

CMPE323 Signals and Systems

Lecture 21: Decimation in Time FFT

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The DFT and the FT



Let x(t) be an energy signal, $\int_{-\infty}^{\infty} |x(t)|^2 dt < \infty$ (basically not periodic)

$$X(f) = \int_{-\infty}^{\infty} x(t)e^{-j2\pi ft}dt \approx \sum_{n=-\infty}^{\infty} x(n\Delta t)e^{-j2\pi fn\Delta t}\Delta t, \ \Delta t = \frac{1}{f_S}$$

$$\approx \frac{1}{f_S} \sum_{n=-\infty}^{\infty} x(n\Delta t) e^{-j2\pi n(f/f_S)} = \frac{1}{f_S} \sum_{n=-\infty}^{\infty} x[n] e^{-j2\pi nF} = \frac{1}{f_S} X_{DTFT}(F)$$

We make the *time* resolution $\Delta t = t_s$ better by increasing f_s .

We make the *frequency* resolution Δf better by increasing T.

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Truncate
$$x(t)$$
 at T , $x_T(t) = \begin{cases} x(t) & 0 \le t < T \\ 0 & \text{otherwise} \end{cases}$

$$X_T(f) \approx \frac{1}{f_S} \sum_{n=0}^{N-1} x[n] e^{-j2\pi nF}, \ N = \frac{T}{t_S} = Tf_S, \ f_S = \frac{N}{T}$$

$$X_T(f) \approx \frac{T}{N} \sum_{n=0}^{N-1} x[n] e^{-j2\pi nF},$$

so
$$X_T(k\Delta f) \approx \frac{T}{N} \sum_{n=0}^{N-1} x[n] e^{-j2\pi nk/N} = \frac{T}{N} X_{DFT}[k]$$

If we make $x_{\tau}(t)$ periodic with period T, the coefficients of the DFS

$$c_{k} = \frac{1}{T} \int_{0}^{T} x_{T}(t) e^{-j2\pi kt/T} dt \approx \frac{\Delta t}{T} \sum_{n=0}^{N-1} x_{T}(n\Delta t) e^{-j2\pi k\Delta tn/T}$$

$$= \frac{1}{f_{S}T} \sum_{n=0}^{N-1} x_{T}(n\Delta t) e^{-j2\pi kn/N} = \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-j2\pi kn/N} = \frac{1}{N} X_{DFT}[k]_{3}$$

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Back to the DFT

We have the DFT equation

$$X_{DFT}[k] = X\left(\frac{k}{T}\right) = \sum_{n=0}^{N-1} x(n\Delta t)e^{-j2\pi kn\Delta t/(N\Delta t)} = \sum_{n=0}^{N-1} x[n]e^{-j2\pi nk/N} = \sum_{n=0}^{N-1} x[n]W_N^{nk}$$

- ...and the DFT is sampled in both time and frequency!
- To compute this requires

(N complex multiplications $(x[n]W_N^{nk}) + N$ complex additions)

 $\times N$ values of $k \approx N^2$ complex "multiply and accumulate" operations

• The FFT is an algorithm to drastically reduce this computation load to $N\log_2 N$

$$N = 1024$$
, $N^2 = 1,048,576$, $N \log_2 N = 1024 \times 10 = 10,240$

$$N^2 / (N \log_2 N) = 102.4$$

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The FFT algorithm

- First published and documented by Cooley & Tukey in 1965 (others had similar ideas)
- The algorithm has multiple forms...
- ...but all the forms make use of the same ideas
- ...including the symmetry and periodicity of W_{N}
- ...and DFTs of simple sequences

$$\begin{split} W_{N/2} &= e^{-j2\pi/(N/2)} = e^{-j2(2\pi/N)} = W_N^2 \\ W_N^{k+(N/2)} &= e^{-j2\pi(k+N/2)/N} = e^{-j2\pi k/N} e^{-j2\pi N/(2N)} = e^{-j2\pi k/N} e^{-j\pi} \\ &= -e^{-j2\pi k/N} = -W_N^k \end{split}$$

If
$$N = 1$$
, $X(0) = x[0]W_1^0 = x[0]$

If
$$N = 2$$
, $X[0] = x[0]W_2^{0 \times 0} + x[1]W_2^{1 \times 0} = x[0] + x[1]$

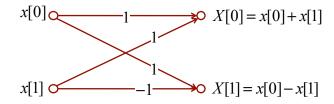
$$X[1] = x[0]W_2^{0 \times 1} + x[1]W_2^{1 \times 1} = x[0] + x[1]W_N^{0 + (N/2)} = x[0] - x[1]$$

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The simple FFT "butterfly"

- The SFG of an FFT operation has a simple shape known as a butterfly...
- ...with the simplest butterfly being an N=2 DFT/FFT



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The Radix-2 Decimation in Time FFT

- "Radix-2" means we group the time-domain points in groups of two...and that we restrict $N = 2^m$
- "Decimation in time" means that we "reorder" the time sequence in preparation for performing the FFT algorithm
- We'll get to that in a few minutes
- First, we start with our defining relationship

$$X_{DFT}[k] = \sum_{n=0}^{N-1} x[n]W_N^{nk}$$

 Second, break the sum into sums on the even and odd indexed terms (not the even and odd parts!)

$$X_{DFT}[k] = \underbrace{\sum_{n=0}^{(N/2)-1} x[2n]W_N^{2nk}}_{\text{even indexed terms}} + \underbrace{\sum_{n=0}^{(N/2)-1} x[2n+1]W_N^{(2n+1)k}}_{\text{odd indexed terms}}$$

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Radix-2 DIT FFT, cont

Third, apply our relationship on the kernel

$$X_{DFT}[k] = \sum_{n=0}^{(N/2)-1} x[2n] \underbrace{W_{N/2}^{nk}}_{W_{N/2}=W_N^2} + W_N^k \sum_{n=0}^{(N/2)-1} x[2n+1] \underbrace{W_{N/2}^{nk}}_{W_{N/2}=W_N^2} W_N^{k}$$

• Let G[k] = DFT(x[2n]), H[k] = DFT(x[2n+1]). These DFTs are N/2 point DFTs, but there are two of them

$$X_{DFT}[k] = G[k] + W_N^k H[k], \ k = 0, 1, \dots \frac{N}{2} - 1$$

• Now, G[k], H[k] are periodic with period N/2

$$X_{DFT} \left[k + \frac{N}{2} \right] = G[k + N/2] + W_N^{(k+N/2)} H[k + N/2]$$

$$= G[k] \underbrace{-W_N^k}_{W_N^{k+N/2} = -W_N^k} H[k]$$
(MRC ENSEAD North Sized Procedure)

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Radix-2 DIT FFT, cont

- So, I can do 2 N/2 point DFTs and combine them to get an N-point DFT
- If I've already done the N/2 point DFT computation (we'll get to that in a minute), then this computation takes

N complex additions (one for each value of k)

 $\frac{N}{2}$ complex multiplications to compute $W_N^k H[k]$, one for each

$$k = 0, 1, 2, \dots, \frac{N}{2} - 1$$

...and this includes special simple cases $(W_N^k = 1, -1, j, -j)$ as full complex multiplications

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Radix-2 DIT FFT, cont

- So now I need two N/2 point DFTs
- ...each of which can be divided into 2 N/4 point DFTs
- ...using the same algorithm requires a total of

$$\underbrace{\frac{2}{\text{one for each}}}_{N/2 \text{ point DFT}} \times \underbrace{\frac{N}{2}}_{\substack{\text{total \# points} \\ \text{in } N/2 \text{ pt. DFT}}} = \underbrace{N}_{\substack{\text{total we} \\ \text{started} \\ \text{with}}} \text{ complex adds}$$

and
$$\underbrace{\frac{2}{2}}_{\text{one for each}} \times \underbrace{\frac{N}{4}}_{\text{one multiply}} = \underbrace{\frac{N}{2}}_{\text{same number as for the } N \text{ pt}} \text{ complex mults.}$$

 ...and then 4 N/4 point DFTs become 8 N/8 point DFTs, all the way down to N/2 2-point DFTs.

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Radix-2 DIT FFT, cont

- It works the same way at each step of the process
- Assume I'm on the r-th step, where $1 \le r \le \log_2 N$
- I decompose the $N/2^{r-1}$ point DFT that I'm trying to compute into two $N/2^r$ point DFTs...
- ...each of which requires

 $N/2^r$ complex adds and $N/2^{r+1}$ complex multiplies

...so there are a total of

$$2 \times \frac{N}{2^r} = \frac{N}{2^{r-1}}$$
 complex adds and $2 \times \frac{N}{2^{r+1}} = \frac{N}{2^r}$ complex mults

• ...per $N/2^{r-1}$ -point DFT, and there are a total of 2^r such DFTs to compute, so the r-th step total is

$$2^{r-1} \times \frac{N}{2^{r-1}} = N$$
 adds and $2^{r-1} \times \frac{N}{2^r} = \frac{N}{2}$ multiplies

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Radix-2 DIT FFT, cont • But $1 \le r \le \log_2 N$, so the total over all of the steps is

 $\log_2 N \times N$ adds and $\log_2 N \times \frac{N}{2}$ multiplies

• ...or approximately $N \log_2 N$ operations!

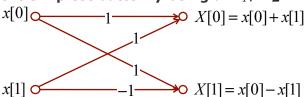
i	FFT/DFT	FFT	Nlog2(N)	DFT	log2(N)	N
	0.75	3	2	4	1	2
C	0.75	12	8	16	2	4
ľ	0.5625	36	24	64	3	8
	0.375	96	64	256	4	16
	0.234375	240	160	1,024	5	32
""	0.140625	576	384	4,096	6	64
	0.08203125	1,344	896	16,384	7	128
	0.046875	3,072	2,048	65,536	8	256
	0.02636719	6,912	4,608	262,144	9	512
	0.01464844	15,360	10,240	1,048,576	10	1024
	0.00439453	73,728	49,152	16,777,216	12	4096
]	0.00036621	1,572,864	1,048,576	4,294,967,296	16	65536
L	2.861E-05	31,457,280	20,971,520	1,099,511,627,776	20	1048576

omplication may not be worth the effort, minimal bang for the buck"

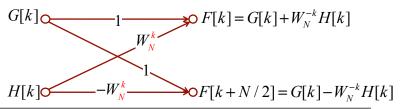
Extra

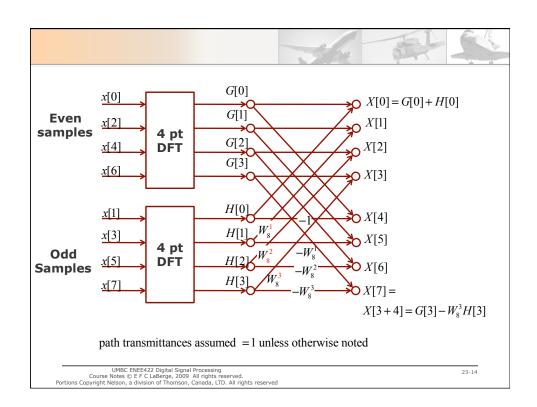
- The simple FFT "butterfly"

 The SFG of an FFT operation has a simple shape known as a butterfly...
- ...with the simplest butterfly being an N=2 DFT/FFT



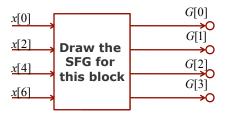
The first stage of our DIT FFT looks like this





You try:

- Draw SFG to implement the "upper" 4-point DFT
- Include any branches necessary for re-ordering...
- ...and the combination branches to generate the 4 point DFT

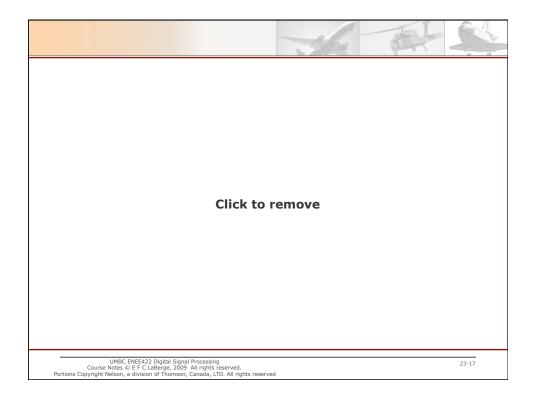


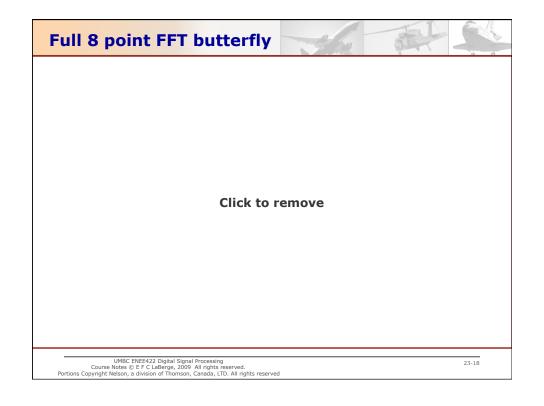
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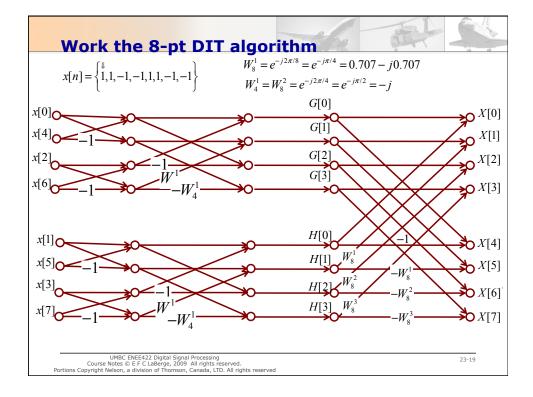
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Bit reversed order

- From this we have that the DIT FFT algorithm must reorder the inputs...
- ...so that the outputs show up in the correct order.
- The decimation in frequency (DIF) FFT algorithm keeps the inputs in their natural (index) order...
- ...and reorders the outputs in exactly the same way!
- We call this reindexing bit reversed order

Binary	LSB on rt	LSB on LF
000	0	0
001	1	4
010	2	2
011	3	6
100	4	1
101	5	5
110	6	3
Die 111	7	7

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The other radix 2 algorithm

The other "famous" radix 2 algorithm is "decimation in frequency"

$$X_{DFT}[k] = \sum_{n=0}^{N/2-1} x[n]W_N^{nk} + \sum_{n=0}^{N/2-1} x[n+N/2]W_N^{(n+N/2)k}$$
first half of input data second half of input data

$$= \sum_{n=-}^{N/2-1} \left(x[n] + W_N^{(N/2)k} x[n+N/2] \right) W_N^{nk}$$

$$= \sum_{n=0}^{N/2-1} \left(x[n] + (-1)^k x[n+N/2] \right) W_N^{nk}$$

 $=\sum_{n=0}^{N/2-1} \left(x[n]+(-1)^k x[n+N/2]\right) W_N^{nk}$ dependence is on the output index being even or odd $k \text{ even: } X[k] = \sum_{n=0}^{N/2-1} \left(x[n]+x[n+N/2]\right) W_{N/2}^{nk}$ and we still have a nice $k \text{ odd: } X[k] = \sum_{n=0}^{N/2-1} \left(x[n]-x[n+N/2]\right) W_N^n W_{N/2}^{nk}$ form

Now the dependence is on

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

$$= DFT(x[n] + x[n+N/2]), k = 0,1,2,...N/2-1$$

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

$$= DFT ((x[n] - x[n+N/2])W_N^k), k = 0,1,2,...N/2-1$$

Let
$$g[n] = x[n] + x[n + N/2]$$

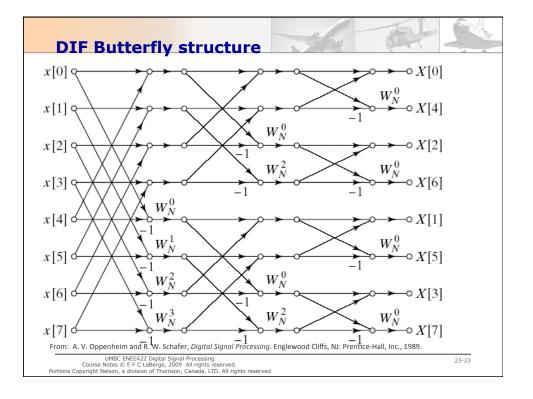
Let
$$h[n] = (x[n] - x[n + N/2])W_N^k$$

$$X[2k] = G[k]$$

$$X[2k+1] = H[k]$$

...and this, too has a butterfly structure

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Wednesday 11/23

- 11 AM Discussion (questions & answers, nothing prepared)
- 2:30 Lecture is your "test" period. We had the 2nd exam as take home, but I continued to lecture, so this is the time when I would have given the exam!
- There is homework, but it is short.

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