The Fast Fourier Transform (FFT)

Lecture 20

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Learning Objectives

By the end of this lecture you will be able to:

- Recall the basic mathematical structure of the Discrete Fourier Transform (DFT)
- Understand how the FFT is used to efficiently compute the DFT
- Be able to sketch a block diagram of the basic blocks needed to implement an FFT on an FPGA.

Outline

- Review of the Fourier Transform
 - Continuous Fourier Transform
 - Discrete Fourier Transform (DFT)
- The Fast Fourier Transform (FFT)
- The FFT on an FPGA

The Discrete Fourier Transform (DFT)

The Discrete Fourier Transform (DFT)

Frequency Domain Coefficients

Time Domain Coefficients

$$X[k] = \sum_{n=0}^{N-1} x[n] \cdot W^{k \cdot n} \qquad x[n] = \sum_{n=0}^{N-1} X[k] \cdot W^{-k \cdot n}$$

W = exp[-j2\pi/N] = cos
$$\left(\frac{2\pi}{N}\right) - j sin\left(\frac{2\pi}{N}\right)$$

- x[n]: time domain samples (complex)
- N: number of samples
- X[k]: frequency domain coefficients (complex)
- W = $exp[-j2\pi/N]$: "roots of unity" or twiddle factors (note: sometimes sign is flipped)
- n: time domain index
- k: frequency domain index

Example Signal

 $x[n] = \cos(2\pi f n\Delta t)$

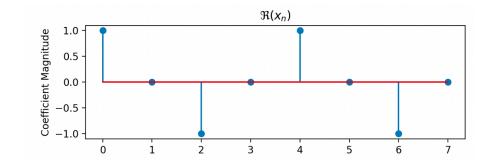
- x[n]: Samples
- f: Signal frequency
- n: Index

Consider a signal with N = 8 samples

- $x[n] = \cos(2\pi f n \Delta t)$
- $n = 0, 1, 2, \dots, N 1$
- $\Delta t = 1/f_s$ where f_s is the sampling frequency
- W = exp $[-j2\pi/N]$ = exp $[-j\pi/4]$

Example Signal

 $x[n] = \cos(2\pi f n \Delta t)$



Set f =2 Hz and f_s =8 Hz. So, this means that we have 4 samples per period of the sinusoid. x[n] = [1,0,-1,0,1,0,-1,0]

Example: DFT computation

- Compute DFT of the signal
- W = exp $[-j2\pi/N]$ = exp $[-j\pi/4]$

$$X[k] = \sum_{n=0}^{N-1} x[n] \cdot W^{k \cdot n}$$

DFT Computational Complexity

- How many multiplications do we need to perform to compute each Fourier coefficient?
 - complex multiplications (real multiplies if signal is real; real multiplications if signal is complex)
 - Assuming a real input signal, each frequency requires _____ multiplies and additions since it costs N – 1 additions to add N numbers and we need to do this for both real and imaginary parts.
 - So, overall computational complexity for a single frequency is and we have N frequencies for a total of

computations.

This means the DFT is

Can we do better?

The Fast Fourier Transform

The Fast Fourier Transform

How can recursion help us to simplify the computation?

$$X[k] = \sum_{n=0}^{N-1} x[n] \cdot W^{k \cdot n}$$

The derivation in the following slides is based on this paper.

A Tutorial-style Single-cycle Fast Fourier Transform Processor

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History: DFT to FFT

- Algorithm invented by Carl Friedrich Gauss around 1805
- Published in 1965 by Cooley (IBM) and Tukey (Princeton)
- Widely used throughout signal processing.

An Algorithm for the Machine Calculation of Complex Fourier Series

By James W. Cooley and John W. Tukey

Symmetries of the roots of unity

- Wⁿ = _____ • Wⁿ =
- Wⁿ =
- $W^{Nk} =$

Decimation in Time (DIT) DFT

• Express the DFT as a sum of two DFTs by splitting the signals into two based on even and odd indices.

$$X[k] = \sum_{n=0}^{N-1} x[n] \cdot W^{k \cdot n}$$

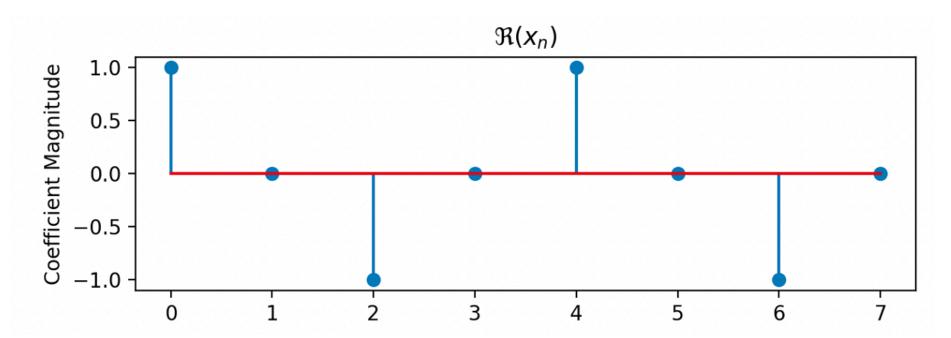
$$X[k] = \sum_{n=0}^{N/2-1} x[2n] \cdot W^{k \cdot 2n} + \sum_{n=0}^{N/2-1} x[2n+1] \cdot W^{k \cdot (2n+1)}$$

$$X[k] = \sum_{n=0}^{N/2-1} x[2n] \cdot W^{k \cdot 2n} + W^k \sum_{n=0}^{N/2-1} x[2n+1] \cdot W^{k \cdot 2n}$$

Decimation in Time (DIT) DFT

$$X[k] = \sum_{n=0}^{N/2-1} x[2n] \cdot W^{k \cdot 2n} + W^k \sum_{n=0}^{N/2-1} x[2n+1] \cdot W^{k \cdot 2n}$$

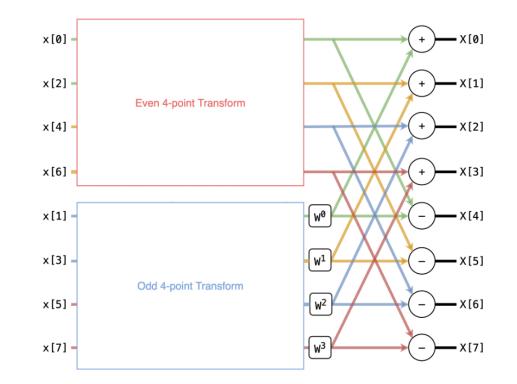
Which samples are in each sum?



DFT Recursion

$$X[k] = \sum_{n=0}^{N-1} x[2n] \cdot W^{k \cdot 2n} + W^k \sum_{n=0}^{N-1} x[2n+1] \cdot W^{k \cdot 2n}$$

This division is equivalent to two separate N/2 DFTs with a final multiplication and addition per output term.



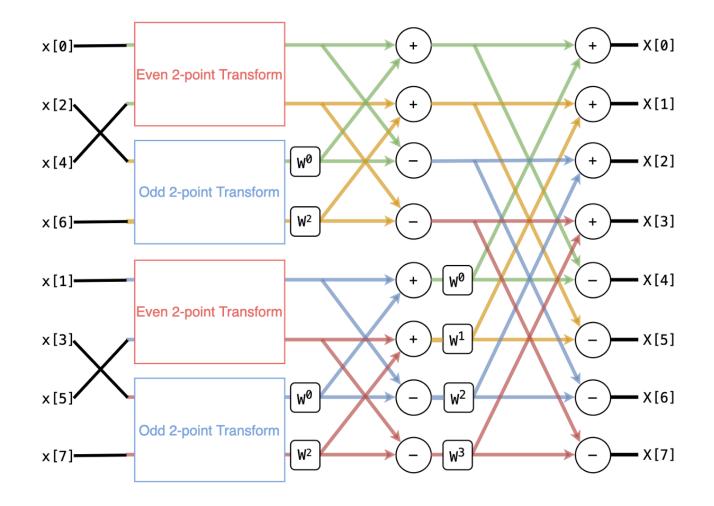
Use recursion to solve 4-point transforms

- Apply the DIT recursion formula to find the 4-point DFT as the sum of two 2-point DFTs.
- Write X[1] (the first output of the 4-point DFT) explicitly using the equation below.

$$X[k] = \sum_{n=0}^{N-1} x[2n] \cdot W^{k \cdot 2n} + W^k \sum_{n=0}^{N-1} x[2n+1] \cdot W^{k \cdot 2n}$$

$$\mathbf{X}[k] = \mathbf{x}[0] \cdot \mathbf{W}^{k \cdot 0} + \mathbf{x}[2] \cdot \mathbf{W}^{k \cdot 2} + \mathbf{W}^{k} \left(\mathbf{x}[1] \cdot \mathbf{W}^{k \cdot 0} + \mathbf{x}[3] \mathbf{W}^{k \cdot 2} \right)$$

Use recursion to solve 4-point transforms



What is the 2-point DFT?

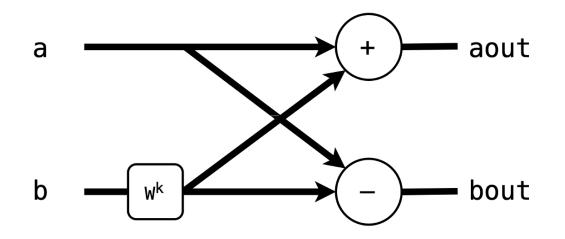
N = 2

$$X[k] = \sum_{n=0}^{N-1} x[2n] \cdot W^{k \cdot 2n} + W^k \sum_{n=0}^{N-1} x[2n+1] \cdot W^{k \cdot 2n}$$



The butterfly unit

- Two complex inputs: A & B
- Multiplies B by the twiddle factor
- Computes sum and difference



Q: Why are the bottom two subtracted instead of added?

A:

FFT Activity

Build your own FFT

Using the template diagrams on the following slides, fill in the missing information.

- The inputs are samples x[n]. You need to determine which samples should be connected to each input.
- The rounded rectangles indicate multiplications by twiddle factors W^k in the butterfly units. You need to determine what the twiddle factors should be.
- The circle indicate either addition or subtraction operations

The most straightforward way to figure this out is to evaluate the time-decimated DFT equation and look for where the nesting occurs. (e.g., the 4-point DFT is the combination of two, 2-point DFTs.)

Step 1: 4-point FFT

Complete the diagram below

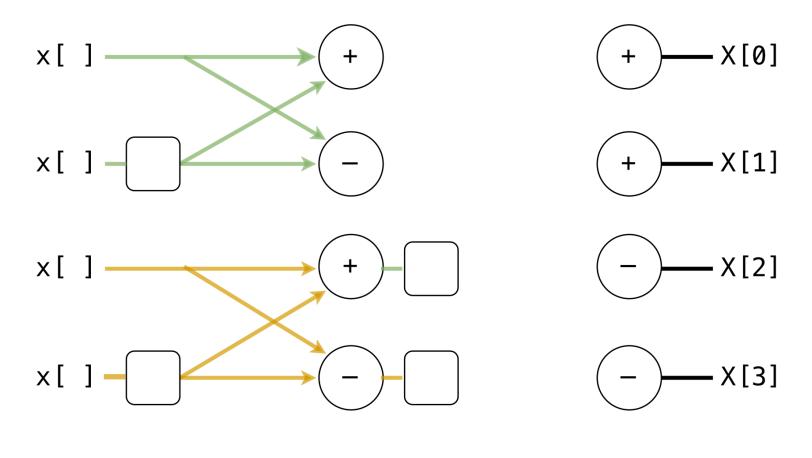
Fill in the rounded rectangles with the appropriate twiddle factors

Draw arrows to route signals properly

Hints:

• Each stage includes $log_2(N)$ butterfly units.

Step 1: 4-point FFT



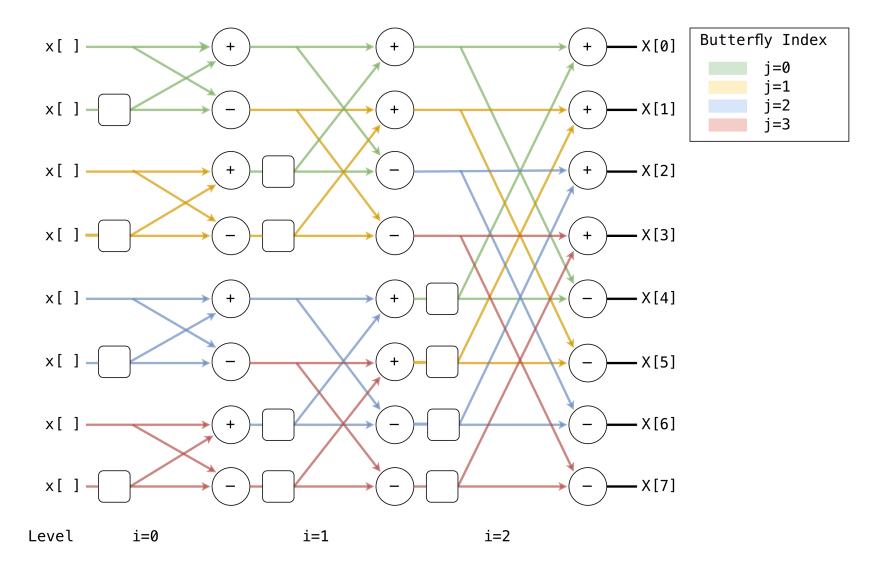
Level i=0 i=1

Step 2: 8-point DFT

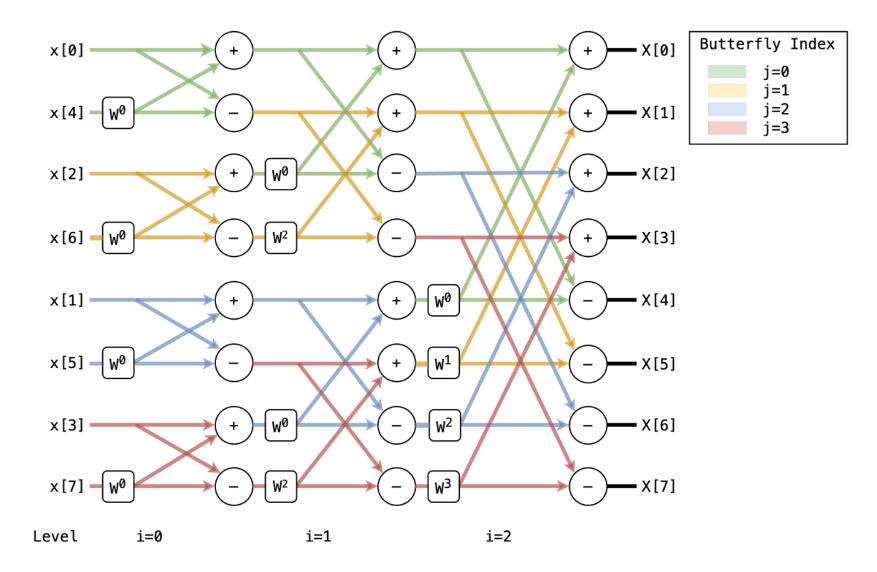
Using the process from the previous slide, repeat this for an 8-point DFT.

Note that the top-left portion of the diagram is identical to the 4-point DFT. You simply add another copy below on the bottom left and then combine them with one more stage of butterflies.

Step 2: 8-point DFT



Solution



Summary

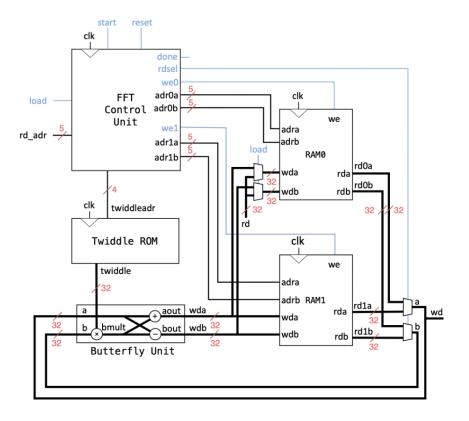
- N-point FFT can be computed with...
 - log₂(N) levels of transforms
 - Each transform has N/2 steps of butterfly operations
 - Each butterfly operation consists of a complex multiplication by a twiddle factor and a complex addition and subtraction.
 - Total of $(N/2) \log_2(N)$ complex multiplications are needed.

FFT in Software

```
1 Bit-reverse-copy (x, X)
                                      // initialize X to bit-reversed order
 2 w=e^(-j*2*pi/N)
                                   // initialize twiddle factor table:
 3 w[0] = 1;
 4 for k = 1 to N/2 - 1
       w[k] = w[k-1] * w
                                      // w[k] holds wk
 5
 6
 7 M = 2
                                  // size of the transform at current level
 8 t = log2(N) - 1
                                  // stride through twiddle factors
9 for s = 0 to log2(N) - 1
                                     // level s
                                  // iterate over M-point transforms
       for k = 0 to N-1 by M
10
11
           for j = 0 to M/2 - 1
                                 // butterflies in M-point transform
               a = X[k+j]
12
13
               b = X[k+j+M/2]
               twiddle = w[j << t]</pre>
14
15
               p = b * twiddle
                                     // multiply by appropriate twiddle factor
               X[k+j] = a + p
16
17
               X[k+j+M/2] = a - p
18
       M = M << 1
                                  // boost size of transform for next level
19
       t = t - 1
                                  // stride of twiddle factors gets smaller
```

FFT in Hardware

- Consider the FFT as a processor like the designs in E85
- Two major components
 - Datapath
 - Controller



A. Vercruysse, M. W. Miller, J. Brake, and D. Harris. 2022. A Tutorial-style Single-cycle Fast Fourier Transform Processor. https://doi.org/10.1145/3526241.3530329

Conclusion

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