



Sampling

- SAMPLING PROCESS
 - Convert $x(t)$ to numbers $x[n]$
 - “n” is an integer; $x[n]$ is a sequence of values
 - Think of “n” as the storage address in memory
- UNIFORM SAMPLING at $t = nT_s$
 - IDEAL: $x[n] = x(nT_s)$
- SAMPLING RATE (f_s)
 - $f_s = 1/T_s$
 - NUMBER of SAMPLES PER SECOND
 - $T_s = 125$ microsec,
 - $f_s = 8000$ samples/sec (Hz)
- HOW OFTEN ?
 - DEPENDS on FREQUENCY of SINUSOID
 - ANSWERED by SHANNON/NYQUIST Theorem
 - ALSO DEPENDS on “**RECONSTRUCTION**”

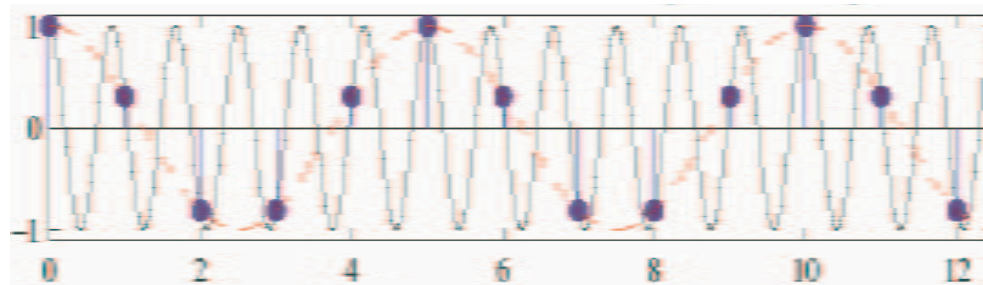


Reconstruction

- Given the samples, draw a sinusoid through the values

When “n” is an integer

$\cos(0.4\pi n)$ or $\cos(2.4\pi n)$



$$x[n] = \cos(0.4\pi n)$$

Time axis n

- CONVERT STREAM of NUMBERS to $x(t)$
 - “CONNECT THE DOTS”
 - INTERPOLATION
 - Math model

$$y(t) = \sum_{n=-\infty}^{\infty} y[n]p(t - nT_s)$$



Linear Filtering

- Background: Signals and Systems

- Let $\delta[\mathbf{k}]$ be a discrete-time impulse function, a.k.a. the Kronecker delta function:

$$\delta[k] = \begin{cases} 1 & k = 0 \\ 0 & k \neq 0 \end{cases}$$


- Impulse response $\mathbf{h}[\mathbf{k}]$: response of a discrete-time LTI system to a discrete impulse function
- We are interested in Finite impulse response filter
 - Non-zero extent of impulse response is finite
 - Can be in continuous time or discrete time
 - Also called a tapped delay line



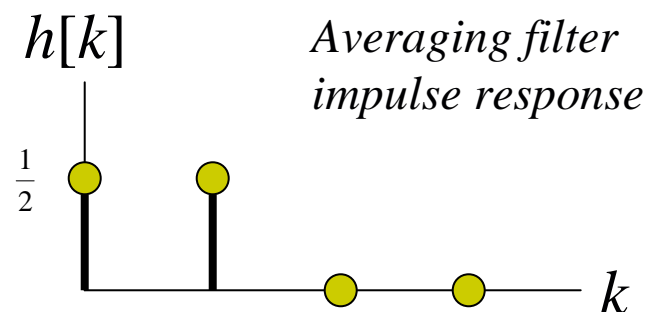
Discrete-time Convolution

- By linear and time-invariant properties, linear convolution
 - For each value of k , compute a different (possibly) infinite summation for $y[k]$

$$y[k] = x[k] * h[k] = \sum_{m=-\infty}^{\infty} x[m] h[k-m] = \sum_{m=-\infty}^{\infty} h[m] x[k-m]$$



$$y[k] = h[0] x[k] + h[1] x[k-1]$$
$$= (x[k] + x[k-1]) / 2$$



Linear Time-Invariant Systems



- The Fundamental Theorem of Linear Systems
 - Inputs a complex sinusoid into an LTI system, the output
 - a complex sinusoid of the same frequency
 - scaled by the response of the LTI system at that frequency
 - Scaling may attenuate the signal and shift it in phase
 - Example in discrete time. Let $x[k] = e^{j\Omega k}$,

$$y[k] = \sum_{m=-\infty}^{\infty} e^{j\Omega(k-m)} h[m] = e^{j\Omega k} \underbrace{\sum_{m=-\infty}^{\infty} h[m] e^{-j\Omega m}}_{H(\Omega)} = e^{j\Omega k} H(\Omega)$$

- $H(\Omega)$ is the discrete-time Fourier transform of $h[k]$ and is also called the frequency response



Frequency Response

- For discrete-time systems, response to complex sinusoid is

frequency response → $e^{j\omega k} \rightarrow H(e^{j\omega}) e^{j\omega k}$

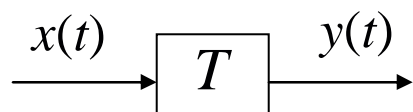
$\cos(\omega k) \rightarrow |H(e^{j\omega})| \cos(\omega k + \angle H(e^{j\omega}))$



Example: Ideal Delay

- Continuous Time

Delay by T seconds



$$y(t) = x(t - T)$$

Impulse response

$$h(t) = \delta(t - T)$$

Frequency response

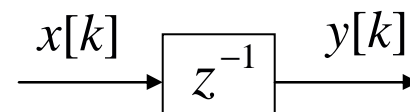
$$H(\Omega) = e^{-j\Omega T}$$

$$|H(\Omega)| = 1$$

$$\angle H(\Omega) = -\Omega T$$

- Discrete Time

Delay by 1 sample



$$y[k] = x[k - 1]$$

Impulse response

$$h[k] = \delta[k - 1]$$

Frequency response

$$H(\omega) = e^{-j\omega}$$

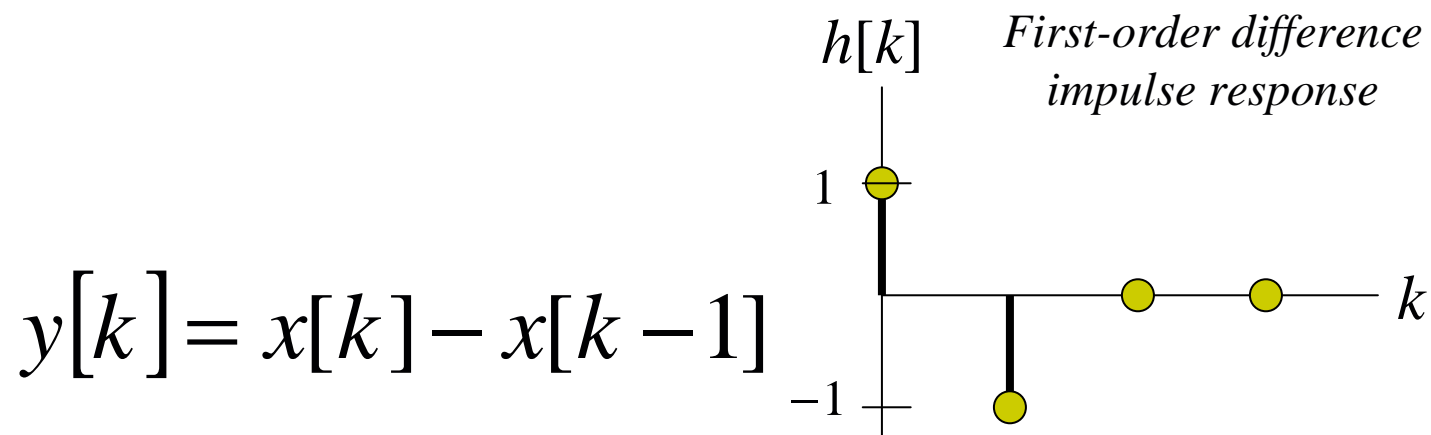
$$|H(\omega)| = 1$$

$$\angle H(\omega) = -\omega$$



First-order difference FIR filter

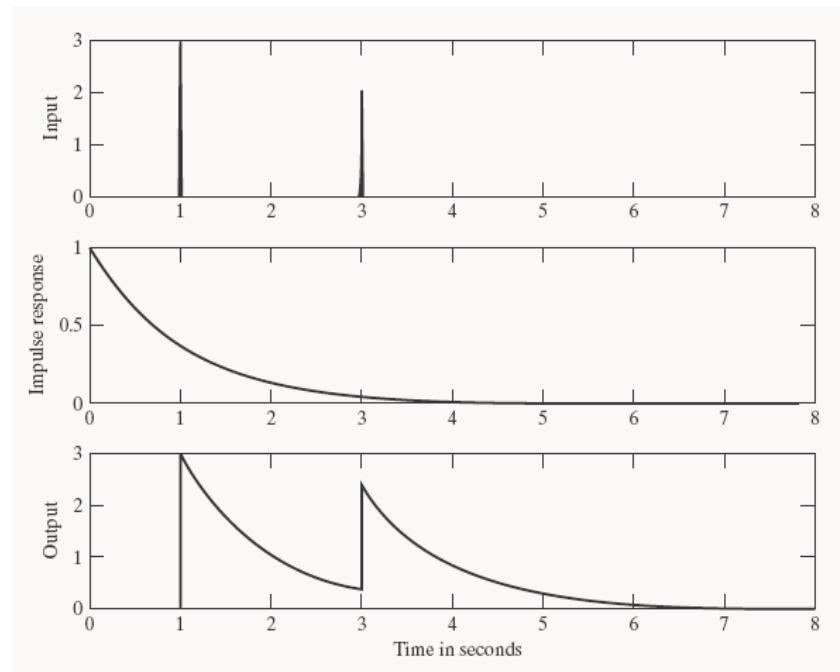
- Highpass filter (sharpening input signal)
 - Impulse response is $\{1, -1\}$





Example

- input: $u(t) = 3\delta(t - 1) + 2\delta(t - 3)$
- impulse response:
 - $h(t) = e^{-t}$ for $t \geq 0$ and 0 Otherwise
- output: $y(t) = 3h(t - 1) + 2h(t - 3)$



Mandrill Demo (*DSP First*)



- From lowpass filter to highpass filter
 - original → blurry → sharpened
- From highpass to lowpass filter
 - original → sharpened → blurry
- Frequencies that are zeroed out (e.g. DC) can never be recovered
- Order of two LTI systems in cascade can be switched under the assumption that the computations are performed in exact precision

Finite Impulse Response (FIR) Filters



- Duration of impulse response $h[k]$ is finite,

$$y[k] = x[k] * h[k] = \sum_{m=-\infty}^{\infty} h[m] x[k-m] = \sum_{m=0}^{N-1} h[m] x[k-m]$$

- Output depends on current input and previous $N-1$ inputs
- N input samples in the vector

$$\{ x[k], x[k-1], \dots, x[k-(N-1)] \}$$

- N nonzero values of the impulse response in vector

$$\{ h[0], h[1], \dots, h[N-1] \}$$

- *What instruction set/architecture features would you add to accelerate FIR filtering?*

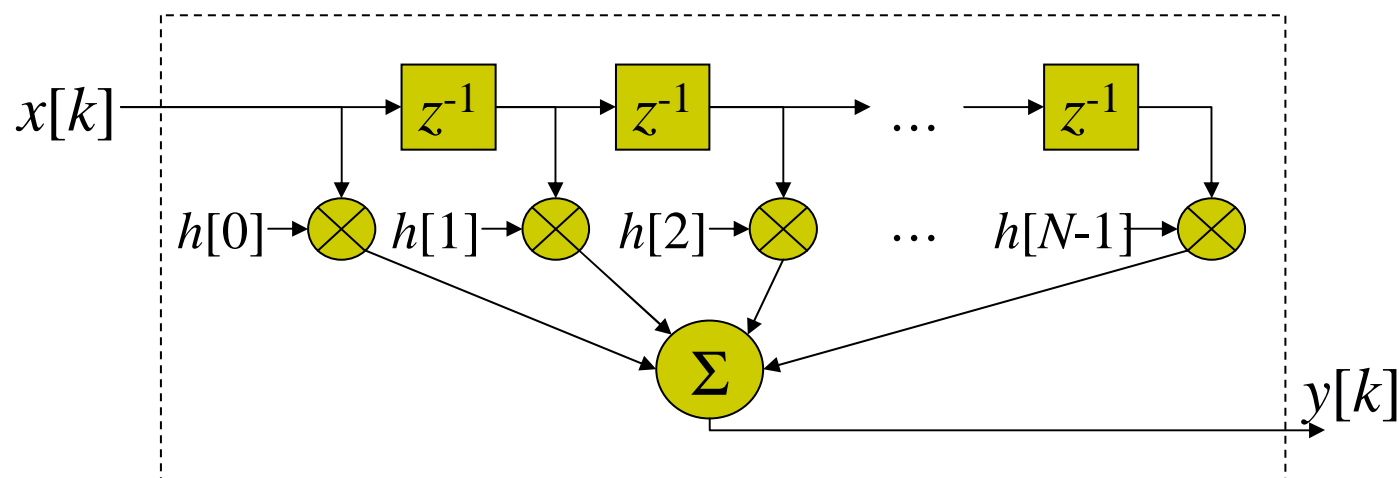


Discrete-time Tapped Delay Line

- Assuming that $h[k]$ has finite duration from $k = 0, \dots, N-1$

$$y[k] = \sum_{m=0}^{N-1} h[m] x[k - m]$$

- Block diagram of an implementation (*Direct Form*)



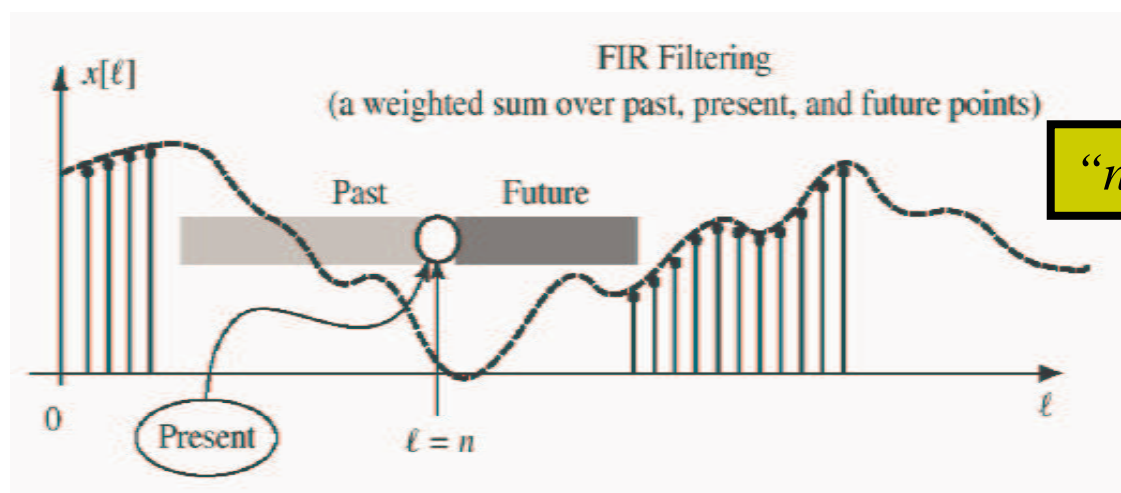


Operation of FIR Filter

- The filter output calculation within a sliding window
- $x[n]$ is a list of numbers indexed by “ n ”

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

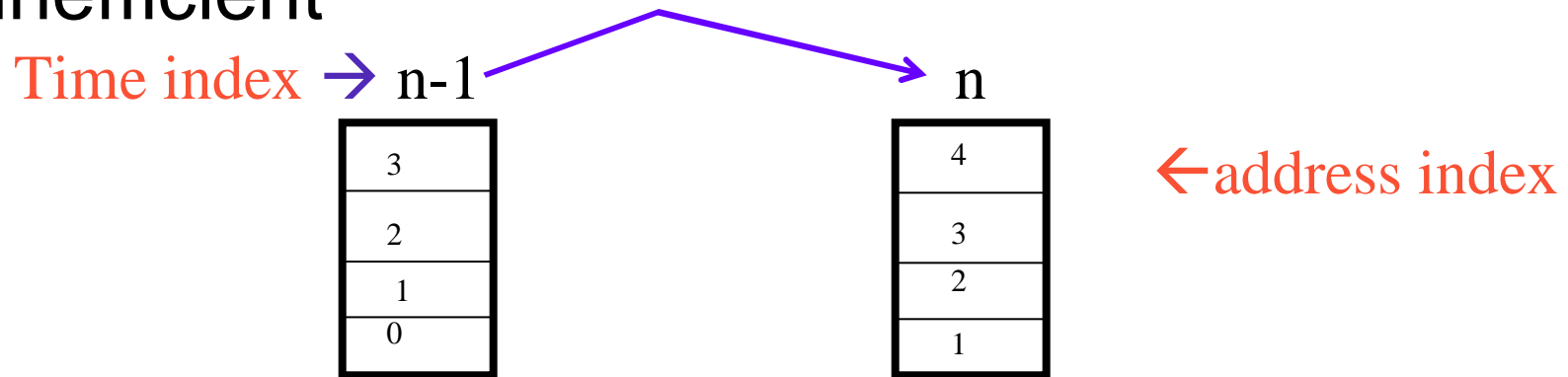
n	$n < -2$	-2	-1	0	1	2	3	4	5	$n > 5$
$x[n]$	0	0	0	2	4	6	4	2	0	0
$y[n]$	0	$\frac{2}{3}$	2	4	$\frac{14}{3}$	4	2	$\frac{2}{3}$	0	0



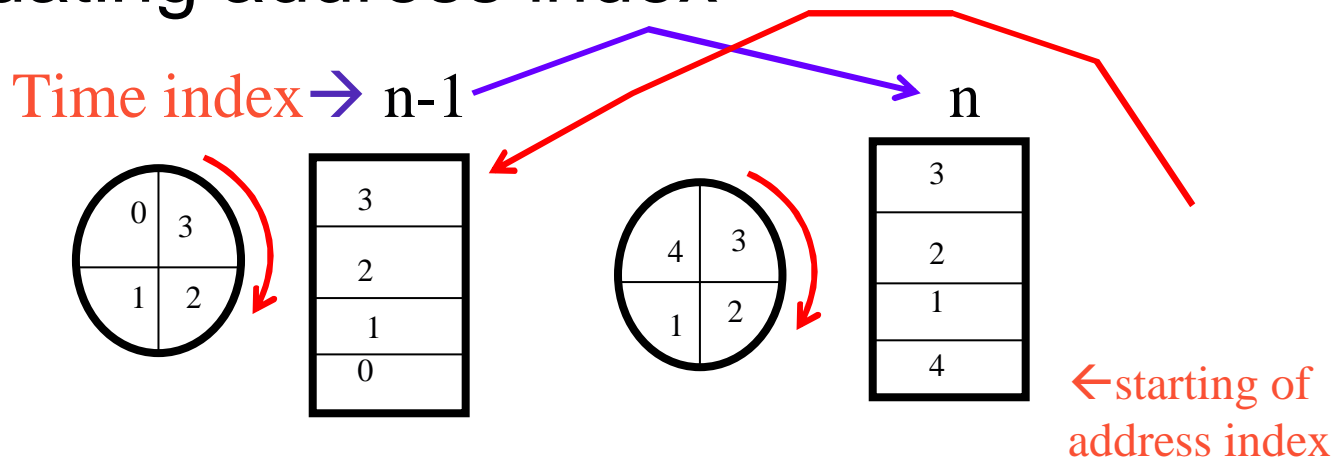
FIR Implementation: Circular Buffer



- Shifting the elements in the entire array is inefficient



- Better approach is to use circular buffers and updating address index



Circular Buffer Implementation in C



- Oldest input sample $x[n-(N-1)]$ is $h[N-1]$ with the largest index
- The newest sample $x[n]$ is multiplied by the $h[0]$ with the smallest index.
- When a new sample is received at time n , it is written over the sample at location $oldest=newest+1$ modulo N and $newest$ is incremented modulo N

Array Index	Coeff. $h[]$	Circ buf $x[]$
0	$h[0]$	$x[n-newest]$
1	$h[1]$	$x[n-newest+1]$
:		
:		$x[n-1]$
<i>newest</i>		$x[n]$
<i>oldest</i>		$x[n-N+1]$
:		
:		
$N-2$	$h[N-2]$	$x[n-newest-2]$
$N-1$	$h[N-1]$	$x[n-newest-1]$

- Thus, data samples are written into the array in a circular fashion.

$$y[n] = \sum_{k=0}^{N-1} h[k] xcirc[mod(newest - k, N)]$$

Convolution Demos



- Johns Hopkins University Demonstrations
 - <http://www.jhu.edu/~signals>
 - Convolution applet to animate convolution of simple signals and hand-sketched signals
 - Convoluting two rectangular pulses of same width gives a triangle whose width is twice the width of the rectangular pulses