1}{2} + g_N(t)$$

Then the derivative of  $g_N(t)$  is:

$$g'_N(t) = \sum_{n=0}^N 2(-1)^{n+1} \sin((2n+1)\pi t)$$

$$\begin{aligned}
&= 2 \sum_{n=0}^N \sin((2n+1)\pi t + (n+1)\pi) \\
&= 2 \operatorname{Im} \sum_{n=0}^N e^{i(2\pi t + \pi)n} \cdot e^{i(\pi t + \pi)} \\
&= 2 \operatorname{Im} \left( \frac{1 - e^{i(2\pi t + \pi)(N+1)}}{1 - e^{i(2\pi t + \pi)}} \cdot e^{i(\pi t + \pi)} \right) \\
&= 2 \operatorname{Im} \left( \frac{e^{i\frac{(2\pi t + \pi)(N+1)}{2}}}{e^{i\frac{(2\pi t + \pi)}{2}}} \cdot \frac{e^{-i\frac{(2\pi t + \pi)(N+1)}{2}} - e^{i\frac{(2\pi t + \pi)(N+1)}{2}}}{e^{-i\frac{(2\pi t + \pi)}{2}} - e^{i\frac{(2\pi t + \pi)}{2}}} \cdot e^{i(\pi t + \pi)} \right) \\
&= 2 \operatorname{Im} \left( e^{i\left(\frac{(2\pi t + \pi)(N+1)}{2} + \frac{\pi}{2}\right)} \cdot \frac{\sin\left(\frac{(2\pi t + \pi)(N+1)}{2}\right)}{\sin\left(\frac{2\pi t + \pi}{2}\right)} \right) \\
&= 2 \cdot \frac{\sin\left(\left(\pi t + \frac{\pi}{2}\right)(N+1) + \frac{\pi}{2}\right) \cdot \sin\left(\left(\pi t + \frac{\pi}{2}\right)(N+1)\right)}{\sin\left(\frac{2\pi t + \pi}{2}\right)} = 0
\end{aligned}$$

Then:

$$\begin{aligned}
&\sin\left(\left(\pi t + \frac{\pi}{2}\right)(N+1) + \frac{\pi}{2}\right) \cdot \sin\left(\left(\pi t + \frac{\pi}{2}\right)(N+1)\right) = 0 \\
&\cos\left(\left(\pi t + \frac{\pi}{2}\right)(N+1)\right) \cdot \sin\left(\left(\pi t + \frac{\pi}{2}\right)(N+1)\right) = 0 \\
&\sin((2\pi t + \pi)(N+1)) = 0 \\
&(2\pi t + \pi)(N+1) = h\pi \quad \Rightarrow \quad t = \frac{h}{2(N+1)} - \frac{1}{2}
\end{aligned}$$

Gibbs phenomenon appears at  $t = \pm \frac{1}{2}$ . For instance, take  $h = 1$ , such that  $t \rightarrow -\frac{1}{2}$  then:

$$t = \frac{1}{2(N+1)} - \frac{1}{2}$$

Evaluate the overshoot:

$$\begin{aligned}
f_N(t) &= \frac{1}{2} + \sum_{n=0}^N \frac{2(-1)^n}{(2n+1)\pi} \cos\left((2n+1)\pi \left(\frac{1}{2(N+1)} - \frac{1}{2}\right)\right) \\
&= \frac{1}{2} + \sum_{n=0}^N \frac{2(-1)^n}{(2n+1)\pi} \cos\left(\frac{(2n+1)\pi}{2(N+1)} - n\pi - \frac{\pi}{2}\right) \\
&= \frac{1}{2} + \sum_{n=0}^N \frac{2(-1)^n}{(2n+1)\pi} \sin\left(\frac{(2n+1)\pi}{2(N+1)} - n\pi\right) \\
&= \frac{1}{2} + \sum_{n=0}^N \frac{2(-1)^{2n}}{(2n+1)\pi} \sin\left(\frac{(2n+1)\pi}{2(N+1)}\right) \\
&= \frac{1}{2} + \sum_{n=0}^N \frac{\sin\left(\frac{n\pi}{N+1} + \frac{\pi}{2(N+1)}\right)}{n\pi + \frac{\pi}{2}}
\end{aligned}$$

$$= \frac{1}{2} + \frac{1}{\pi} \sum_{n=0}^N \frac{\sin\left(\frac{n\pi}{N+1} + \frac{\pi}{2(N+1)}\right)}{\frac{n\pi}{N+1} + \frac{\pi}{2} \cdot \frac{1}{N+1}} \cdot \frac{\pi}{N+1}$$

As  $N \rightarrow \infty$ , the sum approaches:

$$f_N(t) \rightarrow \frac{1}{2} + \frac{1}{\pi} \int_0^\pi \frac{\sin x}{x} dx = \frac{1}{2} + \frac{1}{\pi} \text{Si}(\pi)$$

Therefore, the overshoot ratio is:

$$\eta = \frac{\frac{1}{2} + \frac{1}{\pi} \text{Si}(\pi) - 1}{1} = \frac{\text{Si}(\pi)}{\pi} - \frac{1}{2} \approx 0.0895 = 8.95\%$$

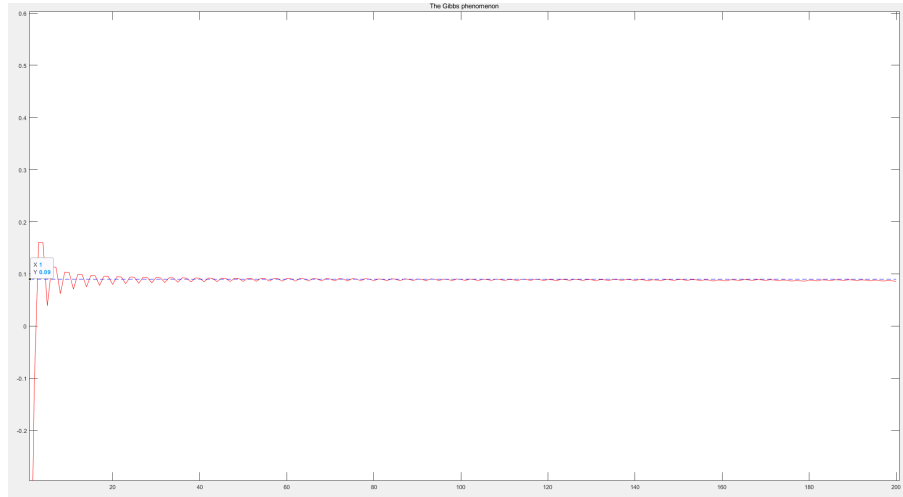


Figure 1: The Gibbs Phenomenon

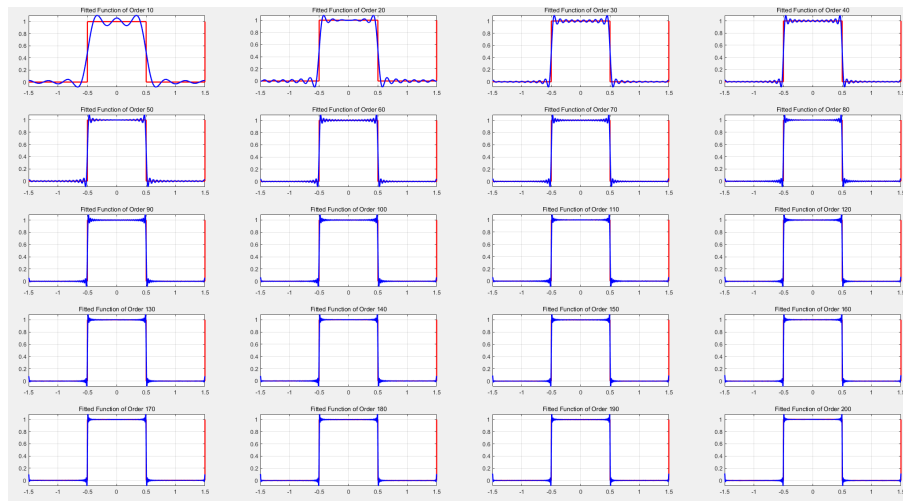


Figure 2: Fourier Approximation of Square Wave

## 3.2 MATLAB Code for Fourier Series Approximation

Reference the file titled "Fourier Transform Pre.pdf".

```
1 %% Variable allocation
2 n = 200; % Fitting order
3 maxgibbis = zeros(1, n); % Maximum value of the fitting function
   in each iteration
4 a = -1.5; b = 1.5; % Define the interval
5 N = 3e3; delta = (b - a) / N; x_lin = transpose(a:delta:b); %
   Sample points
6 N = N + 1;
7
8 % Basis functions: cosine
9 createfun = @(ii, x) cos((ii - 1) * 2 * pi / (b - a) * (x - (b +
   a) / 2));
10
11 vec_fun = zeros(N, n);
12 for ii = 1:n
13     vec_fun(:, ii) = createfun(ii, x_lin);
14 end
15
16 % Square wave target function
17 orifun = @(x) 0.5 + 0.5 * square(pi * (x + 0.5));
18 ori_fun = orifun(x_lin);
19
20 % Gram-Schmidt orthogonalization
21 normal_fun = vec_fun;
22 coeff = sum(delta .* normal_fun(:, 1).^2);
23 normal_fun(:, 1) = normal_fun(:, 1) / sqrt(coeff);
24
25 for ii = 2:n
26     for jj = 1:ii-1
27         coeff = sum(normal_fun(:, jj) .* normal_fun(:, ii)) *
           delta;
28         normal_fun(:, ii) = normal_fun(:, ii) - coeff *
           normal_fun(:, jj);
29     end
30     coeff = sum(delta .* normal_fun(:, ii).^2);
31     normal_fun(:, ii) = normal_fun(:, ii) / sqrt(coeff);
32 end
33
34 % Fitting process
35 coeff = zeros(n, 1);
36 fit_fun = zeros(N, 1);
37 figure; rows = ceil(n / 40); cols = 4;
38
39 for ii = 1:n
40     coeff(ii) = sum(ori_fun .* normal_fun(:, ii)) * delta;
41     fit_fun = fit_fun + coeff(ii) * normal_fun(:, ii);
42     maxgibbis(ii) = max(fit_fun);
43     if mod(ii, 10) == 0
```

```

44     subplot(rows, cols, ii / 10);
45     plot(x_lin, ori_fun, 'r', 'LineWidth', 2);
46     hold on;
47     plot(x_lin, fit_fun, 'b', 'LineWidth', 2);
48     title(['Fitted Function of Order ' num2str(ii)]);
49     grid on;
50     end
51 end
52
53 % Gibbs phenomenon plot
54 figure;
55 plot(1:n, maxgibbis - 1, 'r'); hold on;
56 y_constant = 0.09;
57 plot([1, n], [y_constant, y_constant], 'b--');
58 title('The Gibbs Phenomenon');

```