Department of Electrical Engineering Indian Institute of Technology Kharagpur

Digital Signal Processing Laboratory (EE39203)

Autumn, 2022-23

Expe:	riment 7	': Fast Fourier T	`rans	forn	a
Slot	X	Date	· 13	/10/	/2023

Student Name: P. Manoj Kumar Roll No.: 21IE10027

Grading Rubric

	Tick the best applicable per row			
	Below	Lacking	Meets all	Points
	Expectation	in Some	Expectation	
Completeness of the report				
Organization of the report (5 pts) With cover sheet, answers are in the same order as questions in the lab, copies of the questions are included in report, prepared in LaTeX				
Quality of figures (5 pts) Correctly labelled with title, x-axis, y-axis, and name(s)				
Understanding of the frequency range of DFT and effects of zero-padding (35 pts) DFT and DTFT plots, Matlab code (DTFT samples), questions				
Implementation of Divide-and- Conquer DFT and FFT (40 pts) Matlab codes (dcDFT, fft2, fft4, fft8, fft stages), questions				
Computation time comparison (15 pts) Runtimes, questions				
TOTAL (100 pts)				

Total Points (100):	TA Name:	TA Initials:

Digital Signal Processing Laboratory (EE39203)

P Manoj Kumar (21IE10027)

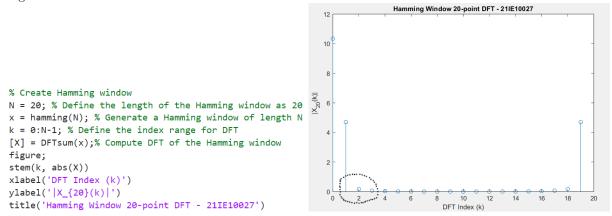
Experiment 7 - Fast Fourier Transform

1 Introduction

1.1 Shifting the Frequency Range

1.1.1 Using DFTSum

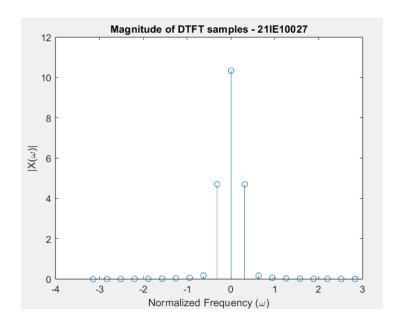
The MATLAB Code and the magnitude plot of DTFT of Hamming Window for 20 points using DFTsum is given below.



The rounded part in the above figure corresponds to the low frequency range.

1.1.2 Using DTFTSamples

The MATLAB Code and the magnitude plot of DTFT of Hamming Window for 20 points using DTFTsamples is given below.



1.2 Zero Padding

1.2.1 Plot of DTFT Samples for N=50 and N=200

The MATLAB Code and the magnitude plot of DTFT of $sin(0.1\pi n)$ for N=50 and N=200 is given below.

```
DTFT samples for N = 50 - 21IE10027
                                                      20
                                                      15
% Define parameters
                                                      10
N1 = 50; N2 = 200;
% Step 1: Compute the vector x
n = 0:N1-1; x n1 = sin(0.1*pi*n);
n = 0:N2-1;x_n2 = sin(0.1*pi*n);
% Step 2: Compute DTFT samples
                                                               -3
                                                                       -2
                                                                               -1
                                                                                        0
[X1, w1] = DTFTsamples(x_n1);
                                                                             Frequency (ω in rad/sample)
[X2, w2] = DTFTsamples(x_n2);
                                                                         DTFT samples for N = 200 - 21 E10027
                                                     100
subplot(2,1,1)
stem(w1, abs(X1))
                                                      80
xlabel('Frequency (\omega in rad/sample)')
ylabel('|X_{50}(\omega)|')
                                                     60
title('DTFT samples for N = 50
                                 - 21IE10027')
                                                     40
subplot(2,1,2)
                                                      20
stem(w2, abs(X2))
xlabel('Frequency (\omega in rad/sample)')
ylabel('|X_{200}(\omega)|')
title('DTFT samples for N = 200 - 21IE10027')
                                                                             Frequency (ω in rad/sample)
```

1.2.2 Discussion

- The plot for N = 200 looks more like the true DTFT.
- The difference in appearance between the two plots can be attributed to *spectral leakage*, a phenomenon that occurs when the number of points in the Discrete Fourier Transform (DFT) is small. In the case of N=50, the frequency resolution of the resulting DTFT samples is limited. As a result, the DTFT samples may fail to capture fine details in the frequency domain.
- On the other hand, with a larger value of N (such as N=200), the frequency resolution becomes higher. This allows for a more accurate representation of the true DTFT. Additionally, when N=200, zero-padding is employed to interpolate additional frequency points. This technique enhances the sampling of the DTFT, resulting in a smoother and more precise representation of the frequency content of the signal.

2 The Fast Fourier Transform Algorithm

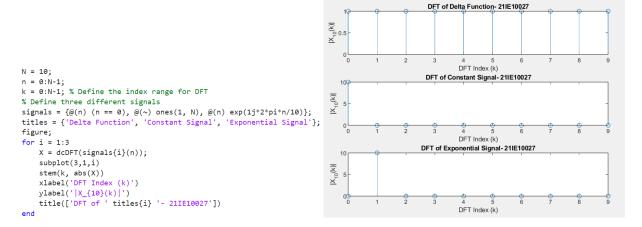
2.1 Implementation of Divide-and-Conquer DFT

2.1.1 MATLAB Code for dcDFT

```
function [X] = dcDFT(x)
   N=length(x);
   j=sqrt(-1);
   % Define indices and split the input signal
   k = 0:N/2-1;
   x0 = x(1:2:N);
   x1 = x(2:2:N);
   % Compute DFT of the even and odd components
   [X0] = DFTsum(x0);
    [X1] = DFTsum(x1);
   % Compute twiddle factors
   W_N = \exp(-j*2*pi.*k/N);
   % Combine the DFT results using twiddle factors
   X_k = X0 + W_N .* X1;
   X_k_2 = X0 - W_N .* X1;
   % Concatenate the DFT results
   X = [X_k X_k_2];
```

2.1.2 Computation of DFT using dcDFT

The MATLAB Code for the magnitude plot of DFT of delta function, constant function and exponential function using dcDFT is given below with the magnitude plots.



2.1.3 Discussion

- 1. Separating samples into even and odd points:
 - No multiplications are performed in this step.
- 2. Computing the two $\frac{N}{2}$ point DFTs (using DFTsum):
 - For each $\frac{N}{2}$ point DFT, there are $\left(\frac{N}{2}\right)^2$ multiplications (since it's a matrix multiplication).
- 3. Multiplying by twiddle factors:
 - There are $\frac{N}{2}$ twiddle factors to compute, each requiring 2 multiplications (real and imaginary parts).
- 4. Combining the two DFTs:
 - For each element of the combined DFT, there are 2 multiplications (real and imaginary parts).

Total number of multiplications:

$$2\left(\left(\frac{N}{2}\right)^2 + \frac{N}{2}\right) = \frac{3}{2}N^2$$

This analysis assumes that the DFTsum function performs the $\frac{N}{2}$ point DFT efficiently. Depending on how DFTsum is implemented, the actual number of multiplications may vary.

2.2 Recursive Divide and Conquer

2.2.1 MATLAB Code for FFT2, FFT4 and FFT8

```
function [X] = FFT8(x)
                            function [X] = FFT4(x)
                                                             N=length(x);
                                N=length(x);
                                                             j=sqrt(-1);
                                j=sqrt(-1);
                                                             k=0:N/2-1;
                                k=0:N/2-1;
                                                             W_N=\exp(-j*2*pi.*k/N);
                                W_N=\exp(-j*2*pi.*k/N);
                                x0=x(1:2:N);
                                                             x0=x(1:2:N);
                                x1=x(2:2:N);
                                                             x1=x(2:2:N);
                                                             [X0]=FFT4(x0);
                                 [X0]=FFT2(x0);
                                                             [X1]=FFT4(x1);
                                 [X1] = FFT2(x1);
function [X] = FFT2(x)
                                                             X_k=X0+W_N.*X1;
                                X k=X0+W N.*X1;
    X(1) = x(1) + x(2);
                                                             X_k_2=X0-W_N.*X1;
                                X_k_2=X0-W_N.*X1;
    X(2) = x(1) - x(2);
                                                             X=[X_k X_k_2];
                                X=[X_k X_k_2];
end
                            end
```

2.2.2 Output of FFT8 for x[n] = 1 for N = 8

```
>> A = FFT8(ones(1,8));
>> A
A =
```

2.2.3 Computation of DFT using FFT8

The MATLAB Code for the magnitude plot of DFT of delta function, constant function and exponential function using FFT8 is given below with the magnitude plots.

- $x[n] = \delta[n]$ for N = 8.
- x[n] = 1 for N = 8.
- $x[n] = e^{j\frac{2\pi n}{8}}$ for N = 8.

```
N = 8;n = 0:N-1;k = 0:N-1;
signals = {\(\emptyred{0}\)(n) \(n = 0\), \(\emptyred{0}\)(n) \(\emptyred{0}\)(
```

FFT8 - 21IE10027

2.2.4 Calculations

- Number of Multiplies with Twiddle Factor in FFT8:
 - In FFT4, there are 2 multiplications for every call. Since FFT4 is called twice, there are a total of 4 multiplications with the twiddle factor.
 - Then, FFT8 is called once, and for each call, 4 multiplications occur. Hence, in total, there will be 8 multiplications with the twiddle factor.
- Number of Multiplies with Twiddle Factor in General FFT (N = 2^p):
 - In the FFT, the number of levels gets reduced to $\log_2(N)$ due to halving at each stage.
 - Each stage (except the last one with 2 elements) will involve $\frac{N}{2}$ multiplications. So, in total, there will be $\left(\frac{N}{2}\right) \cdot (\log_2(N) 1)$ multiplications with the twiddle factor.
- For $N = 2^p$:
 - Number of Multiplies = $p \cdot (\log_2(2p) 1)$
- For $N = 2^p$:
 - Number of Multiplies = $\frac{N}{2} \cdot (p-1)$
- If $N = 2^{10}$ (i.e., p = 10):
 - The number of multiplies is $9 \cdot 2^9 = 4608$.

2.3 Recursive FFT Algorithm for Power-of-2 Length Signal

2.3.1 MATLAB Code for fft_stage

```
function [X] = fft_stage(x)
    N=length(x);
    if (N==2)
       X=FFT2(x);
       return;
    elseif (N>2)
        j=sqrt(-1);
        % Define indices and twiddle factors
        k = 0:N/2-1;
        W_N = \exp(-j*2*pi.*k/N);
        % Split the input signal into even and odd indices
        x0 = x(1:2:N); x1 = x(2:2:N);
        % Recursively apply FFT on even and odd components
        [X0] = fft_stage(x0);[X1] = fft_stage(x1);
        % Combine the FFT results using twiddle factors
               X0 + W_N .* X1; X_k_2 = X0 - W_N .* X1;
        X=[X_k X_k_2];
    end
end
```

2.3.2 Computation of DFT using fft_stage

The MATLAB Code for the magnitude plot of DFT of delta function, constant function and exponential function using fft_stage is given below with the magnitude plots.

```
fft_stage - 21IE10027
                                                                         8-Point FFT of delta function
figure;
for i = 1:3
                                                                             DFT Index (k)
     [X] = fft_stage(signals{i}(n));
                                                                           Point FFT of constan
     subplot(3,1,i)
     stem(k, abs(X))
     xlabel('DFT Index (k)')
                                                                             DFT Index (k)
     ylabel('|X_8(k)|')
     title(['8-Point FFT of
                                  ' titles{i}])
end
sgtitle('fft\_stage - 21IE10027')
                                                                             DFT Index (k)
```

2.3.3 Discussion

• Matching Results:

- The outputs of the recursive FFT implementation (fft_stage) are identical to the results obtained from the non-recursive FFT (FFT8) when applied to the same input signals.

• General Solution:

The recursive approach (fft_stage) is designed to handle signals of various lengths, not just limited to a specific length like the original FFT8 function.

The recursive FFT excels in flexibility, effortlessly handling signals of any power-of-2 length. This versatility outshines the need for separate functions for specific lengths, resulting in leaner and more efficient code. Additionally, its modular structure simplifies maintenance and updates, while its scalability effortlessly processes larger lengths by breaking them into manageable steps. Overall, the recursive FFT is a versatile and efficient tool for computing FFTs across varying signal lengths.