



<http://elec3004.com>

Filter Analysis & Discrete Systems

Thursday: **z-Transforms**

Next Tuesday: **Intro to Control (With Dr. Pounds!!)**

ELEC 3004: Digital Linear Systems: Signals & Controls

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Lecture 6

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Lecture Schedule:

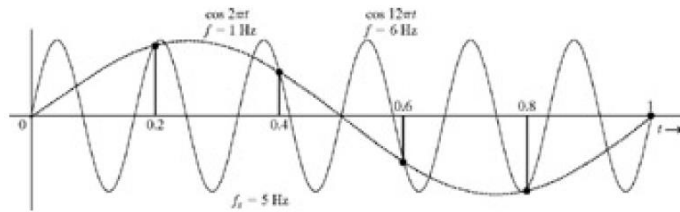
Week	Date	Lecture Title
1	4-Mar	Introduction & Systems Overview
	6-Mar	[Linear Dynamical Systems]
2	11-Mar	Signals as Vectors & Systems as Maps
	13-Mar	[Signals]
3	18-Mar	Sampling & Data Acquisition & Antialiasing Filters
	20-Mar	[Sampling]
4	25-Mar	System Analysis & Convolution
	27-Mar	[Convolution & FT]
5	1-Apr	Frequency Response & Filter Analysis
	3-Apr	[Filters]
6	8-Apr	Discrete Systems & Z-Transforms
	10-Apr	[Z-Transforms]
7	15-Apr	Introduction to Control
	17-Apr	[Feedback]
8	29-Apr	Digital Filters
	1-May	[Digital Filters]
9	6-May	Introduction to Digital Control
	8-May	[Digital Control]
10	13-May	Stability of Digital Systems
	15-May	[Stability]
11	20-May	State-Space
	22-May	Controllability & Observability
12	27-May	PID Control & System Identification
	29-May	Digital Control System Hardware
13	3-Jun	Applications in Industry & Information Theory & Communications
	5-Jun	Summary and Course Review



ELEC 3004: Systems

8 April 2014 - 2

Refresher: Aliasing & Sampling

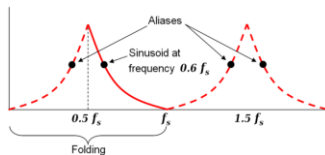


- Nyquist:

$$f_h < \frac{f_s}{2}$$

- Spectral Folding:

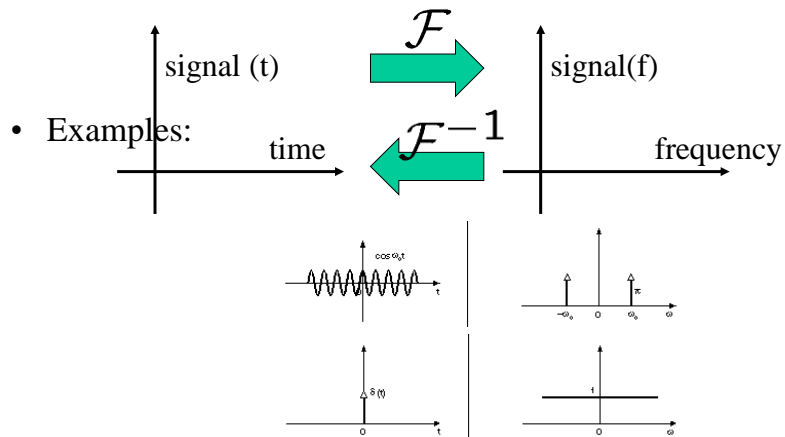
$$f_{\text{image}}(N) = f - Nf_s.$$



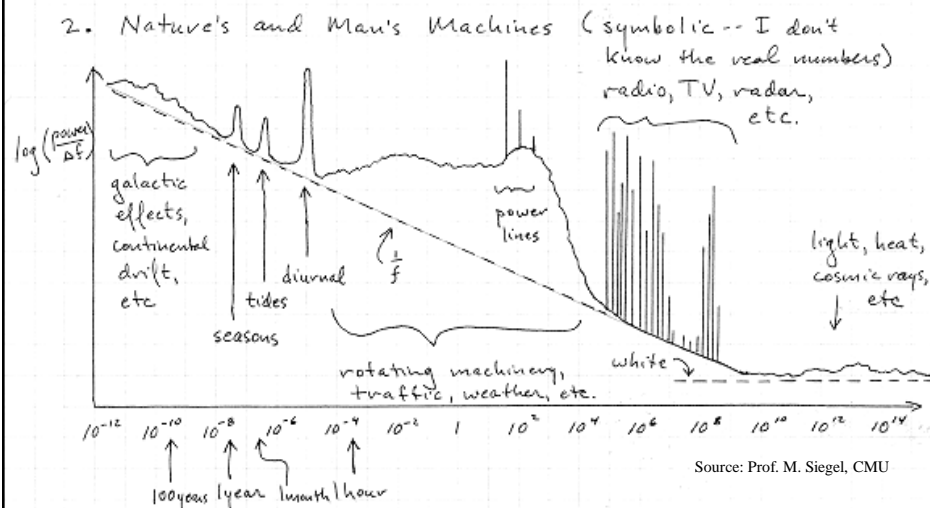
First Some Noise!

Frequency

- How often the signal repeats
- Can be analyzed through Fourier Transform



Noise



Note: this picture illustrates the concepts but it is not quantitatively precise



Noise [2]

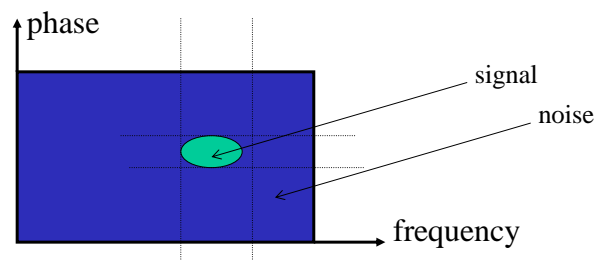
Various Types:

- Thermal (white):
 - Johnson noise, from thermal energy inherent in mass.
- Flicker or 1/f noise:
 - Pink noise
 - More noise at lower frequency
- Shot noise:
 - Noise from quantum effects as current flows across a semiconductor barrier
- Avalanche noise:
 - Noise from junction at breakdown (circuit at discharge)



How to beat the noise

- Filtering (Narrow-banding): Only look at particular portion of **frequency space**
- Multiple measurements ...
- Other (modulation, etc.) ...



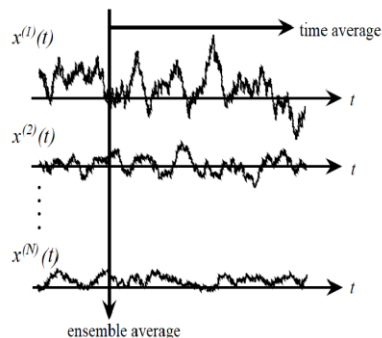
Noise \subseteq Uncertainty

- **Uncertainty:**
All measurement has some approximation
 - A. Statistical uncertainty: quantified by mean & variance
 - B. Systematic uncertainty: non-random error sources
- **Law of Propagation of Uncertainty**
 - Combined uncertainty is root squared

$$u_c = \sqrt{u_1^2 + u_2^2 + \dots + u_n^2}$$



Treating Uncertainty with Multiple Measurements



1. **Over time:** multiple readings of a quantity over time
 - “stationary” or “ergodic” system
 - Sometimes called “integrating”
 2. **Over space:** **single** measurement (summed) from multiple sensors each distributed in space
 3. **Same Measurand:** multiple measurements take of the **same observable quantity** by multiple, related instruments
e.g., measure position & velocity simultaneously
- Basic “sensor fusion”

$$\sigma_{\text{final}} = [\sigma_1^{-1} + \sigma_2^{-1} + \dots + \sigma_n^{-1}]^{-1}$$



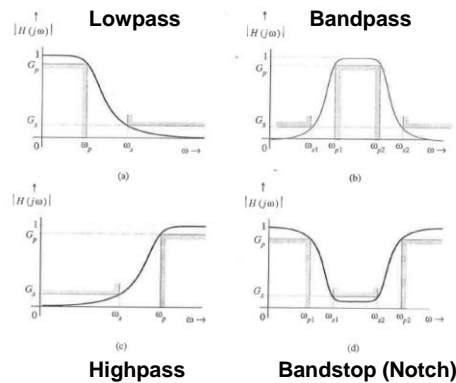
Multiple Measurements Example

- What time was it when this picture was taken?
- What was the temperature in the room?



Now: (analog) Filters!

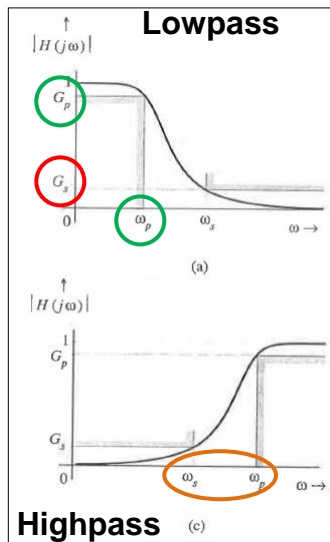
Filters



- *Frequency-shaping filters*: LTI systems that change the shape of the spectrum
- *Frequency-selective filters*: Systems that pass some frequencies undistorted and attenuate others



Filters



Specified Values:

- G_p = minimum passband gain

Typically:

$$G_p = \frac{1}{\sqrt{2}} = -3dB$$

- G_s = maximum stopband gain

- **Low**, not zero (sorry!)
- For realizable filters, the gain cannot be zero over a finite band (Paley-Wiener condition)

- **Transition Band**:

transition from the passband to the stopband $\rightarrow \omega_p \neq \omega_s$



Filter Design & z-Transform

Filter Type	Mapping	Design Parameters
Low-pass	$z^{-1} \rightarrow \frac{z^{-1} - \alpha}{1 - \alpha z^{-1}}$	$\alpha = \frac{\sin[(\omega_c - \omega'_c)/2]}{\sin[(\omega_c + \omega'_c)/2]}$ $\omega'_c = \text{desired cutoff frequency}$
High-pass	$z^{-1} \rightarrow -\frac{z^{-1} + \alpha}{1 + \alpha z^{-1}}$	$\alpha = -\frac{\cos[(\omega_c + \omega'_c)/2]}{\cos[(\omega_c - \omega'_c)/2]}$ $\omega'_c = \text{desired cutoff frequency}$
Bandpass	$z^{-1} \rightarrow -\frac{z^{-2} - [2\alpha\beta/(\beta + 1)]z^{-1} + [(\beta - 1)/(\beta + 1)]}{[(\beta - 1)/(\beta + 1)]z^{-2} - [2\alpha\beta/(\beta + 1)]z^{-1} + 1}$	$\alpha = \frac{\cos[(\omega_{c2} + \omega_{c1})/2]}{\cos[(\omega_{c2} - \omega_{c1})/2]}$ $\beta = \cot[(\omega_{c2} - \omega_{c1})/2] \tan(\omega_c/2)$ $\omega_{c1} = \text{desired lower cutoff frequency}$ $\omega_{c2} = \text{desired upper cutoff frequency}$
Bandstop	$z^{-1} \rightarrow \frac{z^{-2} - [2\alpha/(\beta + 1)]z^{-1} + [(1 - \beta)/(1 + \beta)]}{[(1 - \beta)/(1 + \beta)]z^{-2} - [2\alpha/(\beta + 1)]z^{-1} + 1}$	$\alpha = \frac{\cos[(\omega_{c1} + \omega_{c2})/2]}{\cos[(\omega_{c1} - \omega_{c2})/2]}$ $\beta = \tan[(\omega_{c2} - \omega_{c1})/2] \tan(\omega_c/2)$ $\omega_{c1} = \text{desired lower cutoff frequency}$ $\omega_{c2} = \text{desired upper cutoff frequency}$



Butterworth Filters

- Butterworth: Smooth in the pass-band
- The amplitude response $|H(j\omega)|$ of an n^{th} order Butterworth low pass filter is given by:

$$|H(j\omega)| = \frac{1}{\sqrt{1 + \left(\frac{\omega}{\omega_c}\right)^{2n}}}$$

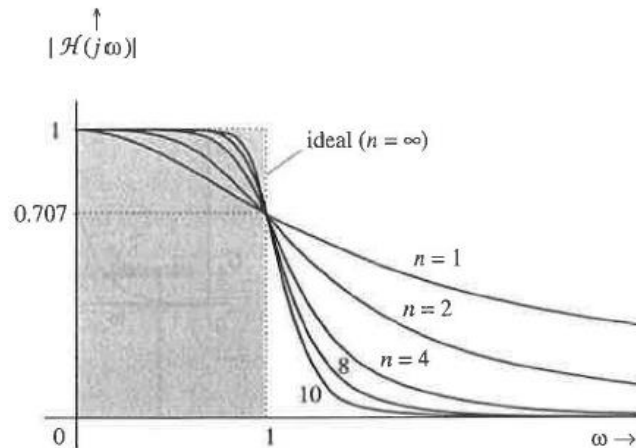
- The normalized case ($\omega_c=1$)

$$|\mathcal{H}(j\omega)| = \frac{1}{\sqrt{1 + \omega^{2n}}} \quad \Rightarrow \quad \mathcal{H}(j\omega)\mathcal{H}(-j\omega) = |\mathcal{H}(j\omega)|^2 = \frac{1}{1 + \omega^{2n}}$$

Recall that: $|H(j\omega)|^2 = H(j\omega)H(-j\omega)$

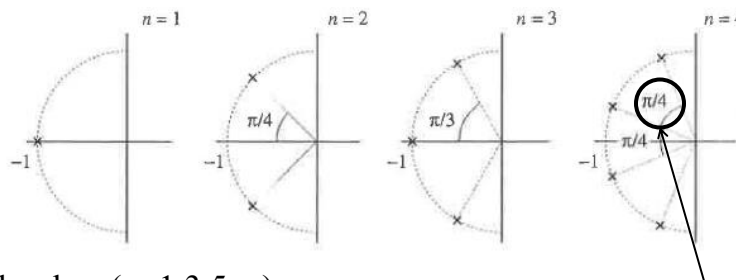


Butterworth Filters



Butterworth Filters of Increasing Order: Seeing this Using a Pole-Zero Diagram

- Increasing the order, increases the number of poles:



→ Odd orders ($n=1, 3, 5, \dots$):

- Have a pole on the Real Axis

→ Even orders ($n=2, 4, 6, \dots$):

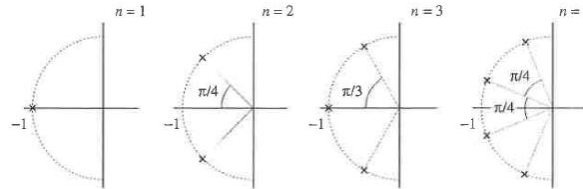
- Have a pole on the off axis

Angle between
poles:

$$\frac{\pi}{n}$$



Butterworth Filters: Pole-Zero Diagram



- Since $H(s)$ is stable and causal, its poles must lie in the LHP
- Poles of $-H(s)$ are those in the RHP
- Poles lie on the unit circle (for a normalized filter)

$$\rightarrow H(s) = \frac{1}{(s - s_1)(s - s_2) \dots (s - s_n)}$$

Where:

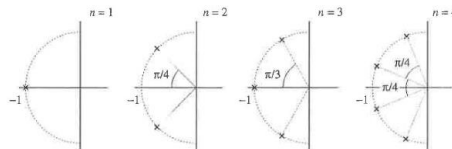
$$s_k = e^{j\frac{\pi}{2n}(2k+n-1)}$$

$$= \cos \frac{\pi}{2n}(2k+n-1) + j \sin \frac{\pi}{2n}(2k+n-1) \quad k = 1, 2, 3, \dots, n$$

n is the order of the filter



Butterworth Filters: 4th Order Filter Example



- Plugging in for $n=4$, $k=1, \dots, 4$:

$$\mathcal{H}(s) = \frac{1}{(s + 0.3827 - j0.9239)(s + 0.3827 + j0.9239)(s + 0.9239 - j0.3827)(s + 0.9239 + j0.3827)}$$

$$= \frac{1}{(s^2 + 0.7654s + 1)(s^2 + 1.8478s + 1)}$$

$$= \frac{1}{s^4 + 2.6131s^3 + 3.4142s^2 + 2.6131s + 1}$$

- We can generalize \rightarrow Butterworth Table

n	a_1	a_2	a_3	a_4	a_5
2	1.41421356				
3	2.00000000	2.00000000			
4	2.61312593	3.41421356	2.61312593		
5	3.23606798	5.23606798	5.23606798	3.23606798	
6	3.86370331	7.46410162	9.14162017	7.46410162	3.86370331

This is for 3dB bandwidth at $\omega_c=1$



Butterworth Filters: Scaling Back (from Normalized)

- Start with Normalized equation & Table
- Replace ω with $\frac{\omega}{\omega_c}$ in the filter equation
- For example:
for $f_c=100\text{Hz} \rightarrow \omega_c=200\pi \text{ rad/sec}$

From the Butterworth table: for $n=2$, $a_1=\sqrt{2}$
Thus:

$$H(s) = \frac{1}{\left(\frac{s}{200\pi}\right)^2 + \sqrt{2}\left(\frac{s}{200\pi}\right) + 1}$$

$$= \frac{1}{s^2 + 200\pi\sqrt{2}s + 40,000\pi^2}$$



Butterworth: Determination of Filter Order

- Define G_x as the gain of a lowpass Butterworth filter at $\omega = \omega_x$
- Then:

$$\hat{G}_x = 20 \log_{10} |H(j\omega_x)| = -10 \log \left[1 + \left(\frac{\omega_x}{\omega_c} \right)^{2n} \right]$$

And thus:

$$\hat{G}_p = -10 \log \left[1 + \left(\frac{\omega_p}{\omega_c} \right)^{2n} \right]$$

$$\hat{G}_s = -10 \log \left[1 + \left(\frac{\omega_s}{\omega_c} \right)^{2n} \right]$$

Or alternatively:

$$\omega_c = \frac{\omega_p}{\left[10^{-\hat{G}_p/10} - 1 \right]^{1/2n}} \quad \& \quad \omega_c = \frac{\omega_s}{\left[10^{-\hat{G}_s/10} - 1 \right]^{1/2n}}$$

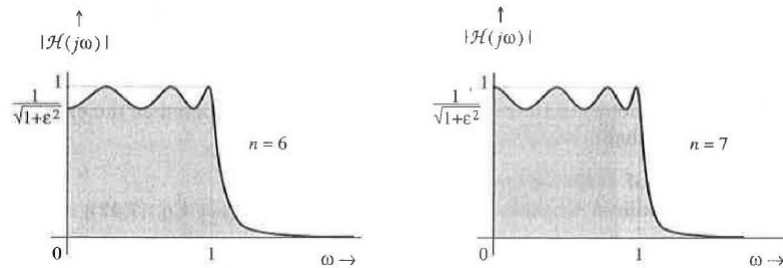
Solving for n gives:

$$n = \frac{\log \left[\left(10^{-\hat{G}_s/10} - 1 \right) / \left(10^{-\hat{G}_p/10} - 1 \right) \right]}{2 \log(\omega_s / \omega_p)}$$

PS. See Lathi 4.10 (p. 453) for an example in MATLAB



Chebyshev Filters



- **equal-ripple:**
Because all the ripples in the passband are of equal height
- If we reduce the ripple, the passband behaviour improves, but it does so at the cost of stopband behaviour



Chebyshev Filters

- Chebyshev Filters: Provide tighter transition bands (sharper cutoff) than the same-order Butterworth filter, but this is achieved at the expense of inferior passband behavior (rippling)
- ➔ For the lowpass (LP) case: at higher frequencies (in the stopband), the Chebyshev filter gain is smaller than the comparable Butterworth filter gain by about **6(n - 1) dB**
- The amplitude response of a normalized Chebyshev lowpass filter is:

$$|\mathcal{H}(j\omega)| = \frac{1}{\sqrt{1 + \epsilon^2 C_n^2(\omega)}}$$

Where $C_n(\omega)$, the n th-order Chebyshev polynomial, is given by:

$$C_n(\omega) = \cos(n \cos^{-1} \omega)$$

$$C_n(\omega) = \cosh(n \cosh^{-1} \omega)$$

and where C_n is given by:

n	$C_n(\omega)$
0	1
1	ω
2	$2\omega^2 - 1$
3	$4\omega^3 - 3\omega$
4	$8\omega^4 - 8\omega^2 + 1$
5	$16\omega^5 - 20\omega^3 + 5\omega$
6	$32\omega^6 - 48\omega^4 + 18\omega^2 - 1$



Normalized Chebyshev Properties

- It's normalized: The passband is $0 < \omega < 1$
- **Amplitude response:** has **ripples** in the passband and is **smooth** (monotonic) in the stopband
- **Number of ripples:** there is a total of n maxima and minima over the passband $0 < \omega < 1$
- $C_n^2(0) = \begin{cases} 0, & n : \text{odd} \\ 1, & n : \text{even} \end{cases} \quad \Rightarrow \quad |H(0)| = \begin{cases} 1, & n : \text{odd} \\ \frac{1}{\sqrt{1+\epsilon^2}}, & n : \text{even} \end{cases}$
- ϵ : ripple height $\rightarrow r = \sqrt{1 + \epsilon^2}$
- The Amplitude at $\omega=1$: $\frac{1}{r} = \frac{1}{\sqrt{1 + \epsilon^2}}$
- For Chebyshev filters, the ripple r dB takes the place of G_p



Determination of Filter Order

- The gain is given by: $\hat{G} = -10 \log [1 + \epsilon^2 C_n^2(\omega)]$
- Thus, the gain at ω_s is: $\epsilon^2 C_n^2(\omega_s) = 10^{-\hat{G}_s/10} - 1$

- Solving:

$$n = \frac{1}{\cosh^{-1}(\omega_s)} \cosh^{-1} \left[\frac{10^{-\hat{G}_s/10} - 1}{10^{\hat{\epsilon}^2/10} - 1} \right]^{1/2}$$

- General Case:

$$n = \frac{1}{\cosh^{-1}(\omega_s/\omega_p)} \cosh^{-1} \left[\frac{10^{-\hat{G}_s/10} - 1}{10^{\hat{\epsilon}^2/10} - 1} \right]^{1/2}$$



Chebyshev Pole Zero Diagram

- Whereas **Butterworth** poles lie on a **semi-circle**,
The poles of an n^{th} -order normalized **Chebyshev** filter lie on a **semiellipse** of the major and minor semiaxes:

$$a = \sinh \left(\frac{1}{n} \sinh^{-1} \left(\frac{1}{\epsilon} \right) \right) \quad \& \quad b = \cosh \left(\frac{1}{n} \sinh^{-1} \left(\frac{1}{\epsilon} \right) \right)$$

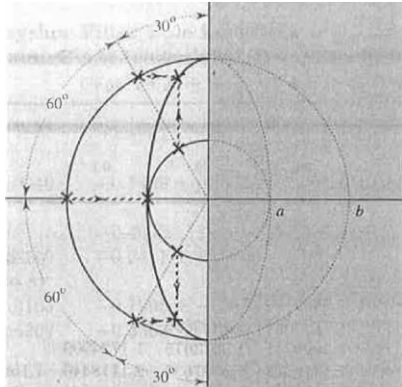
And the poles are at the locations:

$$H(s) = \frac{1}{(s - s_1)(s - s_2) \dots (s - s_n)}$$

$$s_k = -\sin \left[\frac{(2k-1)\pi}{2n} \right] \sinh x + j \cos \left[\frac{(2k-1)\pi}{2n} \right] \cosh x, \quad k = 1, \dots, n$$



Ex: Chebyshev Pole Zero Diagram for $n=3$



Procedure:

1. Draw two semicircles of radii **a** and **b** (from the previous slide).
2. Draw radial lines along the corresponding Butterworth angles (π/n) and locate the n^{th} -order Butterworth poles (shown by crosses) on the two circles.
3. The location of the k^{th} Chebyshev pole is the intersection of the horizontal projection and the vertical projection from the corresponding k^{th} Butterworth poles on the outer and the inner circle, respectively.



Chebyshev Values / Table

$$\mathcal{H}(s) = \frac{K_n}{C'_n(s)} = \frac{K_n}{s^n + a_{n-1}s^{n-1} + \dots + a_1s + a_0}$$

$$K_n = \begin{cases} a_0 & n \text{ odd} \\ \frac{a_0}{\sqrt{1+\epsilon^2}} = \frac{a_0}{10^{\hat{r}/20}} & n \text{ even} \end{cases}$$

n	a_0	a_1	a_2	a_3
1	1.9652267			
2	1.1025103	1.0977343		
3	0.4913067	1.2384092	0.9883412	
4	0.2756276	0.7426194	1.4539248	0.9528114

1 db ripple
($\hat{r} = 1$)



Other Filter Types:

Chebyshev Type II = Inverse Chebyshev Filters

- Chebyshev filters passband has ripples and the stopband is smooth.
- Instead:** this has **passband** have **smooth** response and **ripples** in the stopband.
- Exhibits maximally flat passband response and equi-ripple stopband
- **Cheby2** in MATLAB

$$|\mathcal{H}(\omega)|^2 = 1 - |\mathcal{H}_C(1/\omega)|^2 = \frac{\epsilon^2 C_n^2(1/\omega)}{1 + \epsilon^2 C_n^2(1/\omega)}$$

Where: \mathcal{H}_c is the Chebyshev filter system from before

- Passband behavior, especially for small ω , is **better** than Chebyshev
- Smallest transition band** of the 3 filters (Butter, Cheby, Cheby2)
- Less time-delay (or phase loss) than that of the **Chebyshev**
- Both needs the **same order n** to meet a set of specifications.
- \$\$\$ (or number of elements):
Cheby < Inverse Chebyshev < Butterworth (of the same **performance** [not order])



Other Filter Types: Elliptic Filters (or Cauer) Filters

- Allow **ripple** in **both** the passband and the stopband,
→ we can achieve **tighter** transition band

$$|\mathcal{H}(j\omega)| = \frac{1}{\sqrt{1 + \epsilon^2 R_n^2(\omega)}}$$

Where: R_n is the n^{th} -order Chebyshev rational function determined from a given ripple spec.
 ϵ controls the ripple

$$G_p = \frac{1}{\sqrt{1 + \epsilon^2}}$$

- Most efficient (η)
 - the **largest ratio** of the passband gain to stopband gain
 - **or** for a given ratio of passband to stopband gain, it requires the **smallest transition band**

→ in MATLAB: **ellipord** followed by **ellip**



In Summary

Filter Type	Passband Ripple	Stopband Ripple	Transition Band	MATLAB Design Command
Butterworth	No	No	Loose	butter
Chebyshev	Yes	No	Tight	cheby
Chebyshev Type II (Inverse Chebyshev)	No	Yes	Tight	cheby2
Elliptic	Yes	Yes	Tightest	ellip



Linear, Discrete Dynamical Systems

Linear Difference Equations

$$u_k = f(e_0, \dots, e_k; u_0, \dots, u_{k-1}).$$

$$u_k = -a_1 u_{k-1} - a_2 u_{k-2} - \dots - a_n u_{k-n} + b_0 e_k + b_1 e_{k-1} + \dots + b_m e_{k-m}.$$

$$\nabla u_k = u_k - u_{k-1} \quad (\text{first difference}),$$

$$\nabla^2 u_k = \nabla u_k - \nabla u_{k-1} \quad (\text{second difference}),$$

$$\nabla^n u_k = \nabla^{n-1} u_k - \nabla^{n-1} u_{k-1} \quad (nth \text{ difference}).$$

$$u_k = u_k,$$

$$u_{k-1} = u_k - \nabla u_k,$$

$$u_{k-2} = u_k - 2\nabla u_k + \nabla^2 u_k.$$

$$a_2 \nabla^2 u_k - (a_1 + 2a_2) \nabla u_k + (a_2 + a_1 + 1) u_k = b_0 e_k.$$



Assume a form of the solution

z^k :

- k: “order of difference”
- k: delay

$$Az^k = Az^{k-1} + Az^{k-2}.$$

$$1 = z^{-1} + z^{-2}$$

$$z^2 = z + 1.$$



z Transforms

(Digital Systems Made eZ)

The z-Transform

- It is defined by:

$$z = re^{j\omega}$$

Or in the Laplace domain:

$$z = e^{sT}$$

- Thus: $Y(z) = \sum_{k=-\infty}^{\infty} h[k]z^{-k}$ or $y[n] \xleftrightarrow{Z} Y(z)$

- I.E., It's a discrete version of the Laplace:

$$f(kT) = e^{-akT} \Rightarrow \mathcal{Z}\{f(k)\} = \frac{z}{z - e^{-aT}}$$



The z-transform

- In practice, you'll use look-up tables or computer tools (ie. Matlab) to find the z-transform of your functions

$F(s)$	$F(kt)$	$F(z)$
$\frac{1}{s}$	1	$\frac{z}{z-1}$
$\frac{1}{s^2}$	kT	$\frac{Tz}{(z-1)^2}$
$\frac{1}{s+a}$	e^{-akT}	$\frac{z}{z-e^{-aT}}$
$\frac{1}{(s+a)^2}$	kTe^{-akT}	$\frac{zTe^{-aT}}{(z-e^{-aT})^2}$
$\frac{1}{s^2+a^2}$	$\sin(akT)$	$\frac{z \sin aT}{z^2 - (2 \cos aT)z + 1}$



An example!

- Back to our difference equation:

$$y(k) = x(k) + Ax(k-1) - By(k-1)$$

becomes

$$\begin{aligned} Y(z) &= X(z) + Az^{-1}X(z) - Bz^{-1}Y(z) \\ (z+B)Y(z) &= (z+A)X(z) \end{aligned}$$

which yields the transfer function:

$$\frac{Y(z)}{X(z)} = \frac{z+A}{z+B}$$

Note: It is also not uncommon to see systems expressed as polynomials in z^{-n}



This looks familiar...

- Compare:

$$\frac{Y(s)}{X(s)} = \frac{s+2}{s+1} \quad \text{vs} \quad \frac{Y(z)}{X(z)} = \frac{z+A}{z+B}$$

How are the Laplace and z domain representations related?



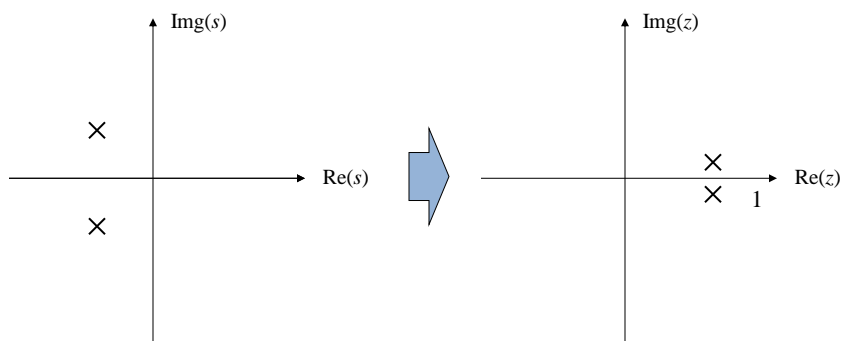
Linearity:

$$a_1y_1[n] + a_2y_2[n] \xleftrightarrow{Z} a_1Y_1(z) + a_2Y_2(z)$$



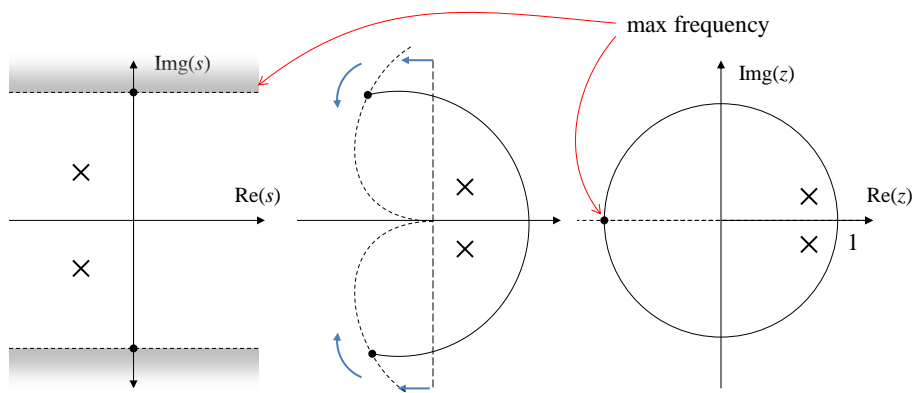
The z -Plane

- z -domain poles and zeros can be plotted just like s -domain poles and zeros:



Deep insight #1

The mapping between continuous and discrete poles and zeros acts like a distortion of the plane



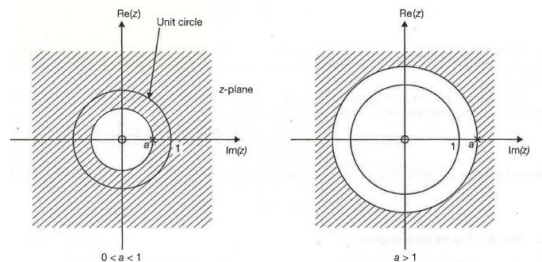
Region of Convergence

- For the convergence of $X(z)$ we require that

$$\sum_{n=0}^{\infty} |az^{-1}|^n < \infty$$

- Thus, the ROC is the range of values of z for which $|az^{-1}| < 1$ or, equivalently, $|z| > |a|$. Then

$$X(z) = \frac{z}{z-a} \quad |z| > |a|$$



Z-Transform Properties: Time Shifting

$$y[n - n_0] \xleftrightarrow{\mathcal{Z}} z^{-n_0} Y(z)$$

$$\begin{aligned} y_2[n] &= y[n - n_0] \\ Y_2(e^{j\omega}) &= \sum_{k=-\infty}^{\infty} y[k - n_0] z^{-k} \\ &= \sum_{l=-\infty}^{\infty} y[l] z^{-(l+n_0)} \\ &= z^{-n_0} Y(z) \end{aligned}$$

- Two Special Cases:
- z^{-1} : the *unit-delay operator*:

$$x[n - 1] \leftrightarrow z^{-1} X(z) \quad R' = R \cap \{0 < |z|\}$$

- z : *unit-advance operator*:

$$x[n + 1] \leftrightarrow z X(z) \quad R' = R \cap \{|z| < \infty\}$$



More Z-Transform Properties

- Time Reversal

$$x[n] \leftrightarrow X(z) \quad \text{ROC} = R$$

$$x[-n] \leftrightarrow X\left(\frac{1}{z}\right) \quad R' = \frac{1}{R}$$

- Multiplication by z^n

$$x[n] \leftrightarrow X(z) \quad \text{ROC} = R$$

$$z_0^n x[n] \leftrightarrow X\left(\frac{z}{z_0}\right) \quad R' = |z_0| R$$

- Multiplication by n (or Differentiation in z):

$$x[n] \leftrightarrow X(z) \quad \text{ROC} = R$$

$$nx[n] \leftrightarrow -z \frac{dX(z)}{dz} \quad R' = R$$

- Convolution

$$x_1[n] \leftrightarrow X_1(z) \quad \text{ROC} = R_1$$

$$x_2[n] \leftrightarrow X_2(z) \quad \text{ROC} = R_2$$

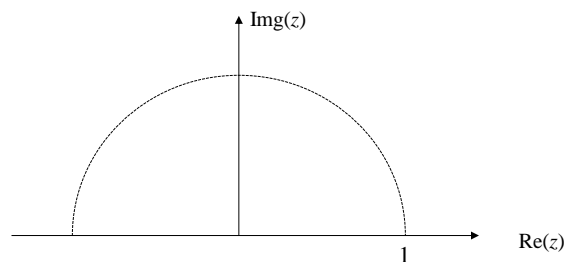
$$x_1[n] * x_2[n] \leftrightarrow X_1(z)X_2(z) \quad R' \supset R_1 \cap R_2$$



The z -plane

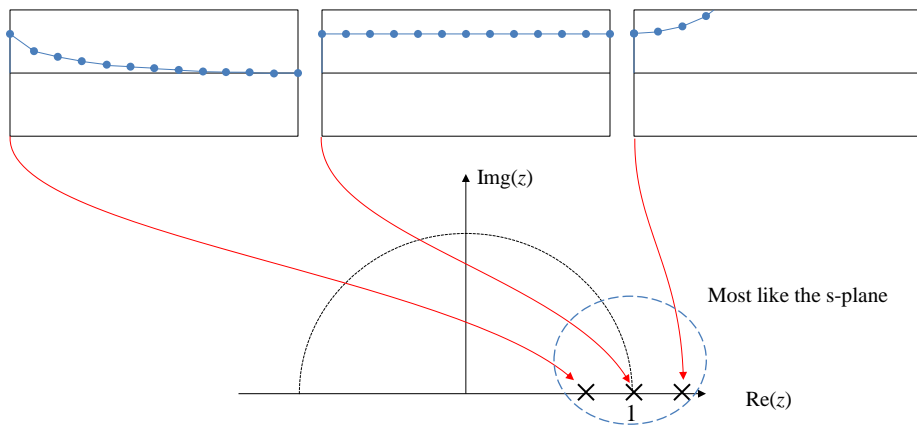
- We can understand system response by pole location in the z -plane

[Adapted from Franklin, Powell and Emami-Naeini]



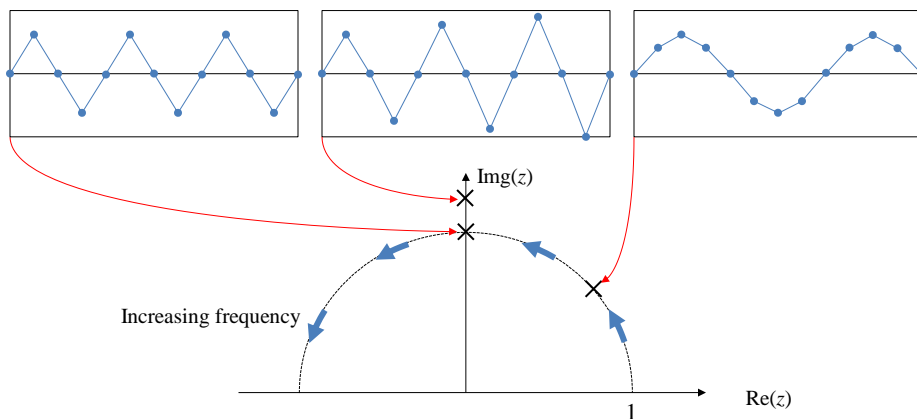
Effect of pole positions

- We can understand system response by pole location in the z -plane



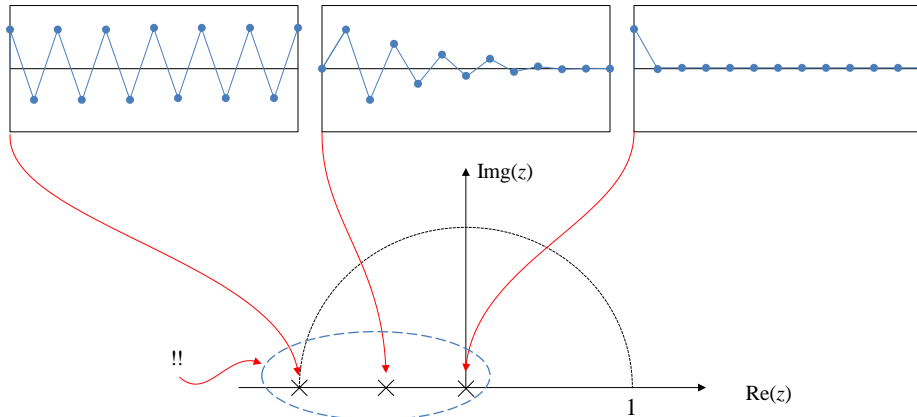
Effect of pole positions

- We can understand system response by pole location in the z -plane



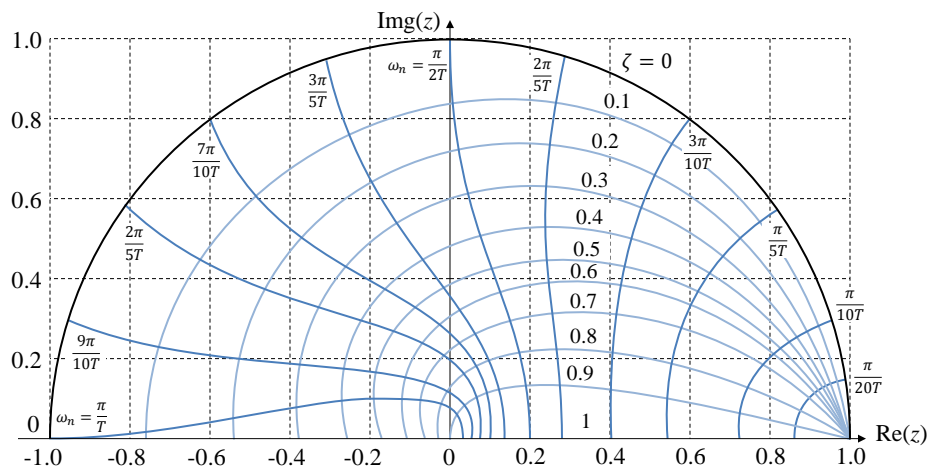
Effect of pole positions

- We can understand system response by pole location in the z -plane



Damping and natural frequency

$$z = e^{sT} \text{ where } s = -\zeta\omega_n \pm j\omega_n\sqrt{1-\zeta^2}$$



[Adapted from Franklin, Powell and Emami-Naeini]

