Fast Fourier Transform (FFT) and Short-Time Fourier Transform (STFT)

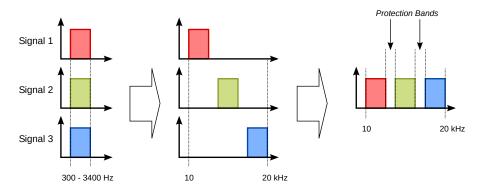
Digital Signal Processing

September 23, 2025



Application of FFT: Communications

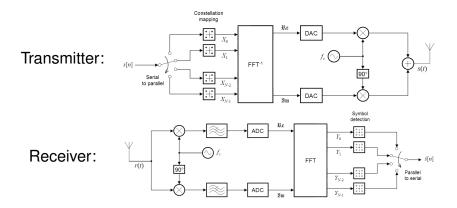
Frequency Division Multiplexing (FDM)



https://commons.wikimedia.org/w/index.php?curid=18327256

Application of FFT: Communications

Orthogonal Frequency Division Multiplexing (OFDM)



Technology behind most digital wireless communication! (WiFi, 4G, 5G, HD Radio, HDTV)

The Fast Fourier Transform (FFT)

Naïve DFT Algorithm

Recall the DFT equation:

$$X[k] = \frac{1}{\sqrt{L}} \sum_{n=0}^{L-1} e^{-i\omega_0 nk} x[n]$$

Naïve DFT Algorithm

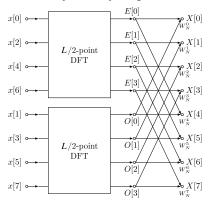
Loop over $k = 0, \dots, L-1$

Compute X[k] by sum (loop) over $n = 0, \dots, L-1$

Complexity is $O(L^2)$

Fast Fourier Transform

Cooley-Tukey Algorithm



Divide-and-Conquer

Radix-2 version:

- 1 compute "even" DFT
- 2 compute "odd" DFT
- ${f 3}$ combine and reuse results Recursively apply to each L/2 block

Complexity is $O(L \log L)$

Fast Fourier Transform

- More general radix-p FFT breaks the DFT into p blocks, where p is a prime factor of the signal length L
- Recursively applied to each L/p block
- Recursion stops when the remaining block lengths are prime numbers (can't be factored any further)
- Bottom line: The FFT is most efficient when the input signal length has small prime factors, preferrably L is a power of 2.
- Sometimes it is more efficient to pad a signal with zeros to get a good prime factorization.

First Half of FFT

Compute X[k] for $k = 0, 1, \dots, \frac{L}{2} - 1$,

$$\begin{split} X[k] &= \frac{1}{\sqrt{L}} \sum_{n=0}^{L-1} e^{-i\omega_0 n k} x[n] \\ &= \frac{1}{\sqrt{L}} \sum_{m=0}^{L/2-1} e^{-i\omega_0 2m k} x[2m] + \frac{1}{\sqrt{L}} \sum_{m=0}^{L/2-1} e^{-i\omega_0 (2m+1)k} x[2m+1] \\ &= \underbrace{\frac{1}{\sqrt{L}} \sum_{m=0}^{L/2-1} e^{-i\omega_0 2m k} x[2m]}_{E[k] = \text{sum of even terms}} + e^{-i\omega_0 k} \underbrace{\frac{1}{\sqrt{L}} \sum_{m=0}^{L/2-1} e^{-i\omega_0 2m k} x[2m+1]}_{O[k] = \text{sum of odd terms}} \end{split}$$

 $= E[k] + e^{-i\omega_0 k} O[k]$

Second Half of FFT

Compute $X[k+\frac{L}{2}]$ for $k=0,1,\ldots,\frac{L}{2}-1$,

$$X\left[k + \frac{L}{2}\right] = \frac{1}{\sqrt{L}} \sum_{m=0}^{L/2-1} e^{-i\omega_0 2m(k + \frac{L}{2})} x[2m] + e^{-i\omega_0(k + \frac{L}{2})} \frac{1}{\sqrt{L}} \sum_{m=0}^{L/2-1} e^{-i\omega_0 2m(k + \frac{L}{2})} x[2m + 1]$$
$$= E[k] - e^{-i\omega_0 k} O[k]$$

Using $e^{-i\omega_0 2m(k+\frac{L}{2})}=e^{-i\omega_0 2mk}$ and $e^{-i\omega_0(k+\frac{L}{2})}=-e^{-i\omega_0 k}$

Second Half of FFT

Compute $X[k+\frac{L}{2}]$ for $k=0,1,\ldots,\frac{L}{2}-1$,

$$\begin{split} X\left[k+\frac{L}{2}\right] &= \frac{1}{\sqrt{L}} \sum_{m=0}^{L/2-1} e^{-i\omega_0 2m(k+\frac{L}{2})} x[2m] + \\ &\quad + e^{-i\omega_0(k+\frac{L}{2})} \frac{1}{\sqrt{L}} \sum_{m=0}^{L/2-1} e^{-i\omega_0 2m(k+\frac{L}{2})} x[2m+1] \\ &= E[k] - e^{-i\omega_0 k} O[k] \quad \text{Reused!} \end{split}$$

Using $e^{-i\omega_0 2m(k+\frac{L}{2})}=e^{-i\omega_0 2mk}$ and $e^{-i\omega_0(k+\frac{L}{2})}=-e^{-i\omega_0 k}$

FFT Historical Trivia

- FFT actually invented by Gauss in 1805! (but lost)
- Re-invented by Cooley and Tukey in 1965
- Tukey coined the term "bit" (for "binary digit") and was first to use the term "software" in writing



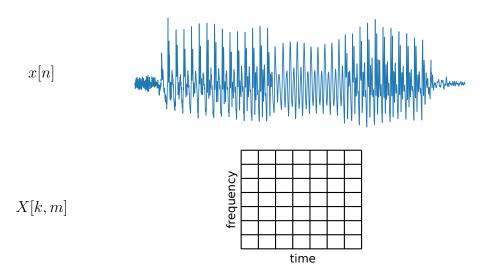
Carl Friedrich Gauss



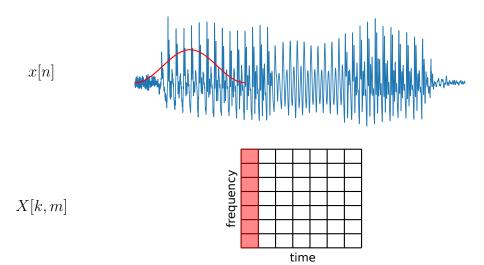
John Tukey

The Short-Time Fourier Transform (STFT)

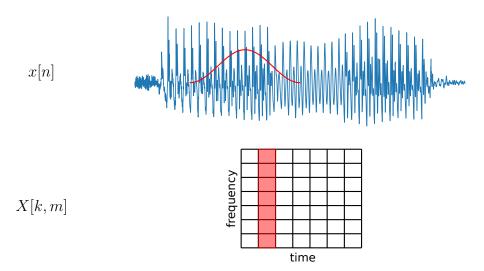
Multiply signal $\boldsymbol{x}[n]$ by a sliding window $\boldsymbol{w}[n]$ and take FFT.



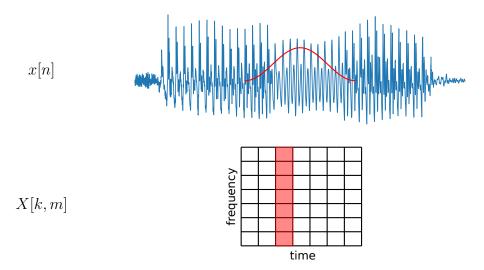
Multiply signal $\boldsymbol{x}[n]$ by a sliding window $\boldsymbol{w}[n]$ and take FFT.



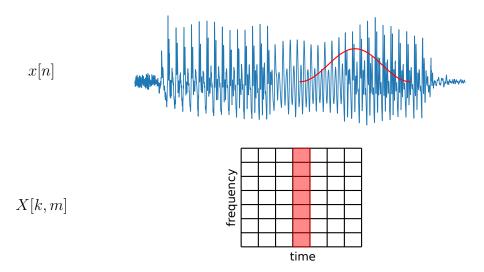
Multiply signal x[n] by a sliding window w[n] and take FFT.



Multiply signal $\boldsymbol{x}[n]$ by a sliding window $\boldsymbol{w}[n]$ and take FFT.



Multiply signal $\boldsymbol{x}[n]$ by a sliding window $\boldsymbol{w}[n]$ and take FFT.



STFT Definition

The STFT of a signal, x[n], is a function of frequency, k, and time, m, given by:

$$X[k,m] = \frac{1}{\sqrt{W}} \sum_{n=0}^{W-1} x[n+mh]w[n]e^{-\frac{i2\pi kn}{W}},$$

where W is the window length, h is the hop length.

STFT Pseudocode

```
x = input signal
w = window function
h = hop length
H = number of hops
X = output STFT

for m = 0 .. H
    x_clip = x[m*h : m*h + len(w)]
    X[:, m] = FFT(x_clip * w)
```

Overlap-Add (OLA)

We can reconstruct a signal x[n] from its STFT, X[k, m], using a method called **Overlap-Add (OLA)**:

① Compute the inverse fast Fourier transform on each column of X[k,m] to get

$$s[n,m] = \mathcal{DFT}^{-1}(X[k,m])$$

2 Scale s by a window, w[n], and sum over m:

$$\tilde{x}[n] = \sum_{m} s[n - mh, m]w[n - mh],$$

where h is the hop length used to compute X[k,m], and the range of the summation is $\left\lceil \frac{1-W+n}{h} \right\rceil \leq m \leq \left\lfloor \frac{n}{h} \right\rfloor$.

Pseudocode for OLA

```
X = STFT
w = window function
x = output signal
initialize x[n] = 0 for all n
for m = 0 .. H
    s = IFFT(X[:, m])
    x[m*h : m*h + len(w)] += s * w
```

Perfect Reconstruction Conditions

The $\tilde{x}[n]$ resulting from OLA is a reconstruction of x[n], but are they equal?

Yes, if window satisfies the constant overlap-add (COLA) condition:

$$\sum_{m} w[n - mh]^2 = 1$$

Note: We can also apply two different windows $w_f[n]$ and $w_b[n]$ during forward STFT and backward OLA, respectively. Then the condition is that $\sum_m w_f[n-mh]w_b[n-mh]=1$.

Perfect Reconstruction Conditions

First, we have

$$s[n,m] = \mathcal{DFT}^{-1}\{X[k,m]\}$$

= $\mathcal{DFT}^{-1}\{\mathcal{DFT}\{x[n+mh]w[n]\}\},$
= $x[n+mh]w[n]$

Perfect Reconstruction Conditions

So, assuming w[n] is COLA:

$$\begin{split} \tilde{x}[n] &= \sum_{m} s[n-mh,m]w[n-mh] \\ &= \sum_{m} (x[n]w[n-mh])w[n-mh] \\ &= x[n] \underbrace{\sum_{m} w[n-mh]^2}_{=1 \text{ if COLA}} \\ &= x[n] \end{split}$$