Lecture 2

Digital Signal Processing

Chapter 2

Convolution Impulse response Difference equations Correlation functions

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Convolution (page 71-80)

The most important connection between input signal and output signal is called *convolution*. If we know the impulse response h(n) of a system, we can calculate the output signal for any input signal. We are only assuming the properties of linearity and time invariance (LTI).

| Input signal | \rightarrow | Output signal |
|----------------------------|---------------|-----------------------|
| x(n) | \rightarrow | y(n) |
| $\delta(n)$ | \rightarrow | h(n) |
| $\delta(n-k)$ | \rightarrow | h(n-k) |
| $x(k)\delta(n-k)$ | \rightarrow | x(k)h(n-k) |
| $\sum_{k} x(k)\delta(n-k)$ | \rightarrow | $\sum_{k} x(k)h(n-k)$ |

$$y(n) = \sum_{k} x(k)h(n-k) = \sum_{k} h(k)x(n-k) = h(n) * x(n)$$
(1)

This dependence is called is called convolution and is the most common and diverse formula in the course.

Example of convolution

Given: Input signal x(n) and impulse response h(n).

$$x(n) = \left\{ \begin{array}{ccc} \underline{2} & 4 & 6 & 4 & 2 \end{array} \right\}$$
(2)

$$h(n) = \left\{ \begin{array}{ccc} \underline{3} & 2 & 1 \end{array} \right\} \tag{3}$$

Find: Output signal y(n).

$$y(n) = \sum_{k} h(n-k)x(k) = \sum_{k} h(k)x(n-k)$$
 (4)

$$= h(0)x(n) + h(1)x(n-1) + h(2)x(n-2)$$
(5)

$$= 3x(n) + 2x(n-1) + x(n-2)$$
(6)

Solution: We solve the convolution graphically with the following visual procedure For n = 0:

| h(0-k) | 1 | 2 | <u>3</u> | | | | | |
|------------|---|---|----------|---|---|---|---|-------------------------------|
| x(k) | | | <u>2</u> | 4 | 6 | 4 | 2 | |
| h(0-k)x(k) | | | <u>6</u> | | | | | $\sum = \underline{6} = y(0)$ |

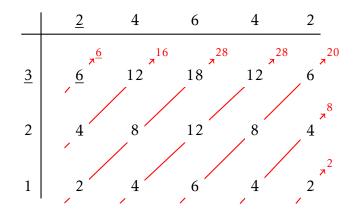
For *n* = 1:

For *n* = 2:

| h(1-k) | 1 | 2 | <u>3</u> | | | | |
|---------------|---|----------|----------|---------------|---|---|--------------------|
| x(k) | | <u>2</u> | 4 | 6 | 4 | 2 | |
| h(1-k)x(k) | | 4 | 12 | | | | $\sum = 16 = y(1)$ |
| | | | | | | | |
| | | | | | | | |
| h(2-k) | | 1 | 2 | <u>3</u> | | | |
| h(2-k) $x(k)$ | | | | <u>3</u> 6 | 4 | 2 | |

Multiply the components of each rows and add the results. Shift the impulse response one step to the right and repeat. Repeat as long as h(n - k) covers the signal x(k). The output is

Equivalent solution with a table.



Multiply rows and columns in the matrix. Sum along the anti-diagonals and read the result in the direction of the diagonal.

Properties of convolution (page 81)

The usual properties apply.

Commutativity

$$x_1(n) * x_2(n) = x_2(n) * x_1(n)$$
(8)

Associativity

$$x_1(n) * [x_2(n) * x_3(n)] = [x_1(n) * x_2(n)] * x_3(n)$$
(9)

Distributivity

$$x_1(n) * [x_2(n) + x_3(n)] = x_1(n) * x_2(n) + x_1(n) * x_3(n)$$
(10)

Input-output

$$y(n) = x(n) * h(n) \tag{11}$$

$$x(n) \longrightarrow h(n) \longrightarrow y(n)$$

Cascade or Serial coupling

$$y(n) = x(n) * h_1(n) * h_2(n)$$
(12)

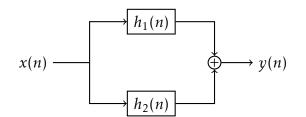
$$h(n) = h_1(n) * h_2(n)$$
(13)

$$x(n) \longrightarrow h_1(n) \longrightarrow h_2(n) \longrightarrow y(n)$$

Parallel coupling

$$y(n) = [x(n) * h_1(n)] + [x(n) * h_2(n)] = x(n) * [h_1(n) + h_2(n)]$$
(14)

$$h(n) = h_1(n) + h_2(n) \tag{15}$$



Stability (sid 85)

A system is BIBO-stable (bounded input-bounded output) if

$$|x(n)| \le M_x \quad \Rightarrow \quad |y(n)| \le M_y \tag{16}$$

or equivalently

$$\left| y(n) \right| = \left| \sum_{k=-\infty}^{\infty} h(k) x(n-k) \right|$$
(17)

$$\leq \sum_{k=-\infty}^{\infty} |h(k)| |x(n-k)| \tag{18}$$

$$\leq M_x \cdot \sum_{k=-\infty}^{\infty} |h(k)| \tag{19}$$

The system is therefore stable if

$$\sum_{k=-\infty}^{\infty} |h(k)| < \infty$$
(20)

Difference equations (page 93–95)

General:

$$y(n) + \sum_{k=1}^{N} a_k y(n-k) = \sum_{k=0}^{N} b_k x(n-k)$$
(21)

Example

The FIR-filter

$$y(n) = 0.5x(n) + 0.25x(n-1) + 0.15x(n-2)$$
⁽²²⁾

immediately gives us the impulse response

 $h(n) = \left\{ \begin{array}{ccc} 0.5 & 0.25 & 0.15 \end{array} \right\}$ (23)

A first order IIR-filter:

$$y(n) = 0.5y(n-1) + 2x(n)$$
(24)

A second order IIR-filter:

$$y(n) = 0.5y(n-1) + 0.5y(n-2) + x(n)$$
⁽²⁵⁾

For IIR-filters we have to solve the difference equation in order to determine the impulse response h(n). We will solve a first order difference equation (page 94).

$$y(n) = -a_1 y(n-1) + b_0 x(n)$$
(26)

Solve iteratively for $n \ge 0$.

$$y(0) = -a_1 y(-1) + b_0 x(0) \tag{27}$$

$$y(1) = -a_1 y(0) + b_0 x(1) = (-a_1)^2 y(-1) + b_0 x(1) + (-a_1) b_0 x(0)$$
(28)

$$y(2) = -a_1 y(1) + b_0 x(2) = (-a_1)^3 y(-1) + b_0 x(2) + (-a_1) b_0 x(1) + (-a_1)^2 b_0 x(0)$$
(29)

$$y(n) = \sum_{k=0}^{n} (-a_1)^k \cdot b_0 x(n-k) + \underbrace{(-a_1)^{n+1} \cdot y(-1)}_{\text{often 0}}$$
(30)

We will wait until chapter 3 and the *z*-transform to solve higher order difference equations.

Example

Given:

$$h(n) = \left(\frac{1}{2}\right)^n \cdot u(n) \tag{31}$$

$$x(n) = u(n) \tag{32}$$

Find:

$$y(n) = h(n) * x(n) \tag{33}$$

Solution: Convolution gives

$$y(n) = \sum_{k=-\infty}^{\infty} h(k)x(n-k) \qquad [h(k) = 0 \text{ if } k < 0 \text{ and } x(n-k) = 0 \text{ if } k > n]$$
(34)

$$=\sum_{k=-\infty}^{\infty} \left(\frac{1}{2}\right)^{k} \cdot u(k) \cdot u(n-k) = \sum_{k=0}^{n} \left(\frac{1}{2}\right)^{k}$$
(35)

$$=\frac{1-\left(\frac{1}{2}\right)^{n+1}}{1-\frac{1}{2}}=2-\left(\frac{1}{2}\right)^{n}\qquad n\geq0$$
(36)

The solution is therefore

$$y(n) = \left[2 - \left(\frac{1}{2}\right)^n\right] \cdot u(n) \tag{37}$$

Correlation functions (sid 118)

How similar are two signals?

Auto correlation function

$$r_{xx}(k) = \sum_{n = -\infty}^{\infty} x(n)x(n-k) = x(k) * x(-k)$$
(38)

Cross correlation function

$$r_{yx}(k) = \sum_{n = -\infty}^{\infty} y(n)x(n-k) = y(k) * x(-k)$$
(39)

Cross correlation for input and output signals

$$x(n) \longrightarrow h(n) \longrightarrow y(n)$$

The auto correlation for the input signal:

$$r_{xx}(k) = x(k) * x(-k)$$
(40)

The cross correlation between the input signal and the output signal:

$$r_{yx}(k) = y(k) * x(-k)$$
 (41)

$$= h(k) * x(k) * x(-k)$$
(42)

$$=h(k)*r_{xx}(k) \tag{43}$$

The auto correlation for the output signal:

$$r_{yy}(k) = y(k) * y(-k)$$
 (44)

$$= h(k) * x(k) * h(-k) * x(-k)$$
(45)

$$= r_{hh}(k) * r_{xx}(k) \tag{46}$$

We can determine an unknown system h(n) by using an input signal x(n). For example, if x(n) is white noise, then

$$r_{xx}(k) = \delta(k) \tag{47}$$

and therefore the impulse response becomes

$$h(k) = r_{yx}(k) \tag{48}$$

Example of IIR-filter

Determine the balance of a bank account with interest.

Given: Deposit is 100 every year with 5% interest.

$$x(n) = 100 \cdot u(n) \tag{49}$$

 $y(n) = \text{balance at year } n \tag{50}$

Find: Balance after 1, 2, 5 and 20 years.

Solution: The current balance is the balance from last year plus 5% interest and the deposit for the current year.

$$y(n) = 1.05y(n-1) + x(n)$$
(51)

We have a recursive system where the new balance depends on both the previous balance (old output signal) and the deposit (input signal). This is an IIR-filter.

Iterative solution gives:

$$y(0) = 1.05y(-1) + x(0) = 100$$
(52)

y(n) = 0 for n < 0 before the saving started.

$$y(1) = 1.05y(0) + x(1) = 1.05 \cdot 100 + 100$$
(53)

$$y(2) = 1.05y(1) + x(2) = 1.05 \cdot (1.05 \cdot 100 + 100) + 100$$
(54)

$$y(3) = \dots \tag{55}$$

Using the *z*-transform we can determine a formula for y(n) (more on that later).

$$Y(z) \cdot (1 - 1.05z^{-1}) = X(z)$$
(56)

$$X(z) = \frac{100}{1 - z^{-1}} \tag{57}$$

$$Y(z) = \frac{100}{(1 - 1.05z^{-1})(1 - z^{-1})}$$
(58)