

# **FAST FOURIER TRANSFORM**

**EEEN 462 – ANALOGUE COMMUNICATION**

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# FOURIER TRANSFORM (RECAP)

1. **Fourier Transform** decomposes a function into its constituent frequencies.

$$X(f) = \int_{-\infty}^{\infty} x(t) e^{-j2\pi ft} dt$$

1. **For continuous signals**, we use the Continuous Fourier Transform (CFT).
2. **For discrete signals**, we use the Discrete Fourier Transform (DFT).
3. **Key Insight:** Any periodic signal can be represented as a sum of sinusoids with different frequencies, amplitudes, and phases.

# DISCRETE FOURIER TRANSFORM - RECAP

**1. Discrete-time signals**, we use the DFT:

$$X[k] = \sum_{n=0}^{N-1} x[n] \cdot e^{-j(2\pi/N)kn}, \text{ for } k = 0, 1, \dots, N-1$$

where:

- $x[n]$  is the discrete-time signal (N samples)
- $X[k]$  is the DFT output (frequency bins)
- N is the number of samples
- k represents frequency index

**2. Inverse DFT (IDFT)** reconstructs the time-domain signal from frequency components:

$$x[n] = (1/N) \sum_{k=0}^{N-1} X[k] \cdot e^{j(2\pi/N)kn}, \text{ for } n = 0, 1, \dots, N-1$$

# THE NEED FOR FAST FOURIER TRANSFORM

Direct computation of DFT requires  $O(N^2)$  operations:

- For each of  $N$  frequency bins ( $k$  values), we need  $N$  complex multiplications and  $N-1$  additions
- Total:  $N \times N = N^2$  complex multiplications

N (Samples)	DFT Operations	FFT Operations	Speed-up Factor
64	4,096	384	10.7x
256	65,536	2,048	32x
1024	1,048,576	10,240	102x
4096	16,777,216	49,152	341x

# FFT USES DIVIDE AND CONQUER APPROACH

The key insight behind FFT is the divide-and-conquer strategy as follows:

1. Divide the  $N$ -point DFT into two  $N/2$ -point DFTs
2. Exploit symmetry and periodicity of the complex exponential (twiddle factors)
3. Recursively apply until we reach 2-point DFTs (butterflies)

# RADIX-2 FFT ALGORITHM

- The most common FFT is the **radix-2 Cooley-Tukey algorithm**, which requires  $N$  to be a power of 2.
- Algorithm steps:
  1. If  $N = 1$ , return  $x[0]$  (base case)
  2. Separate  $x[n]$  into even and odd indexed samples:
    - Even:  $x_e[m] = x[2m]$  for  $m = 0, 1, \dots, N/2-1$
    - Odd:  $x_o[m] = x[2m+1]$  for  $m = 0, 1, \dots, N/2-1$
  3. Compute  $N/2$ -point FFT of both sequences:  $X_e[k]$  and  $X_o[k]$
  4. Combine using the butterfly operation:

$$X[k] = X_e[k] + W_N^{-k} \cdot X_o[k]$$

$$X[k+N/2] = X_e[k] - W_N^{-k} \cdot X_o[k]$$

where  $W_N^{-k} = e^{-j(2\pi/N)k}$  are the twiddle factors

The algorithm recursively applies this decomposition until reaching 2-point DFTs (butterflies). Each stage has  $N/2$  butterflies, and there are  $\log_2 N$  stages.

# FFT BUTTERFLY STRUCTURE

1. The butterfly is the fundamental computational unit of the FFT:
2. Each butterfly performs:
  1. One complex multiplication ( $B \times W_N^k$ )
  2. One complex addition ( $A + WB$ )
  3. One complex subtraction ( $A - WB$ )
3. **Key Property:** The butterfly structure exploits the symmetry of twiddle factors:  $W_N^{k+N/2} = -W_N^k$ , which reduces computations by half.

# DECIMATION-IN-TIME (DIT) FFT

1. In Decimation-in-Time (DIT) FFT, we split the time-domain sequence into even and odd parts:

$$\begin{aligned} X[k] &= \sum_{n \text{ even}} x[n] W_N^{kn} + \sum_{n \text{ odd}} x[n] W_N^{kn} \\ &= \sum_{m=0}^{N/2-1} x[2m] W_N^{2mk} + \sum_{m=0}^{N/2-1} x[2m+1] W_N^{k(2m+1)} \end{aligned}$$

2. Since  $W_N^2 = W_{N/2}$ , we get:

$$\begin{aligned} X[k] &= \sum_{m=0}^{N/2-1} x[2m] W_{N/2}^{mk} + W_N^k \sum_{m=0}^{N/2-1} x[2m+1] W_{N/2}^{mk} \\ &= X_e[k] + W_N^k X_o[k] \end{aligned}$$

# DECIMATION-IN-FREQUENCY (DIF) FFT

- In Decimation-in-Frequency (DIF) FFT, we split the frequency-domain sequence into even and odd parts:
- For k even:  $X[2r] = \sum_{n=0}^{N/2-1} (x[n] + x[n+N/2]) W_{N/2}^{rn}$
- For k odd:  $X[2r+1] = \sum_{n=0}^{N/2-1} (x[n] - x[n+N/2]) W_N^n W_{N/2}^{rn}$

# KEY DIFFERENCES BETWEEN DIT AND DIF

Aspect	DIT FFT	DIF FFT
Decomposition	Time sequence split	Frequency sequence split
Butterfly Input	Natural order	Natural order
Butterfly Output	Bit-reversed order	Natural order
Twiddle Factor	Between stages	Before butterfly
Complexity	Same: $O(N \log N)$	Same: $O(N \log N)$

# FFT COMPLEXITY ANALYSIS

1. The computational savings of FFT compared to DFT are dramatic:
  - a) DFT:  $N^2$  complex multiplications
  - b) FFT:  $(N/2) \log_2 N$  complex multiplications
2. For an  $N$ -point FFT:
  - Number of stages:  $\log_2 N$
  - Butterflies per stage:  $N/2$
  - Complex multiplications per butterfly: 1
  - Total complex multiplications:  $(N/2) \log_2 N$
  - Complex additions:  $N \log_2 N$

# WRITING FFT FUNCTION IN MATLAB - (Radix-2 Cooley-Tukey)

```
function X = myFFT(x)
N = length(x);
if bitand(N, N-1) ~= 0 % Check if N is power of 2
    error('Input length must be a power of 2');
end
if N == 1 % Base case
    X = x;
    return;
end
even = myFFT(x(1:2:end)); % Split into even and odd indices
odd = myFFT(x(2:2:end));
k = 0:N/2-1; % Compute twiddle factors
W = exp(-2*pi*1i*k/N); % Twiddle factors
X = [even + W.*odd, even - W.*odd]; % Combine results
end
```

# APPLICATIONS OF FFT

## 1. Spectral Analysis

- Analyzing frequency content of signals (audio, vibrations, RF)

## 2. Communications

OFDM in 4G/5G, DSL, WiFi, software-defined radio

## 3. Image Processing

JPEG compression, filtering, convolution via multiplication

## 4. Medical

MRI, ECG analysis, ultrasound imaging

## 5. Other applications:

audio compression (MP3), speech recognition, radar, sonar, seismic analysis, and solving partial differential equations.

# EXAMPLE: USING INBUILT FFT FUNCTION IN MATLAB

- % Generate a test signal  
Fs = 1000; % Sampling frequency  
t = 0:1/Fs:1-1/Fs; % Time vector  
f1 = 50; f2 = 120; % Frequencies  
x = 0.7\*sin(2\*pi\*f1\*t) + sin(2\*pi\*f2\*t);  
  
% Compute FFT  
N = length(x);  
X = fft(x);  
X\_mag = abs(X); % Magnitude spectrum  
  
% Frequency vector  
f = (0:N-1)\*(Fs/N);  
plot(f, X\_mag)  
xlabel('Frequency (Hz)')  
ylabel('Magnitude')

# PRACTICAL CONSIDERATIONS

## 1. Windowing

- Apply window functions (Hamming, Hanning) to reduce spectral leakage from finite observation intervals.

## 2. Aliasing

- Ensure sampling rate  $\geq 2 \times$  maximum frequency (Nyquist theorem) to avoid aliasing.

## 3. Zero Padding

- Add zeros to increase FFT size for better frequency resolution (interpolation in frequency domain).

# CONCLUSION

## 1. Key Takeaways:

- FFT reduces DFT complexity from  $O(N^2)$  to  $O(N \log N)$
- Radix-2 FFT uses divide-and-conquer with butterfly operations
- Both DIT and DIF approaches provide the same computational benefits
- FFT enables real-time spectral analysis in countless applications

## 2. Why FFT Matters for EE Students

- Understanding FFT is essential for:
- Digital signal processing (DSP) system design
- Communications engineering (OFDM, software-defined radio)
- Image and audio processing applications
- Embedded systems with real-time signal processing requirements