

# The Fast Fourier Transform (FFT)

Digital Signal Processing

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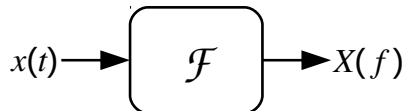
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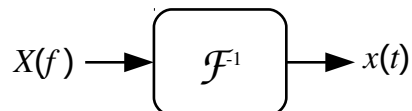
## Fourier Transform: Calculate the frequencies in a signal

- Fourier transform



where  $x(t)$  is the signal,  $t$  is time or sample index,  $X(f)$  is the frequency representation and  $f$  is the frequency.

- Inverse Fourier transform



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# Discrete Fourier Transform (DFT)

- Discrete Fourier transform of a discrete periodic signal  $x(t)$

$$X(f) = \mathcal{F}(X(f)) = \frac{1}{N} \sum_{t=0}^{N-1} x(t) \exp\left(-j \frac{2\pi}{N} ft\right)$$

- Inverse Discrete Fourier transform

$$x(t) = \mathcal{F}^{-1}(x(t)) = \sum_{f=0}^{N-1} X(f) \exp\left(j \frac{2\pi}{N} ft\right)$$

But how to calculate these on a computer?

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## Simple Algorithm to Compute DFT

Simple DFT algorithm can compute the real and imaginary parts independently then combine at the end.

Separate the real and imaginary parts of the complex exponential using Euler's identity:

$$\exp\left(-j \frac{2\pi}{N} ft\right) = \cos\left(\frac{2\pi}{N} ft\right) - j \sin\left(\frac{2\pi}{N} ft\right)$$

so that

$$X(f) = \frac{1}{N} \sum_{t=0}^{N-1} x(t) \left( \cos\left(\frac{2\pi}{N} ft\right) - j \sin\left(\frac{2\pi}{N} ft\right) \right).$$

Then complex multiplication:

$$X(f) = \frac{1}{N} \sum_{t=0}^{N-1} \left( \cos\left(\frac{2\pi}{N} ft\right) \operatorname{Re}(x(t)) + \sin\left(\frac{2\pi}{N} ft\right) \operatorname{Im}(x(t)) - j \left( \sin\left(\frac{2\pi}{N} ft\right) \operatorname{Re}(x(t)) + \cos\left(\frac{2\pi}{N} ft\right) \operatorname{Im}(x(t)) \right) \right).$$

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## Simple Algorithm to Compute DFT:

```
simpleDFT(xRe, xIm, XRe, XIm, N)
a = 2 × π / N
for f = 0 to N - 1
  XRe[f] = 0, XIm[f] = 0, b = a × f
  for t = 0 to N - 1
    arg = b × t, cosarg = cos(arg), sinarg = sin(arg)
    XRe[f] = XRe[f] + cosarg × xRe[t] + sinarg × xIm[t]
    XIm[f] = XIm[f] - sinarg × xRe[t] + cosarg × xIm[t]
  end for
end for
```

×2 for loops, one nested inside other, both length  $N$ .

∴ running time  $\propto N^2$ .

A bit slow, especially for large  $N$ .

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How to Reduce Running Time?

Divide and Conquer

Consider the DFT equation:

$$X(f) = \frac{1}{N} \sum_{t=0}^{N-1} x(t) \exp(-j \frac{2\pi}{N} ft)$$

The sum can be split into even and odd components:

$$\begin{aligned} N \times X(f) &= \sum_{\tau=0}^{N/2-1} x(2\tau) \exp\left(-j \frac{2\pi}{N} f 2\tau\right) + \sum_{\tau=0}^{N/2-1} x(2\tau+1) \exp\left(-j \frac{2\pi}{N} f (2\tau+1)\right) \\ &= \underbrace{\sum_{\tau=0}^{N/2-1} x(2\tau) \exp\left(-j \frac{2\pi}{N/2} f \tau\right)}_{\text{even}} + \underbrace{\exp\left(-j \frac{2\pi}{N} f\right)}_{\text{twiddle}} \underbrace{\sum_{\tau=0}^{N/2-1} x(2\tau+1) \exp\left(-j \frac{2\pi}{N/2} f \tau\right)}_{\text{odd}} \end{aligned}$$

So that

$$N \times X(f) = X_{N/2}^{x_e}(f) + T(N, f) \times X_{N/2}^{x_o}(f)$$

where  $x_e(\tau) = x(2\tau)$  and  $x_o(\tau) = x(2\tau+1)$ .

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DFT Iterative Calculations

Some useful points to remember:

- The DFT is symmetric about the point  $N/2$  which means

$$X(f) = X(f + N/2).$$

- Also noting

$$\begin{aligned} \exp(-j2\pi(f + N/2)/N) &= \exp(-j\pi) \exp(-j2\pi f/N) \\ &= -\exp(-j2\pi f/N). \end{aligned}$$

Therefore we now have

$$N \times X(f) = \begin{cases} X_{N/2}^{x_e}(f) + T(N, f) \times X_{N/2}^{x_o}(f) & \text{for } 0 \leq f < N/2, \\ X_{N/2}^{x_o}(f - \frac{N}{2}) - T(N, f - \frac{N}{2}) \times X_{N/2}^{x_e}(f - \frac{N}{2}) & \text{for } N/2 \leq f < N. \end{cases}$$

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How to Reduce Running Time?

Splitting into even and odd components can be performed repeatedly (recursively) until  $N/2 - 1 = 1$ .

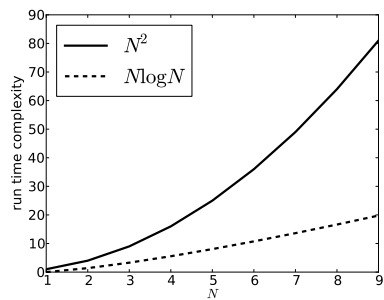
```
recursiveFFT(x, X, N, s)
  if N = 1
    XRe[0] = xRe[0], XIm[0] = xIm[0]
  else
    N2 = N/2
    recursiveFFT(x, X, N2, 2s) //even
    recursiveFFT(x + s, X + s, N2, 2s) //odd
    a = 2 * pi / N
    for f = 0 to N2 - 1
      T = exp(-j * f * a) //twiddle
      even = X[f], odd = X[f + s]
      X[f] = even + T * odd
      X[f + N2] = even - T * odd
    end for
  end if
```

- Decimation in Time (DIT) principle:
  - Proposed by Cooley and Tukey in 1965.
- Single for loop.
- However this is a recursive function...

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So what is the running time?

- Total running time is approximately  $T(N) \propto N \log N$
- This is much faster than simpleDFT algorithm with  $N^2$ .



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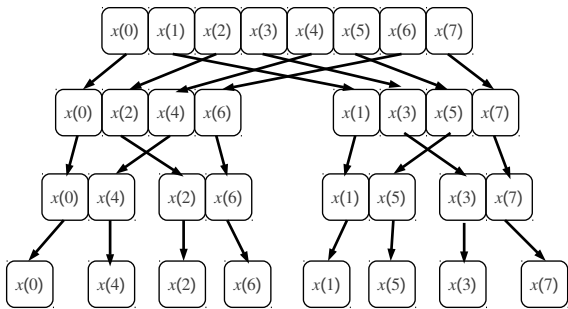
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Any Further Improvements?

Remove Recursion

Recursion tree for input sequence  $x(t)$  with 8 samples:



- Recursion not good for computer systems with small stacks.
- Stack overflow may occur on many levels (4 levels here).

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Removing the Recursion

Observe the sequence in the bottom level of the recursion tree:

$t$ (base 10)	0	4	2	6	1	5	3	7
$t$ (base 2)	000	100	010	110	001	101	011	111

Reversing the bits produces numbers in a natural order:

$t$ (base 2, bit reversed)	000	001	010	011	100	101	110	111
$t$ (base 10, bit reversed)	0	1	2	3	4	5	6	7

This information can be used to *simulate* the recursion.

- The data can be sorted using indices arranged in bit reverse order.
- An algorithm can then:
  - Start from the bottom of the tree and iterate up
  - But without having to perform recursion.

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# Recursionless FFT

```
fft(x,X,N)
  Copy data from x to X
  SwapDataUsingBitReversedIndices(X)
  mmax = 2, n = N << 1
  //calculate transforms length 2,4,8,...,N
  while (n > mmax) //loops log2N times
    istep = mmax << 1
    initialise T //(twiddle factor)
    for m = 1 to mmax in steps 2
      for i = m to n in steps istep
        j = i + mmax
        X[i] = X[i] + T * X[j]
        X[j] = X[i] - T * X[j]
      end for
      update T //using trigonometric recurrences
    end for
    mmax = istep
  end while
```

This algorithm has the same run time complexity of  $N \log N$ .

Notes

# Summary

- Fourier Analysis is made practical using computer algorithms, particularly efficient computer algorithms such as the Fast Fourier Transform (FFT).
- Better algorithms can be designed with insight of the underlying mathematics.

More information can be found in the following books:

- “Numerical Recipes”, by Press, Teukolsky, Vetterling and Flannery, Cambridge University Press, 2007.
- “Introductory Digital Signal Processing with Computer Applications”, 2nd Edition, by Lynn and Fuerst, Wiley 1998.

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