

Lecture 22: Overlap-Add

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ECE 401: Signal and Image Analysis

- 1 Review: Circular Convolution
- 2 Fast Fourier Transform
- 3 Overlap-Add
- 4 Conclusion

Outline

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Review: Circular convolution

Multiplying the DFT means **circular convolution** of the time-domain signals:

$$y[n] = h[n] \circledast x[n] \leftrightarrow Y[k] = H[k]X[k],$$

Circular convolution ($h[n] \circledast x[n]$) is defined like this:

$$h[n] \circledast x[n] = \sum_{m=0}^{N-1} x[m] h[((n - m))_N] = \sum_{m=0}^{N-1} h[m] x[((n - m))_N]$$

Circular convolution is the same as linear convolution if and only if $N \geq L + M - 1$.

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Computational Complexity: Convolution and DFT

Convolution is an $O\{N^2\}$ operation: each of the N samples of $y[n]$ is created by adding up N samples of $x[m]h[n-m]$:

$$y[n] = \sum_m x[m]h[n-m]$$

The way we've learned it so far, the DFT is **also** an $O\{N^2\}$ operation: each of the N samples of $X[k]$ is created by adding up N samples of $x[n]e^{j\omega_k n}$:

$$X[k] = \sum_n x[n]e^{-j\frac{2\pi kn}{N}}$$

However...

The Fast Fourier Transform

- The **fast Fourier transform** (FFT) is a clever divide-and-conquer algorithm that computes all of the N samples of $X[k]$, from $x[n]$, in only $N \log_2 N$ multiplications.
- It does this by computing all N of the $X[k]$, all at once.
- Multiplications ($x[n] \times w_{k,n}$, for some coefficient $w_{k,n}$) are grouped together, across different groups of k and n .
- On average, each of the N samples of $X[k]$ can be computed using only $\log_2 N$ multiplications, for a total complexity of $N \log_2 N$.

What's the difference between N^2 and $N \log_2 N$?

Consider filtering $N = 1024$ samples of audio (about 1/40 second) with a filter, $h[n]$, that is 1024 samples long.

- Time-domain convolution requires $1024 \times 1024 \approx 1,000,000$ multiplications. If a GPU does 40 billion multiplications/second, then it will take an hour of GPU time to apply this operation to a 1000-hour audio database.
- FFT requires $1024 \times \log_2 1024 \approx 10,000$ multiplications. If a GPU does 40 billion multiplications/second, then it will take 36 seconds of GPU time to apply this operation to a 1000-hour audio database.

How is it used?

Suppose we have a 1025-sample $h[n]$, and we want to filter a one-hour audio (144,000,000 samples). Divide the audio into frames, $x[n]$, of length $M = 1024$, zero-pad to $N = L + M - 1 = 2048$, and take their FFTs.

- $H[k] = \text{FFT}\{h[n]\}$: total cost is trivial, because we only need to do this once.
- $X[k] = \text{FFT}\{x[n]\}$: total cost is $N \log N$ per M samples.
- $Y[k] = X[k]H[k]$: total cost is N multiplications per M samples.
- $y[n] = \text{FFT}^{-1}\{Y[k]\}$: total cost is $N \log N$ per M samples.

Grand total: $N \times (2 \log_2 N + 1) = 2048 \times 23 = 47104$ multiplications per 1024 audio samples, or 46 multiplications per sample.

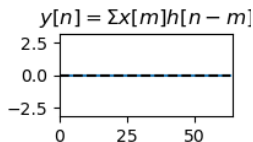
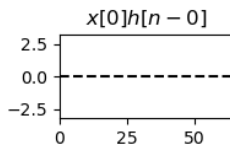
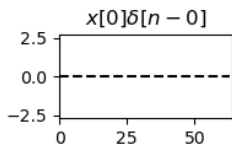
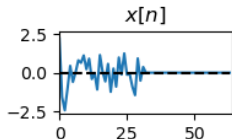
How do we recombine the $y[n]$?

- The main topic of today's lecture: how do we recombine the $y[n]$?
- Remember: each frame of $x[n]$ was 1024 samples, but after zero-padding and convolution, each frame of $y[n]$ is 2048 samples.
- How do we recombine them?

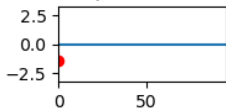
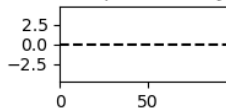
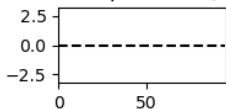
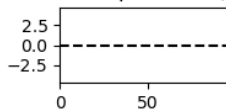
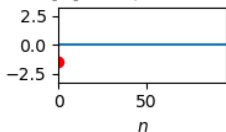
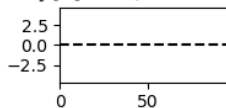
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Let's look more closely at what convolution is. Each sample of $x[n]$ generates an impulse response. Those impulse responses are added together to make the output.



First two lines show the first two frames (input on left, output on right). Last line shows the total input (left) and output (right).

First input frame, $x_1[n]$ First output frame, $y_1[n]$ Second input frame, $x_2[n]$ Second output frame, $y_2[n]$  $x[n]$ total, all frames $y[n]$ total, all frames

The Overlap-Add Algorithm

- 1 Divide $x[n]$ into frames
- 2 Generate the output from each frame
- 3 Overlap the outputs, and add them together

The Overlap-Add Algorithm

- 1 Divide $x[n]$ into frames ($w[n]$ is a length- M rectangle).

$$x_t[n] = x[n + tM]w[n]$$

$$X_t[k] = \text{FFT}\{x_t[n]\}$$

- 2 Generate the output from each frame

$$Y_t[k] = X_t[k]H[k]$$

$$y_t[n] = \text{FFT}^{-1}\{Y_t[k]\}$$

- 3 Overlap the outputs, and add them together

$$y[n] = \sum_t y_t[n - tM]$$

Quiz

Try the quiz!

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