#### Constructor University Bremen

### Lab Report 2: The Fourier Series and The Fourier Transform

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#### 1. Prelab

#### 1.1 Problem 1: Decibels

- 1. Given  $x(t) = 5\cos(2\pi 1000t)$ 
  - What is the amplitude and the  $V_{pp}$  of the signal? The amplitude of the signal is 5V and the  $V_{pp}$  is  $2 \cdot 5V = 10V$ .
  - What is the  $V_{rms}$  of the signal? The  $V_{rms}$  of a sinusoidal signal is given by

$$V_{rms} = \frac{V_A}{2\sqrt{2}} = \frac{5V}{\sqrt{2}} = 3.535V \tag{1.1}$$

• What is the amplitude of the spectral peak in dBVrms? The amplitude of the spectral peak in dBVrms is given by the following relation:

$$A_{dBV_{rms}} = 20 \log_{10} (V_{rms}) = 20 \log_{10} (3.535V) = 10.9 dBV_{rms}$$
(1.2)

- 2. Given a square wave of  $1V_{pp}$  the voltage changes between -0.5V to 0.5V
  - What is the signal amplitude in  $V_{rms}$ ? To find the RMS value of a square wave, the following relation is used:

$$V_{rms} = \sqrt{\frac{1}{T} \left[ \int_0^T f(t)^2 dt \right]}$$
 (1.3)

The square wave can be specified as a piecewise function:

$$f(t) = \begin{cases} -0.5 & \text{if } 0 \le t \le \frac{T}{2} \\ 0.5 & \text{if } \frac{T}{2} \le t \le T \end{cases}$$
 (1.4)

Performing the integral over the function,

$$V_{rms} = \sqrt{\frac{1}{T} \left[ \int_0^{T/2} (-0.5)^2 dt + \int_{T/2}^T (0.5)^2 dt \right]}$$

$$= \sqrt{0.25 \cdot \frac{1}{T} \left( \frac{T}{2} + T - \frac{T}{2} \right)}$$

$$= \sqrt{0.25}V$$

$$= 0.5V$$
(1.5)

• What is the amplitude in dBVrms?

$$V_{dB_{rms}} = 20 \log_{10} (V_{rms}) = 20 \log_{10} (0.5V) = -6.02 dB V_{rms}$$
 (1.6)

# 1.2 Problem 2: Determination of Fourier Series Coefficients

1. Determine the Fourier series coefficients up to the 5th harmonic of the function  $f(t) = 4t^2$ 

To determine the fourier series coefficients, first recognize that the function is an even function by the very fact that it is a polynomial of an even degree. Knowing this, the relations used to compute the fourier series coefficients are:

$$a_n = \frac{4}{T} \int_{-T}^{T} t^2 \cos(n\omega_0 t) dt$$

$$a_0 = \frac{1}{T} \int_{0}^{T} t^2 dt$$

$$b_n = 0$$

$$(1.7)$$

Where it is known that T=1 due to the interval being given as [-0.5, 0.5]. Furthermore,  $\omega_0 = \frac{2\pi}{T} = 2\pi$ . By using the help of an integration table, the integral for  $a_n$  is

$$a_n = 4 \cdot \left[ \frac{2t \cos(n\omega_0 t)}{4\pi^2 n^2} + \frac{n^2 4\pi^2 - 2}{8\pi^3 n^3} sin(n\omega_0 t) \right]_{-T}^T$$
 (1.8)

Evaluating this integral:

$$a_n = 4 \cdot \left(\frac{4 \cdot (-1)^n}{4\pi^2 n^2}\right) = \frac{4 \cdot (-1)^n}{\pi^2 n^2}$$
 (1.9)

For  $a_0$ , the integral is straightforward:

$$a_0 = \frac{1}{T} \int_0^T t^2 dt = \left[\frac{t^3}{3}\right]_0^T = \frac{1}{3}$$
 (1.10)

So the first five coefficients are:

$$a_n = \left\{ \frac{-4}{\pi^2}, \frac{4}{16\pi^2}, \frac{-4}{36\pi^2}, \frac{4}{64\pi^2}, \frac{-4}{100\pi^2} \right\}$$
 (1.11)

2. Use MATLAB to plot the original function and the inverse Fourier transform. Put both graphs into the same diagram.

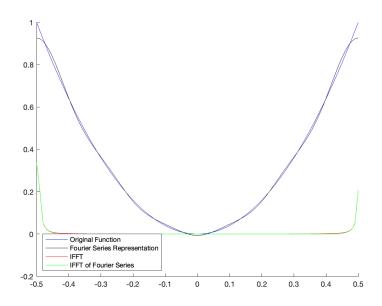


Figure 1.1: The plot of the original function, its inverse fourier transform, the fourier series representation up to the fifth harmonic, and the representation's inverse fourier transform

The code used:

```
t = -0.5:0.01:0.5;
f = 4*t.^2;
f_fs = 1/3;
% Fourier series representation of the function
for i=1:5
    % w_0 = 2pi
    % T = 1
    a_i = (4*(-1)^i)/(pi^2*i^2);
    f_fs = f_fs + (a_i .* cos(i.*2.*pi.*t));
end
hold on
plot(t, f, "blue");
plot(t, f_fs, "black");
xlim([-0.5, 0.5]);
plot(t, abs(ifft(f)), "red");
plot(t, abs(ifft(f_fs)), "green");
legend({"Original Function", ...
    "Fourier Series Representation", ...
    "IFFT", ...
    "IFFT of Fourier Series" ...
    }, "Location", "southwest")
```

#### 1.3 Problem 3: FFT of a Rectangular Wave

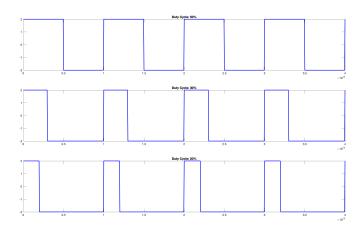


Figure 1.2: A figure showing the different rectangular waves based on their duty cycles. Plotted using MATLAB.

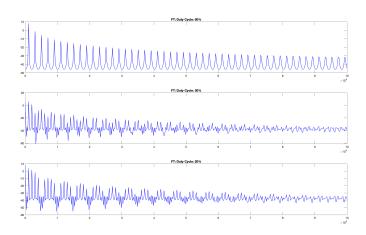


Figure 1.3: A figure showing the different rectangular waves' spectra based on their duty cycles. Plotted using MATLAB.

The code used to plot the figures:

```
%% Time plots
[t_fifty, signal_fifty, f_fifty, y_fifty] = get_square(50);
[t_thirty, signal_thirty, f_thirty, y_thirty] = get_square(30);
[t_twenty, signal_twenty, f_twenty, y_twenty] = get_square(20);
subplot(3, 1, 1);
plot(t_fifty, signal_fifty, "blue", "LineWidth", 2);
title("Duty Cycle: 50%");
subplot(3, 1, 2);
plot(t_thirty, signal_thirty, "blue", "LineWidth", 2);
title("Duty Cycle: 30%");
subplot(3, 1, 3);
plot(t_twenty, signal_twenty, "blue", "LineWidth", 2);
title("Duty Cycle: 20%");
%% Frequency plots
subplot(3, 1, 1);
plot(f_fifty, y_fifty, "blue", "LineWidth", 1);
title("FT: Duty Cycle: 50%");
subplot(3, 1, 2);
plot(f_thirty, y_thirty, "blue", "LineWidth", 1);
title("FT: Duty Cycle: 30%");
subplot(3, 1, 3);
plot(f_twenty, y_twenty, "blue", "LineWidth", 1);
title("FT: Duty Cycle: 20%");
```

Where the function get\_square is defined as:

```
function [t, signal, f, y_single_db] = get_square(duty_cycle)
    period = 1e-3;
    Fs = 200e3;
    frequency = 1/period;
    Vpp = 2;
    duration = period * 4;
    % The signal
    t = 0:1/Fs:duration;
    signal = Vpp*square((2*pi*frequency)*t, duty_cycle);
    plot(t, signal, "blue", "LineWidth", 2);
    % The fourier transform of the signal
    rms_value = sqrt(mean(signal.^2));
    N = length(signal); % The length of the signal
    y = fft(signal, N);
    y_mag = 2*(abs(y)/N); % Magnitudes of y
    y_single = y_mag(1:floor(N/2)) * 2;
    f_nyquist = Fs / 2;
    y_single_db = 20*log10(y_single / rms_value);
    f = linspace(0, f_nyquist, length(y_single));
end
```

Following by the equation:

$$c_k = \frac{1}{k\pi} \sin(k\omega_0 T_1) \tag{1.12}$$

It is observed that for a lower  $T_1$ , for each frequency component the amplitude of the component is lower. This is due to the fact that the pulse width is lower, and thus the signal is more spread out in the time domain. This is also observed in the frequency domain, where the frequency components are more spread out, thereby more frequency components are required to represent the signal.

#### 1.4 Problem 4: FFT of an audio file

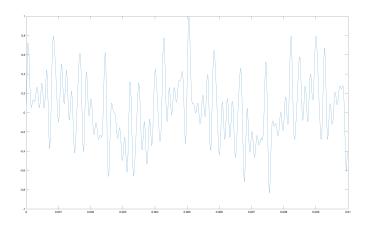


Figure 1.4: A figure showing the plot of the audio file

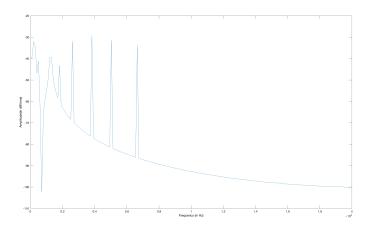


Figure 1.5: A figure showing the frequency spectrum of the audio file

The following code was used for the plots:

```
%% Problem 4:
[y, Fs] = audioread("s_samp.wav");
N_samples = Fs * 10E-3;
y = y(1:N_samples);
t = 0:1/Fs:((10E-3)-1/Fs);
plot(t, y);
%%
Y = fft(y);
rms = sqrt(mean(abs(Y.^2)));
Fs_nyquist = Fs / 2;
Y_{single} = 2 * abs(Y) / N_{samples};
Y_single = Y_single(1:floor(N_samples/2));
f = linspace(0,Fs_nyquist,length(Y_single));
Y_db = 20*log10(Y_single ./ rms);
plot(f, Y_db);
ylabel("Amplitude(in dBVrms)")
xlabel("Frequency (in Hz)");
xlim([0, 20000]);
   The following tones are observed:
   • G3 - 201Hz
   • B4 - 503Hz
   • E6 - 1308Hz
   • A6 - 1812Hz
   • E7 - 2617Hz
   • A#7 - 3826Hz
   • D#8 - 5034Hz
   • G#8 - 6645Hz
```

## 2. Introduction

### 3. Execution

### 4. Evaluation

### 5. Conclusion

### 6. References

## 7. Appendix