

Module FIR

Finite Impulse Response filter

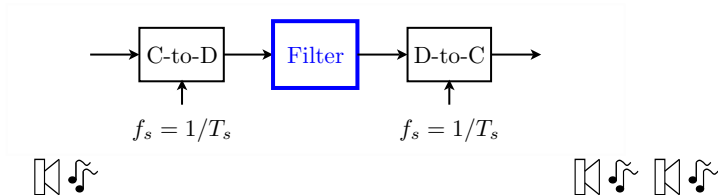
Piet Sommen

TU / **e**

Technische Universiteit
Eindhoven
University of Technology

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Where innovation starts



Example: The running average filter

FIR-2

Compute a **moving (running) average of two or more consecutive samples**, forming a new sequence of the average values

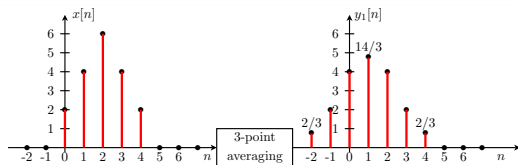
Note: FIR is generalization of running average

Averaging used when data fluctuates, thus smoothing prior interpretation (e.g. view trends). E.g. Stock-market prices, credit-card balances, etc.

Example:

Input-output relation (=difference equation) 3-point averaging

$$\rightarrow y_1[n] = \frac{1}{3} (x[n] + x[n+1] + x[n+2])$$



The running average filter

Observations output signal samples:

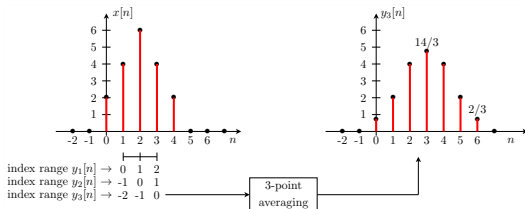
Smooths input , longer than input , finite sequence , starts (becomes nonzero) **before** input starts (=noncausal filter)

Last issue undesirable if input comes directly from A-to-D (e.g. audio).

Causal filter uses only present and past values

$$\text{Noncausal} \quad y_2[n] = \frac{1}{3} (x[n+1] + x[n] + x[n-1])$$

$$\text{Causal} \quad y_3[n] = \frac{1}{3} (x[n] + x[n-1] + x[n-2])$$

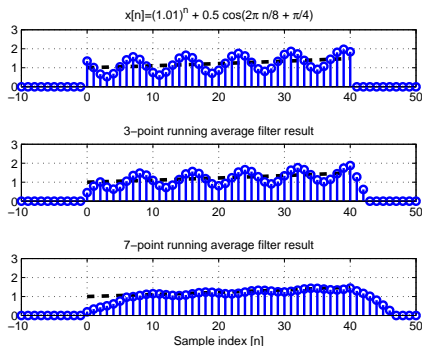


Find trend in the following signal:

$$x[n] = \begin{cases} (1.01)^n + \frac{1}{2} \cos(2\pi n/8 + \pi/4) & 0 \leq n \leq 40 \\ 0 & \text{otherwise} \end{cases}$$

Use M -point averaging filter \rightarrow

$$y[n] = \frac{1}{M} \sum_{k=0}^{M-1} x[n-k]$$



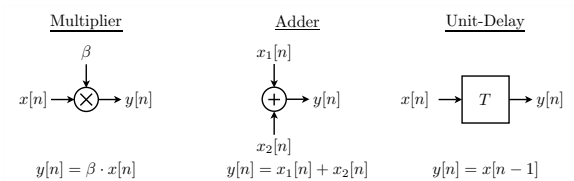
Observations:

- Run-in and -out area $y[n]$
- Length $y[n] \uparrow$
- For $M \uparrow \Rightarrow$ Less fluctuations
- Fluctuations reduced not eliminated

Difference equation (DE):
$$y[n] = \sum_{k=0}^{M-1} b_k x[n - k]$$

- Weighted running average of M samples
- Causal filter
- M is order of filter

⇒ **Basic building blocks** are: multiplier , adder , unit-delay operator

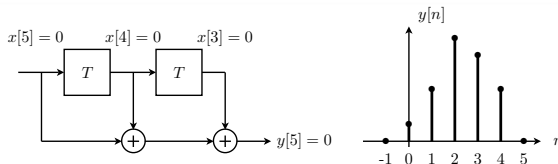
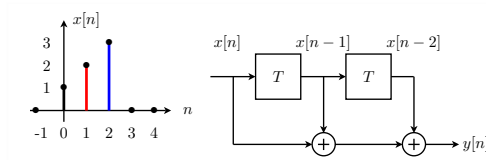


Convolution sum: Example

$$\text{DE} : y[n] = \sum_{k=0}^2 h[k] \cdot x[n - k]$$

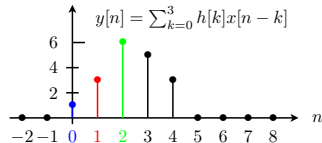
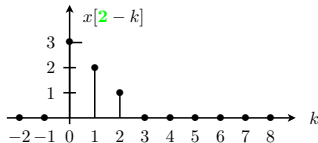
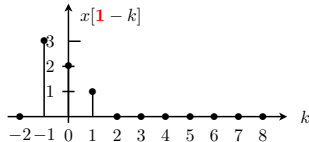
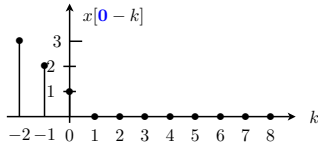
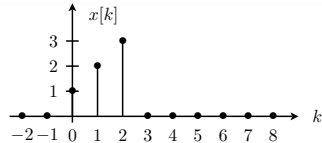
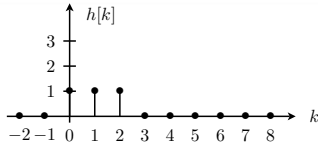
$$\text{Impulse response} : h[n] = \delta[n] + \delta[n - 1] + \delta[n - 2]$$

$$\text{Input} : x[n] = \delta[n] + \textcolor{red}{2}\delta[n - 1] + \textcolor{blue}{3}\delta[n - 2]$$

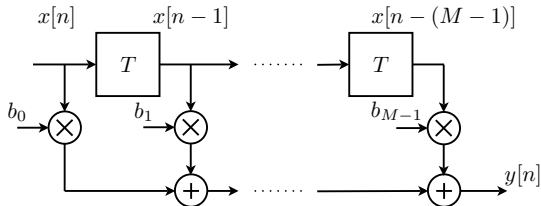


Convolution procedure via plot

FIR-7



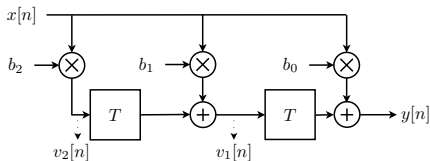
Note for finite filter length : $\text{Length}(y) = \text{Length}(x) + \text{Length}(h) - 1$



$$y[n] = \sum_{k=0}^{M-1} b_k x[n-k]$$

Procedure:

- Plot weights b_k as function of k
- Plot input signal samples $x[k]$ as function of k
- Mirror (reverse) $x[k]$ into $x[-k]$
- Shift mirrored signal $x[-k]$ to index n which results in $x[n-k]$.
- For each new time index n $y[n]$ equals the element by element multiplication of b_k and $x[n-k]$ in the range $k = 0, 1, \dots, M-1$.



$$y[n] = b_0x[n] + v_1[n - 1]$$

$$v_1[n] = b_1x[n] + v_2[n - 1]$$

$$v_2[n] = b_2x[n]$$

$$\Rightarrow v_1[n] = b_1x[n] + b_2x[n - 1]$$

$$\Rightarrow y[n] = b_0x[n] + b_1x[n - 1] + b_2x[n - 2]$$

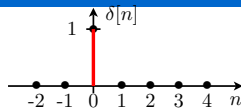
Implementation issues:

Constraints with VLSI or DSP architectures; Finite word-length effects; Memory length; etc

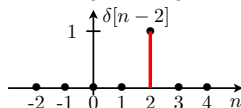
Basic signal: Delta pulse

FIR-10

$$\delta[n] = \begin{cases} 1 & \text{for } n = 0 \\ 0 & \text{for } n \neq 0 \end{cases}$$



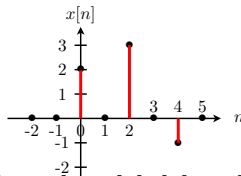
$$\delta[n-2] = \begin{cases} 1 & \text{for } n = 2 \\ 0 & \text{for } n \neq 2 \end{cases}$$



n	...	-2	-1	0	1	2	3	4	5	6	...
$\delta[n]$	0	0	0	1	0	0	0	0	0	0	0
$\delta[n-2]$	0	0	0	0	0	1	0	0	0	0	0

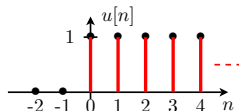
Concept usefull to represent signals and system, e.g.

$$x[n] = 2\delta[n] + 3\delta[n-2] - \delta[n-4]$$



$$x[n] = \sum_k x[k]\delta[n-k] = \dots + x[-1]\delta[n+1] + x[0]\delta[n] + x[1]\delta[n-1] + \dots$$

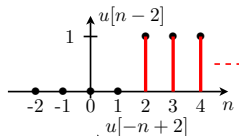
$$u[n] = \begin{cases} 1 & \text{for } n \geq 0 \\ 0 & \text{for } n < 0 \end{cases}$$



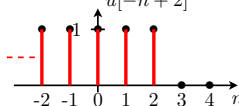
n	-2	-1	0	1	2	3	4	...
$u[n]$	0	0	1	1	1	1	1	...

• Examples:

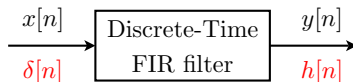
$$u[n-2] = \begin{cases} 1 & \text{for } n \geq 2 \\ 0 & \text{for } n < 2 \end{cases}$$



$$u[-n+2] = \begin{cases} 1 & \text{for } n \leq 2 \\ 0 & \text{for } n > 2 \end{cases}$$



Response for $x[n] = \delta[n]$



Impulse response $h[n]$ of order M FIR:

$$h[n] = \sum_{k=0}^{M-1} b_k \delta[n - k] = \begin{cases} b_n & n = 0, 1, \dots, M-1 \\ 0 & \text{otherwise} \end{cases}$$

In tabular form response to FIR (Finite Impulse Response):

n	$n < 0$	0	1	2	3	\dots	$M-1$	M	$n > M$
$x[n] = \delta[n]$	0	1	0	0	0	0	0	0	0
$y[n] = h[n]$	0	b_0	b_1	b_2	b_3	\dots	b_{M-1}	0	0

Delay or shift input $x[n]$ by n_0 samples, thus $y[n] = x[n - n_0]$

For order M FIR:

$$y[n] = b_0x[n] + b_1x[n - 1] + \cdots b_{n_0}x[n - n_0] + \cdots + b_Mx[n - (M - 1)]$$

Now $y[n] = x[n - n_0]$ when $b_i = 0$ except for $i = n_0$, thus $b_{n_0} = 1$

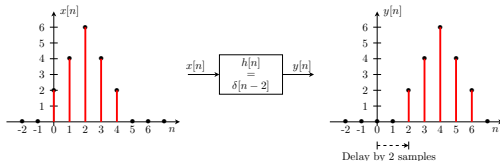
\Rightarrow Impulse response of delay by n_0 samples: $h[n] = \delta[n - n_0]$

Example:

Delay by 2 samples, thus

$h[n] = \delta[n - 2]$ and

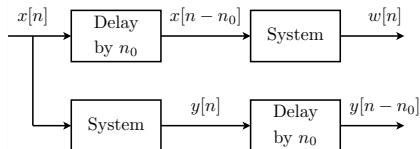
$y[n] = x[n - 2]$



Why LTI: Simplified mathematics and leads to greater insight

Input-output notation: $x[n] \mapsto y[n]$

Time Invariance: $x[n - n_0] \mapsto y[n - n_0]$



System Time-Invariant
 $w[n] = y[n - n_0]$

Example: $y[n] = (x[n])^2$

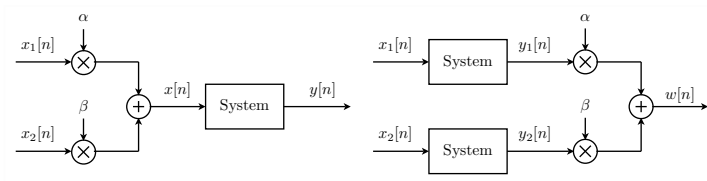
$$x[n] \mapsto y[n] = (x[n])^2 \Rightarrow \text{Delay by one} \Rightarrow y[n - 1] = (x[n - 1])^2$$

$$x[n] \Rightarrow \text{Delay by one} \Rightarrow x[n - 1] \mapsto (x[n - 1])^2 = y[n - 1]$$

Not time-invariant: $y[n] = x[-n]$

Linear system: If $x_1[n] \mapsto y_1[n]$ and $x_2[n] \mapsto y_2[n]$, then

$$x[n] = \alpha x_1[n] + \beta x_2[n] \mapsto y[n] = \alpha y_1[n] + \beta y_2[n]$$



System Linear iff $w[n] = y[n]$

Example: System $y[n] = (x[n])^2$ is **Non**linear since

$$y[n] = (\alpha x_1[n] + \beta x_2[n])^2 \neq \alpha (x_1[n])^2 + \beta (x_2[n])^2$$

Is $y[n] = x[-n]$ linear?

FIR both linear and time-invariant, proof:

$$y[n] = \sum_{k=0}^M b_k x[n-k] \quad \Rightarrow \quad y[\textcolor{red}{n} - \textcolor{red}{n}_0] = \sum_{k=0}^M b_k x[(\textcolor{red}{n} - \textcolor{red}{n}_0) - k]$$

On the other hand with $\textcolor{green}{v}[n] = x[n - n_0]$

$$w[n] = \sum_{k=0}^M b_k \textcolor{green}{v}[n-k] = \sum_{k=0}^M b_k \textcolor{red}{x}[(n-k) - \textcolor{red}{n}_0] = \sum_{k=0}^M b_k x[(n - n_0) - k]$$

FIR is **Linear Time-Invariant** (LTI) system

Main issues of LTI system:

- Impulse response is complete characterization
- Convolution is general formula to compute output from input

We can write $x[n] = \sum_l x[l]\delta[n-l]$

$$(\cdots + x[-2]\delta[n+2] + x[-1]\delta[n+1] + x[0]\delta[n] + x[1]\delta[n-1] + x[2]\delta[n-2] + \cdots)$$

For LTI system:

$$\begin{aligned}\delta[n] \mapsto h[n] &\Rightarrow \delta[n-l] \mapsto h[n-l] \quad \text{for any } l \\ &\Rightarrow x[l]\delta[n-l] \mapsto x[l]h[n-l] \quad \text{for any } l \\ &\Rightarrow x[n] = \sum_l x[l]\delta[n-l] \mapsto y[n] = \sum_l x[l]h[n-l]\end{aligned}$$

No assumptions made about length $x[n]$ or $h[n]$, \Rightarrow

The convolution sum formula:
$$y[n] = \sum_{l=-\infty}^{\infty} x[l]h[n-l]$$

Thus all LTI systems can be represented by a **convolution sum**

Impulse response $h[n]$ of FIR only nonzero for $0 \leq n \leq M$. Thus

$$y[n] = \sum_{l=-\infty}^{\infty} x[l]h[n-l]$$

with index $h[n-l] \in \{0, M\}$ results in

$$y[n] = \sum_{l=n-M}^n x[l]h[n-l]$$

Simple example:

$$x[n] = \delta[n] + 2\delta[n-1] + 3\delta[n-3] \text{ and } h[n] = \delta[n] + \delta[n-1] + \delta[n-2]$$

Convolution as operator:

$$y[n] = x[n] * h[n] = \sum_{l=-\infty}^{\infty} x[l]h[n-l]$$

Convolution with shifted impulse: \Leftrightarrow Delay with n_0

$$x[n] * \delta[n - n_0] = x[n - n_0]$$

Commutative property of $*$:

$$x[n] * h[n] = h[n] * x[n]$$

Proof:

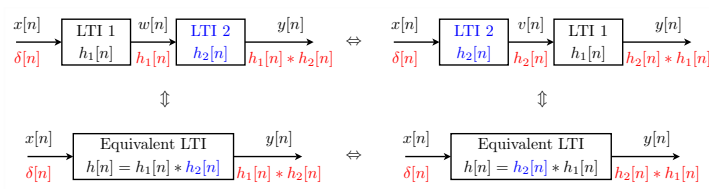
$$\begin{aligned} y[n] &= x[n] * h[n] = \sum_{l=-\infty}^{\infty} x[l]h[n-l] \quad \text{with} \quad k = n - l \\ &= \sum_{k=-\infty}^{-\infty} x[n-k]h[k] = \sum_{k=-\infty}^{\infty} h[k]x[n-k] = h[n] * x[n] \end{aligned}$$

Associative property:

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

Proof:

$$\begin{aligned} x_1[n] * (x_2[n] * x_3[n]) &= \sum_{l=-\infty}^{\infty} x_1[l] \left(\sum_{k=-\infty}^{\infty} x_2[k] x_3[(n-l)-k] \right) \\ \text{with } k = q - l \Rightarrow &= \sum_{l=-\infty}^{\infty} x_1[l] \sum_{q=-\infty}^{\infty} x_2[q-l] x_3[n-q] \\ &= \sum_{q=-\infty}^{\infty} \left(\sum_{l=-\infty}^{\infty} x_1[l] x_2[q-l] \right) x_3[n-q] = (x_1[n] * x_2[n]) * x_3[n] \end{aligned}$$



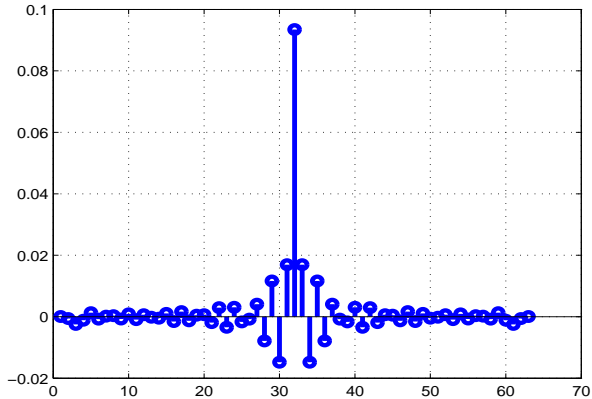
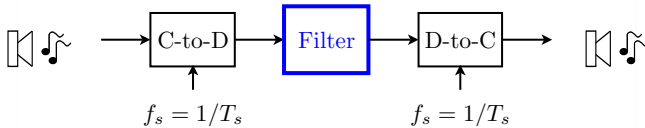
$$\begin{aligned}
 y[n] &= (x[n] * h_1[n]) * h_2[n] = x[n] * (h_1[n] * h_2[n]) \\
 &= x[n] * (h_2[n] * h_1[n]) = (x[n] * h_2[n]) * h_1[n]
 \end{aligned}$$

Example: Evaluate $h[n] = h_1[n] * h_2[n]$ with

$$h_1[n] = \begin{cases} 1 & 0 \leq n \leq 2 \\ 0 & \text{otherwise} \end{cases} \quad h_2[n] = \begin{cases} 1 & 1 \leq n \leq 2 \\ 0 & \text{otherwise} \end{cases}$$

Example FIR filtering

FIR-22



- **Finite Impulse Response (FIR) filters:** Each output sample is sum of **finite number** of weighted samples of input sequence.
- **Difference equation FIR:** $y[n] = \sum_{k=0}^{M-1} b_k x[n-k]$
Weighted running average of M samples; **Causal** filter; M is **order** of FIR filter
- **Unit Impulse Sequence:** $\delta[n] = 1$ for $n = 0$ and zero elsewhere.
Representation signal by unit impulses: $x[n] = \sum_k x[k] \delta[n-k]$
- **Impulse response $h[n]$ of order M FIR:**
$$h[n] = \sum_{k=0}^{M-1} b_k \delta[n-k] = b_n$$
- **Time-Invariance:** If $x[n] \mapsto y[n]$ then $x[n-n_0] \mapsto y[n-n_0]$
- **Linear system:** If $x_1[n] \mapsto y_1[n]$ and $x_2[n] \mapsto y_2[n]$ then
$$x[n] = \alpha x_1[n] + \beta x_2[n] \mapsto y[n] = \alpha y_1[n] + \beta y_2[n]$$

- Main issues Linear Time-Invariance (LTI) system: (FIR is LTI)**
 Completely characterized by **impulse response**; **Convolution** is general formula to compute output from input.
- Convolution sum FIR:** $y[n] = \sum_{k=0}^{M-1} h[k]x[n-k]$
- Convolution operator:** $y[n] = x[n] * h[n] = \sum_{l=-\infty}^{\infty} x[l]h[n-l]$
 - Commutative property:** $x[n] * h[n] = h[n] * x[n]$
 - Associative property:** $(x_1[n] * x_2[n]) * x_3[n] = x_1[n] * (x_2[n] * x_3[n])$
- Cascade of LTI systems:**

