## Analog Filter Design

## Part. 4: Discrete time filters Sect. 4-a: General methods

## Discrete time (DT) signals and sequences

- A discrete time signal is defined only on a "countable set" of time instants. The set, or time series, can be either finite or infinite.
- Definition of a discrete time signal can also disregard the actual times at which each value corresponds. In this way we have a pure, ordered sequence of values.

Discrete time signal:  $x(t_n)$ Pure sequence: x(n)

A discrete time signal <u>may</u> be the result of sampling a Continuous Time (CT)signal. Sampling is generally considered to be uniform. Generally, we are interested to DT signals for their capability to represent CT signals.

## Discrete Time Signals (DTS): Linear operators

- As with CT signals, while dealing with DT signals we are interested in Linear, Time Invariant, Causal systems.
- In DT signals the derivative operator is substituted by the difference operator:

CT domain DT domain  

$$\frac{dx(t)}{dt}$$
  $\longleftrightarrow$   $x(n) - x(n-1)$ 

> More generally, the base operator in DT signal is the unity delay operator:

$$x(n) \implies x(n-1)$$
 "T" operator

### **Difference Equations**

In the DT domain, differential equations are substituted by difference equations, where difference between elements of the sequences taken with different indexes (e.g. n, n-1, n-2 etc.) appears.

First ( $\Delta$ ) and second ( $\Delta^2$ ) difference definitions (non causal operators)

$$\Delta x(n) = x(n+1) - x(n)$$
  
$$\Delta^2 x(n) = \Delta x(n+1) - \Delta x(n) = x(n+2) - 2x(n+1) + x(n)$$

Strictly speaking, difference equations are a particular case of recurrence equations:

$$y(n+1) = f[x(n+1), x(n), \dots, x(n-k), y(n+1), y(n), \dots, y(n-k)]$$

#### Analysis tool: Z transform

- In the case of linear, time invariant and causa recurrence equations, a powerful approach is using the Z-transform, which is the analogue of the Laplace transform.
- With the Z-transform, the unity delay operator is transformed into multiplication by Z<sup>-1</sup>: recurrence equations becomes algebraic equations.

$$x(n) \implies x(n-1)$$

$$X(z) \implies z^{-1}X(z)$$

### **Common Z-Transform pairs**



Note that  $u(n)a^n$  is an exponential function:

$$a^n = e^{\ln(a)n}$$

Exponential functions are eigenvectors of the delay operator

## Z-Transform applied to LTI

LTI (Linear Time Invariant) system representation in the DT domain:

$$Y(z) = H(z)X(z) = \frac{N(z)}{D(z)}X(z)$$

Rational transfer function representations

$$H(z) = \frac{b_0 + b_1 z^{-1} + b_2 z^{-2} + \dots + b_M z^{-M}}{a_0 + a_1 z^{-1} + a_2 z^{-2} + \dots + a_N z^N} \quad \text{Negative powers (preferred for synthesis)}$$
$$H(z) = \frac{b_0 z^M + b_1 z^{M-1} + \dots + b_{M-1} z^{M-1} + b_M}{a_0 z^N + a_1 z^{N-1} + \dots + a_{N-1} z^{N-1} + \dots + a_N} z^{N-M} \quad \text{Positive powers}$$

#### Z-Transform: a few properties

$$\lim_{n \to \infty} x(n) = \lim_{z \to 1} (z - 1) X(z)$$
 Final value theorem

 $x(0) = \lim_{z \to \infty} X(z)$  Initial value theorem

DC gain of a transfer function  $H(z) = \lim_{z \to 1} H(z)$ 

$$H(z) = \frac{N(z)}{D(z)} \quad \longleftrightarrow \quad \text{Stability: for all poles (D(z) roots) } z_i:$$
$$|z_i| \le 1$$

#### DT signals: Discrete-Time Fourier Transform (DTFT)

We consider a signal that is sampled with:

Sampling frequency  $f_c = 1/T$ , where T is the sampling interval

$$x(nT) = \int_{-f_c/2}^{+f_c/2} X_F(f) e^{j2\pi fnT} df$$

Decomposition of x(nT) into an integral of complex exponential functions

The frequecy domain is only fc wide since the exponential sequences are invariant for a frequency shift of  $\rm kf_{\rm c}$ 

$$f \to f + kf_c \Rightarrow e^{j2\pi fTn} \to e^{j2\pi fTn} e^{j2\pi kf_cTn} = e^{j2\pi fTn} e^{j2\pi kn} = e^{j2\pi fTn}$$

$$X_F(f) = T \sum_{n=-\infty}^{\infty} x(nT) e^{-j2\pi fnT}$$

With this definition the DTFT is periodical with period=  $f_c$ . Then, only the [- $f_c/2$ ,  $f_c/2$ ] domain is considered.

 $-\frac{f_{s}}{2}$ 

#### DT filters used to replace CT filters



## DT Filters synthesis (Ideal block diagrams)

*IIR: Infinite Impulse Response FIR: Finite Impulse Response* 

Start from a CT state variable filter and replace the CT integrator with DT integrator

Use synthesis approaches that do not need an analog filter as a starting point: use the delayed impulse response properly windowed FIR

IIR



#### Forward and Backward Euler Integrators



#### Simulation of CT state variable filters: result



Discrete time approximation of the CT filter

#### Example: Transform a 1<sup>st</sup> order low pass filter



#### Method 2: z -> s direct substitution

Euler Backward and Forward approximation can be used to find possible substitution formulas:

Forward 
$$H_{I}(z) = \frac{T}{z-1} \leftrightarrow \frac{1}{s} \implies \frac{z-1}{T} \rightarrow s$$
  
Backward  $H_{I}(z) = \frac{zT}{z-1} \leftrightarrow \frac{1}{s} \implies \frac{z-1}{zT} \rightarrow s$ 

# A more precise formula for z-to-s substitution

#### The Bilinear Transformation



## Bilinear transform: characteristics



- Maintains stability
- $\succ$  all s=j $\omega$  are mapped to z belonging to the unit circle
- "Features" of the CT frequency response (e.g. peaks, notches are preserved
- $\succ$  Pre-warping of the CT singularities is necessary for close matching

 $\omega_{i}T$ 

 $\frac{2}{-}$ tan

#### **Pre-warping**

 $\rightarrow \omega_i$ 

#### DT filter design from the impulse response



#### DT filter design from the impulse response



#### Effect of windowing



#### High pass and band-pass from low-pass





## Architecture of a generic DT filter

Coefficients  $b_i$  and  $a_i$  are called the "taps" of the filter and correspond to the coefficients of the numerator and denominator of H(z), according to:

$$H(z) = \frac{b_0 + b_1 z^{-1} + b_2 z^{-2} + \dots + b_M z^{-M}}{1 + a_1 z^{-1} + a_2 z^{-2} + \dots + a_N z^{-M}}$$

In a FIR filter the coefficients ai are all equal to zero, that is the denominator is = 1