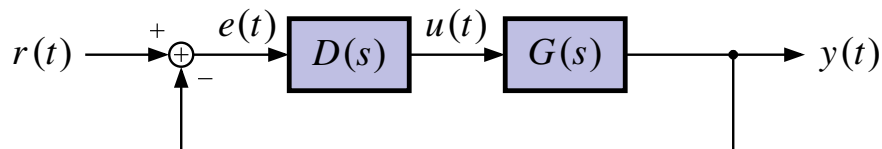


EMULATION OF ANALOG CONTROLLERS

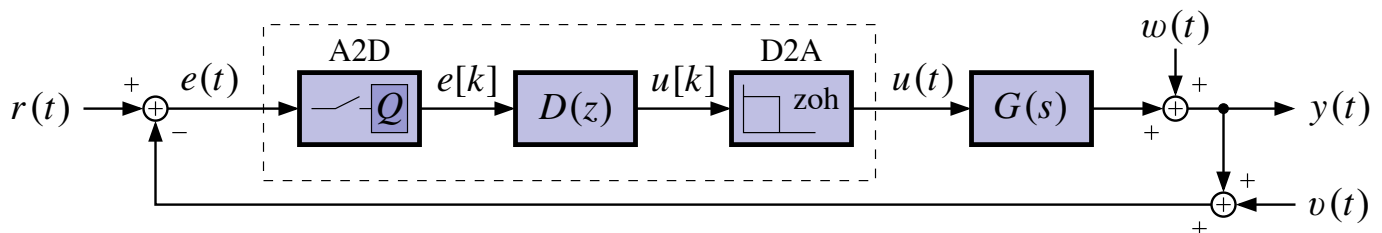
2.1: Control design via time-domain emulation

- There are two main approaches to digital controller design:
 1. Emulation—we look at this now (two methods).
 2. Direct digital design—subject of the rest of the course.
- Emulation is when a digital computer approximates an analog controller design.

■ Analog:

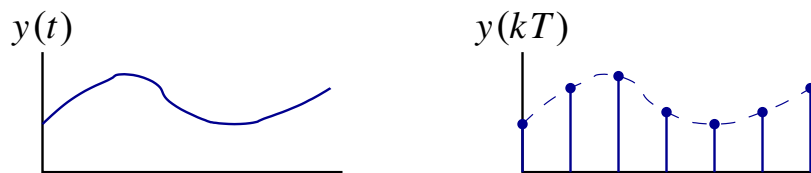


■ Digital:

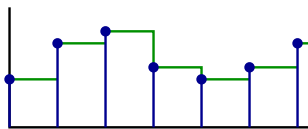


- Analog controller computes $u(t)$ from $e(t)$ using differential equations.
- Digital controller computes $u(kT)$ from $e(kT)$ using difference equations.
- To interface the computer controller to the “real world” we need an analog-to-digital converter (to measure analog signals) and digital-to-analog converter (to output signals).

- Sampling and outputting usually done synchronously, at a constant rate. If sampling period = T , frequency = $1/T$.
- The signals inside the computer (the sampled signals) are noted as $y(kT)$, or simply $y[k]$. $y[k]$ is a discrete-time signal, where $y(t)$ is a continuous-time signal.



- Discrete-time signals are usually converted to continuous-time signals using a zero-order hold:



e.g., to convert $u[k]$ to $u(t)$.

- We will spend more time on these topics as the course progresses. Only a conceptual understanding is needed now.

“Digitization”

- Continuous-time controllers are designed with Laplace-transform techniques. The resulting controller is a function of “ s ”.

$$x(t) \longrightarrow \boxed{s} \longrightarrow y(t) = \frac{dx(t)}{dt}$$

- So, “ s ” is a derivative operator. There are several ways of approximating this in discrete time. We look at one now called the “forward rectangular” rule.

$$\dot{x}(t) \triangleq \lim_{\delta t \rightarrow 0} \frac{\delta x(t)}{\delta t} = \lim_{\delta t \rightarrow 0} \frac{x(t + \delta t) - x(t)}{\delta t}.$$

- If T is small,

$$\dot{x}(kT) \approx \frac{x((k+1)T) - x(kT)}{T} \quad i.e., \quad \dot{x}[k] \approx \frac{x[k+1] - x[k]}{T}.$$

- To use this approximation, we set $T = t_{k+1} - t_k =$ sampling interval.
- We “digitize” a controller $D(s)$ by

1. Noting that controller output is related to controller input via $U(s) = D(s)E(s)$.

2. Performing term-by-term inverse Laplace transform to get a differential equation relationship between $u(t)$ and $e(t)$

$$\sum_{k=0}^n a_k \frac{d^k u(t)}{dt^k} = \sum_{k=0}^m b_k \frac{d^k e(t)}{dt^k}.$$

3. Replace derivatives with differences.

EXAMPLE:

- Consider $D(s) = \frac{U(s)}{E(s)} = k_0 \frac{s+a}{s+b}$... a lead or lag controller.

1. $(s+b)U(s) = k_0(s+a)E(s)$.

2. $\dot{u}(t) + bu(t) = k_0\dot{e}(t) + ak_0e(t)$.

3. Use “forward rectangular rule” to digitize

$$\frac{u[k+1] - u[k]}{T} + bu[k] = k_0 \left(\frac{e[k+1] - e[k]}{T} + ae[k] \right)$$

$$u[k+1] = u[k] +$$

$$T \left[-bu[k] + k_0 \left(\frac{e[k+1] - e[k]}{T} + ae[k] \right) \right]$$

$$u[k+1] = (1 - bT)u[k] + k_0(aT - 1)e[k] + k_0e[k+1].$$

- Note that we often re-index the difference equation to be in more familiar terms of “ k ” instead of “ $k+1$ ”

$$u[k] = (1 - bT)u[k-1] + k_0(aT - 1)e[k-1] + k_0e[k].$$

- Present output of digital controller $u[k]$ depends on previous output $u[k - 1]$ as well as the previous and current errors $e[k - 1]$ and $e[k]$.

Real-Time Controller Implementation

$x = 0$. (initialization of “past” values for first loop through)

Define constants:

$$\alpha_1 = 1 - bT.$$

$$\alpha_2 = k_0(aT - 1).$$

READ A/D to obtain $y[k]$ and $r[k]$.

$$e[k] = r[k] - y[k].$$

$$u[k] = x + k_0e[k].$$

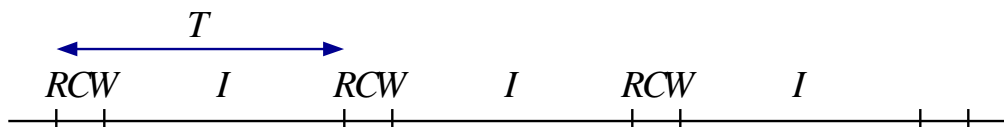
OUTPUT $u[k]$ to D/A and ZOH.

Now compute x for the next loop through:

$$x = \alpha_1u[k] + \alpha_2e[k].$$

Go back to “READ” when T seconds have elapsed since last READ.

- Code is optimized to minimize latency between A2D read and D2A write.



$R = \text{read.}$ $W = \text{write.}$
 $C = \text{compute.}$ $I = \text{idle.}$

- Rule of thumb: Sampling frequency must be ≈ 30 times the bandwidth of the analog system for comparable performance.

EXAMPLE: This one with numbers:

- Let $D(s) = 70 \frac{(s + 2)}{(s + 10)}$, $G(s) = \frac{1}{s(s + 1)}$.

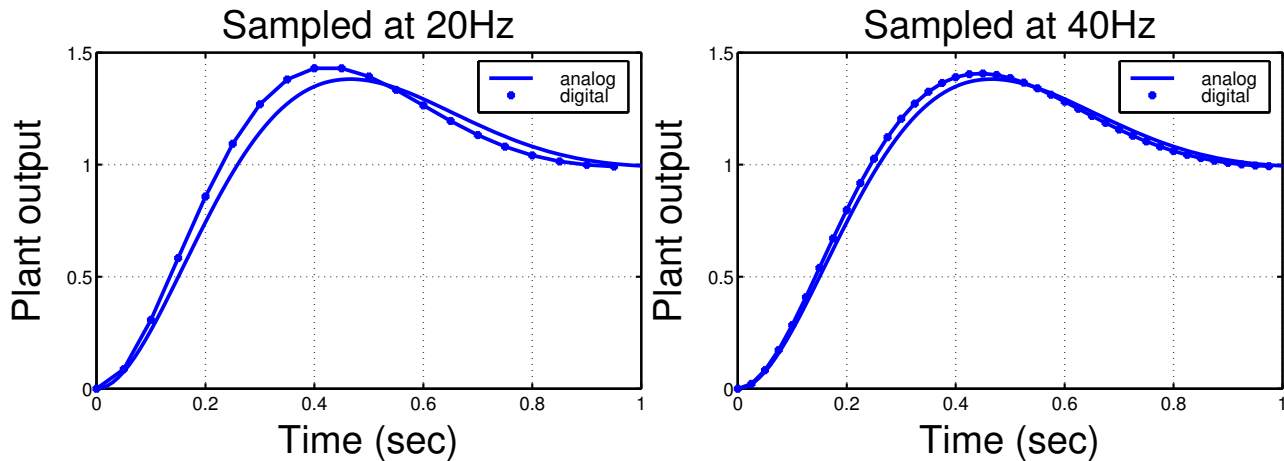
- Choose to try a sample rate of 20 Hz and also try 40 Hz.

(Note, BW of analog system is ≈ 1 Hz or so).

⇒ Use formula from before to digitize $D(s)$.

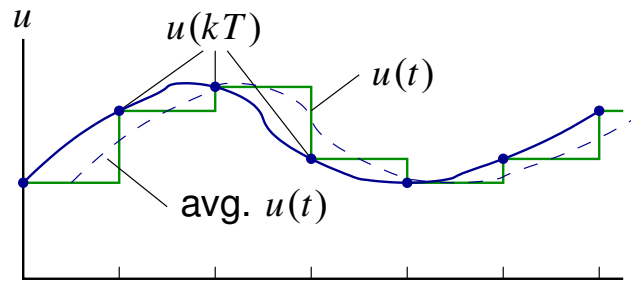
■ 20 Hz: $u[k + 1] = 0.5u[k] + 70e[k + 1] - 63e[k]$.

■ 40 Hz: $u[k + 1] = 0.75u[k] + 70e[k + 1] - 66.5e[k]$.



- **IMPORTANT NOTE:** The closed loop system with digital controller has poorer damping than the original analog system. This will *always* be true when emulating an analog controller. We see why next . . .

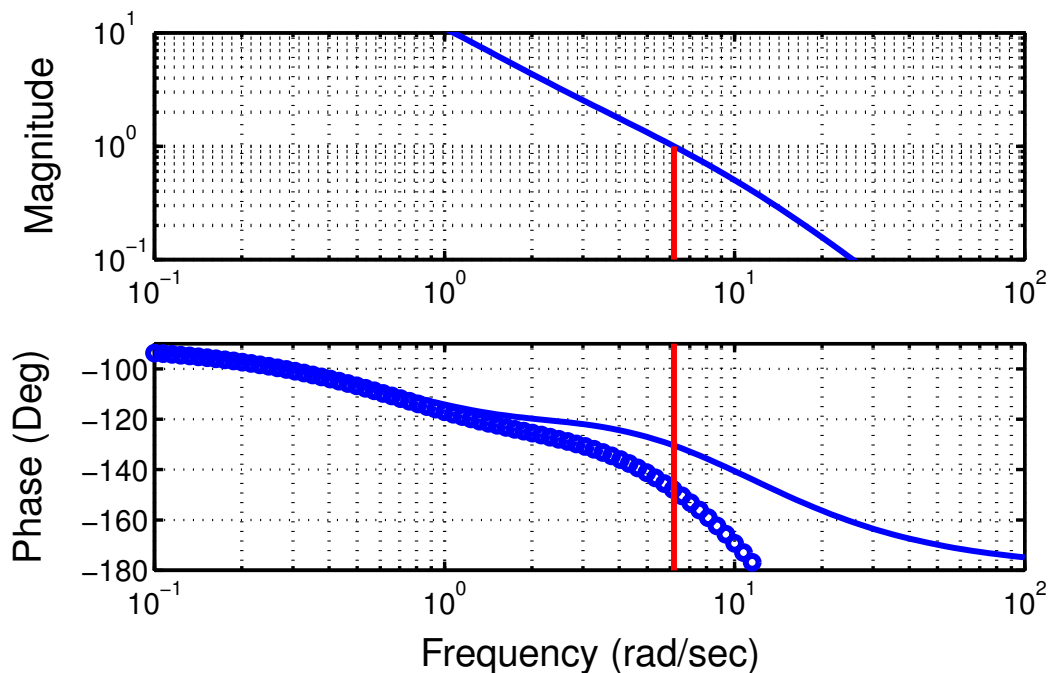
2.2: The importance of the D2A “hold” operation



- Even if $u(kT)$ is a perfect re-creation of the output of the analog controller at $t = kT$, the “hold” in the D2A causes an “effective delay.”
- The delay is approximately equal to half of the sampling period: $T/2$.
- Recall from frequency-response analysis and design, the magnitude of a delayed response stays the same, but the phase changes:

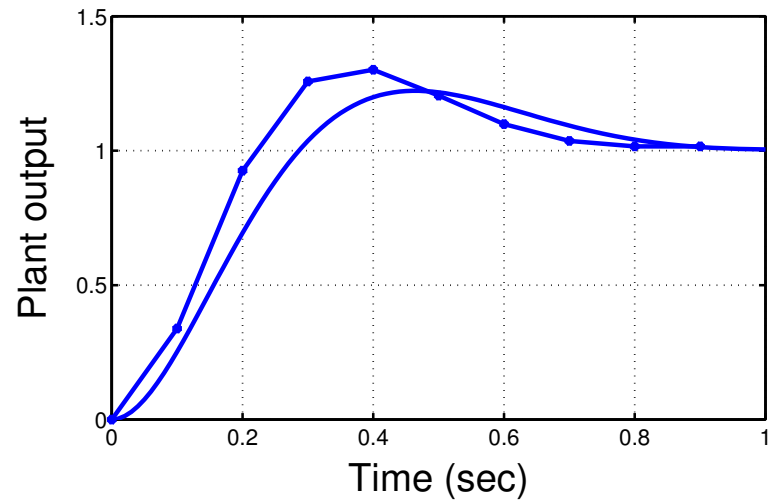
$$\Delta \text{phase} = -\omega \frac{T}{2}.$$

- For the previous example, sampling now at 10 Hz, we have:



- The PM has changed from $\approx 50^\circ$ to $\approx 30^\circ$.
- $\zeta \approx \frac{PM}{100}$... ζ changed from 0.5 to 0.3 ... much less damping.

- M_p from about 20% to about 30% ...
- Faster sampling... smaller T ... smaller delay... smaller change in response.

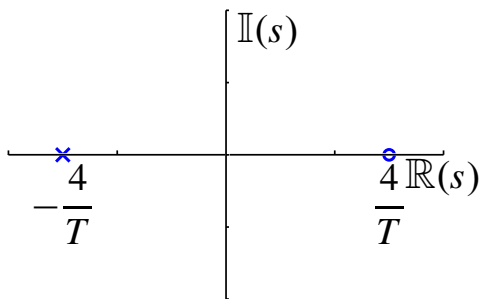


Root-locus view of the delay

- Recall that we can model a delay using a Padé approximation.

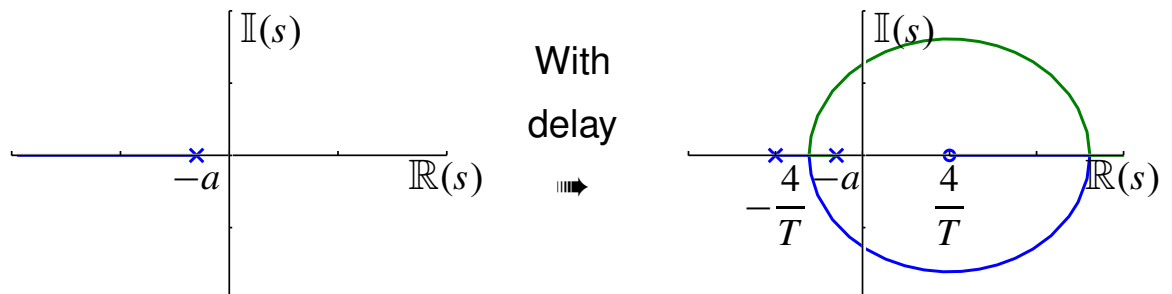
$$\frac{T}{2} \text{ delay} \longrightarrow e^{-sT/2} \approx \frac{1 - sT/4}{1 + sT/4}$$

- Poles and zeros reflected about the $j\omega$ -axis.



As $T \rightarrow 0$, delay dynamics $\rightarrow \infty$.

- Impact of delay: Suppose $D(s)G(s) = \frac{1}{s+a}$.



- Does the delayed locus make sense?

$$\frac{1 - sT/4}{1 + sT/4} = -\frac{(sT/4 - 1)}{(sT/4 + 1)}.$$

- Gain is negative! We need to draw a 0° root locus, not the 180° locus we are more familiar with.
- Conclusion: Delay destabilizes the system.

PID Control via Emulation

$$\left. \begin{array}{l} \text{P: } u(t) = Ke(t) \\ \text{I: } u(t) = \int_0^t \frac{K}{T_I} e(\tau) d\tau \\ \text{D: } u(t) = KT_D \dot{e}(t) \end{array} \right\} \text{PID: } u(t) = K \left[e(t) + \frac{1}{T_I} \int_0^t e(\tau) d\tau + T_D \dot{e}(t) \right]$$

$$\text{or, } \dot{u}(t) = K \left[\dot{e}(t) + \frac{1}{T_I} e(t) + T_D \ddot{e}(t) \right].$$

- Convert to discrete-time (use rule twice for $\ddot{e}(t)$).

$$u[k] = u[k-1] + K \left[\left(1 + \frac{T}{T_I} + \frac{T_D}{T} \right) e[k] - \left(1 + \frac{2T_D}{T} \right) e[k-1] + \frac{T_D}{T} e[k-2] \right].$$

EXAMPLE:

$$G(s) = \frac{360000}{(s+60)(s+600)} \quad K = 5, T_D = 0.0008, T_I = 0.003.$$

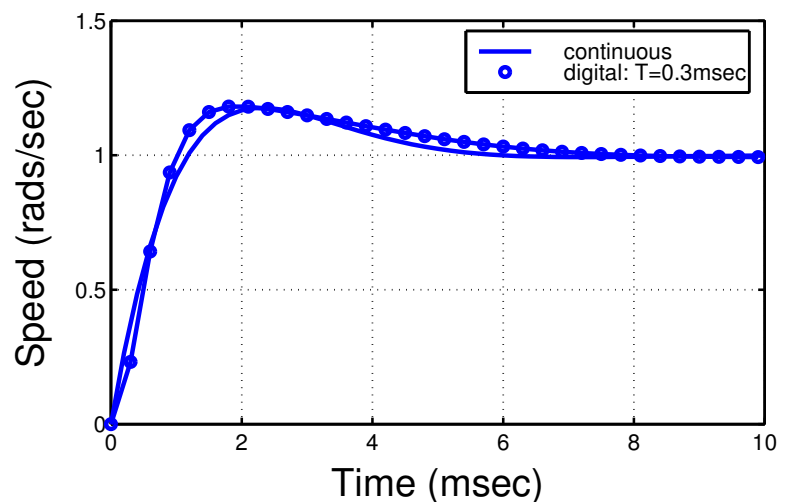
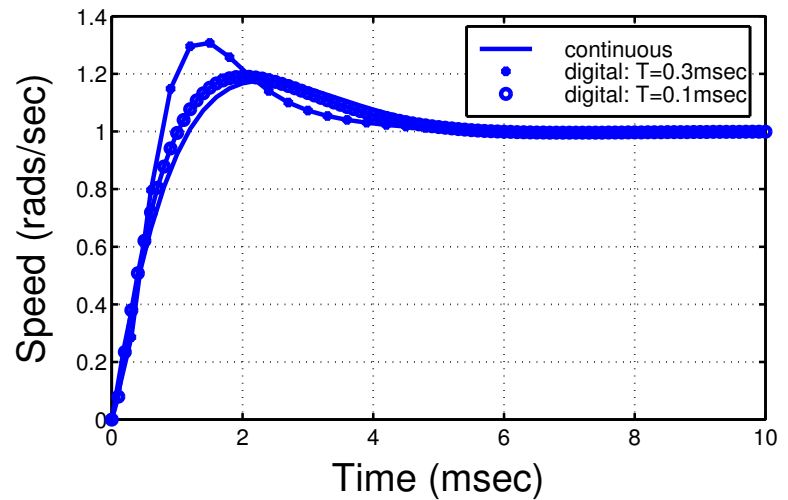
- Bode plot of cts-time OL system $D(s)G(s)$ with the above PID $D(s)$ shows that $\text{BW} \approx 1800 \text{ rad s}^{-1} \approx 320 \text{ Hz}$.

$$10 \times \text{BW} \quad \Rightarrow \quad T = 0.3 \text{ ms.}$$

- From above,

$$u[k] = u[k-1] + 5 \left[3.7667e[k] - 6.333e[k-1] + 2.6667e[k-2] \right].$$

- Performance not great, so tried again with $T = 0.1$ ms. Much better.
- Note, however, that the error is mostly due to the rise time being too fast, and the damping too low.
- *FIDDLE* with parameters \Rightarrow increase K to slow the system down; Increase T_D to increase damping. \Rightarrow New $K = 3.2$, new $T_D = 0.3$ ms.



KEY POINT: We can emulate a desired analog response using the forward-rectangular rule, but the delay added to the system due to the D2A hold circuit will decrease damping. This could even destabilize the system!!! This delay can be minimized by sampling at a high rate. (EXPENSIVE). Or, we can change the digital controller parameters, as in the last example, to achieve the desired system performance BUT NOT BY emulating the specific analog controller $D(s)$.

- How do we design the digital controller if we cannot use $D(s)$ as a prototype? Need Laplace-like tools for discrete-time.

THE z -TRANSFORM!

2.3: Definition of the z -transform

- In ECE4510/5510, we saw that the Laplace transform is a very powerful tool for analysis and design of analog control systems.
- We are now looking at discrete-time (digital) control systems, and we need a similar tool. Enter the z -transform. . .

DEFINITION: The (one-sided or unilateral) z -transform is defined as

$$\begin{aligned} X(z) &= x(0) + x(T)z^{-1} + x(2T)z^{-2} + x(3T)z^{-3} + \dots \\ &= \sum_{k=0}^{\infty} x(kT)z^{-k} \\ \left(\text{or, } X(z) &= \sum_{k=0}^{\infty} x[k]z^{-k}. \right) \end{aligned}$$

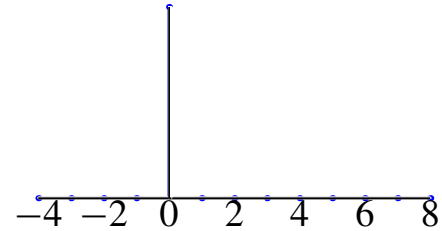
- Like the Laplace-transform variable “ s ,” the z -transform variable “ z ” is a complex number.
- Like the Laplace-transform, convergence of the transform is an important consideration.
 - For the above definition of the z -transform, convergence is always of the form

$$X(z) = \text{something}, \quad |z| > \rho,$$

where ρ is a disc on the z -plane.

EXAMPLE: Define the digital impulse (unit pulse) function to be

$$\delta[k] = \begin{cases} 1, & k = 0; \\ 0, & \text{otherwise.} \end{cases}$$



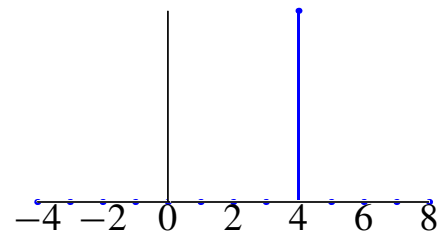
- Note: This is very different from an analog impulse (e.g., $\delta[0]$ is defined), but plays a similar role.

so, let $x[k] = \delta[k]$

$$\begin{aligned} X(z) &= \sum_{k=0}^{\infty} x[k]z^{-k} \\ &= x[0]z^0 + x[1]z^{-1} + x[2]z^{-2} + \dots \\ &= x[0] = 1. \end{aligned}$$

- Since this sum converges to 1 regardless of the value of z ,
ROC = $\{|z| > 0\}$.

EXAMPLE: Consider a delayed impulse:



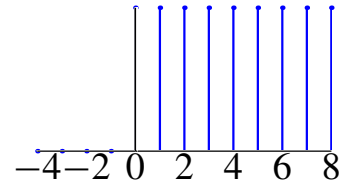
$$\begin{aligned} x[k] &= \delta[k - k_d] \\ X(z) &= \sum_{k=0}^{\infty} x[k]z^{-k} \\ &= x[0] + x[1]z^{-1} + \dots + x[k_d]z^{-k_d} + \dots \\ &= x[k_d]z^{-k_d} = z^{-k_d}. \end{aligned}$$

- Again, ROC = $\{|z| > 0\}$.

EXAMPLE: Let $x[k]$ be the discrete unit-step function.

$$x[k] = 1[k] = \begin{cases} 1, & k \geq 0; \\ 0, & \text{otherwise.} \end{cases}$$

$$\begin{aligned} 1(z) &= \sum_{k=0}^{\infty} 1[k]z^{-k} \\ &= \sum_{k=0}^{\infty} z^{-k} = 1 + z^{-1} + z^{-2} + \dots \end{aligned}$$



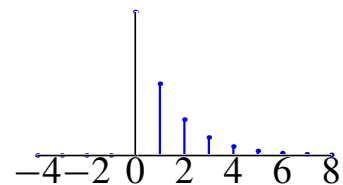
- Multiply both sides of this equation by $(z - 1)$.

$$\begin{aligned} (z - 1)1(z) &= (z + 1 + z^{-1} + z^{-2} + \dots) - (1 + z^{-1} + z^{-2} + \dots) \\ &= z - z^{-\infty} \quad (\text{in the limit}) \\ &= z \quad \text{if } |z| > 1. \end{aligned}$$

$$\text{so } 1(z) = \frac{z}{z - 1}, \quad \text{ROC} = \{|z| > 1\}.$$

EXAMPLE: Let $x[k] = a^k 1[k]$ where a is a complex number.

$$\begin{aligned} X(z) &= \sum_{k=0}^{\infty} a^k z^{-k} \\ &= 1 + az^{-1} + a^2 z^{-2} + \dots \end{aligned}$$



- Multiply both sides of this equation by $(z - a)$.

$$\begin{aligned} (z - a)X(z) &= (z + a + a^2 z^{-1} + \dots) - (a + a^2 z^{-1} + a^3 z^{-2} + \dots) \\ &= z - a^{\infty} z^{-\infty} \quad (\text{in the limit}) \\ &= z \quad \text{if } |z| > |a| \end{aligned}$$

$$\text{so } X(z) = \frac{z}{z - a}, \quad \text{ROC} = \{|z| > |a|\}.$$

2.4: Delay properties of the z -transform

Linearity:

- The standard linearity rule:
 - If $x[k] \leftrightarrow X(z)$ and $v[k] \leftrightarrow V(z)$,
 - Then $ax[k] + bv[k] \leftrightarrow aX(z) + bV(z)$. (prove for yourself)

EXAMPLE: Let $y[k] = 1[k] + a^k 1[k]$.

- Then,

$$Y(z) = \frac{z}{z-1} + \frac{z}{z-a} = \frac{2z^2 - (1+a)z}{(z-1)(z-a)}.$$

Right shift in time (delay); Theorem 1:

- Suppose $x[k] \leftrightarrow X(z)$, and q is a positive integer.

$$\begin{aligned} y[k] &= x[k-q]1[k-q] \\ Y(z) &= \sum_{k=0}^{\infty} x[k-q]1[k-q]z^{-k} \quad \text{let } \bar{k} = k - q \\ &= \sum_{\bar{k}=-q}^{\infty} x[\bar{k}]1[\bar{k}]z^{-(\bar{k}+q)} \\ &= z^{-q} \sum_{\bar{k}=0}^{\infty} x[\bar{k}]z^{-\bar{k}} \\ &= z^{-q} X(z). \end{aligned}$$

- So, a delay of q samples multiplies the z -transform by z^{-q} .

EXAMPLE:

$$p[k] = \begin{cases} 1, & k = 0, 1, \dots, N-1; \\ 0, & \text{otherwise.} \end{cases}$$

- Note that $p[k] = 1[k] - 1[k - N]$.
- By linearity and by the right-shift theorem,

$$\begin{aligned} P(z) &= 1(z) - z^{-N}1(z) \\ &= \frac{z}{z-1} - z^{-N} \frac{z}{z-1} \\ &= \frac{z^N - 1}{z^{N-1}(z-1)}. \end{aligned}$$

Right shift in time (delay); Theorem 2:

- The first delay theorem states: $x[k - q]1[k - q] \iff z^{-q}X(z)$.
- The second delay theorem states: $x[k - q] \iff \dots$
- The first version is generally used on *causal* signals. The second version is generally used on *non-causal* signals or when analyzing effect of *systems* that may have initial conditions.

$$x[k - 1] \iff z^{-1}X(z) + x[-1]$$

$$x[k - 2] \iff z^{-2}X(z) + x[-2] + z^{-1}x[-1]$$

⋮

$$x[k - q] \iff z^{-q}X(z) + x[-q] + z^{-1}x[-q + 1] + \dots + z^{-q+1}x[-1].$$

- Proof for the first case:

$$\begin{aligned} x[k - 1] &\iff \sum_{k=0}^{\infty} x[k - 1]z^{-k} && \text{let } \bar{k} = k - 1 \\ &= \sum_{\bar{k}=-1}^{\infty} x[\bar{k}]z^{-(\bar{k}+1)} \end{aligned}$$

$$\begin{aligned}
 &= x[-1] + \sum_{\bar{k}=0}^{\infty} x[\bar{k}]z^{-(\bar{k}+1)} \\
 &= x[-1] + z^{-1}X(z). \quad \text{Proof of other cases by induction.}
 \end{aligned}$$

EXAMPLE: z -transform of LCCDE.

- Many discrete-time systems are described by Linear Constant Coefficient Difference Equations (LCCDEs). These are of the form

$$\sum_{i=0}^n a_i y[k-i] = \sum_{i=0}^m b_i x[k-i],$$

where $n > m$.

- We can find the transfer function of this system by taking the z -transform of both sides of this equation.
- System transfer functions always assume zero initial conditions, so we can use Delay Theorem 1.

$$\sum_{i=0}^n a_i z^{-i} Y(z) = \sum_{i=0}^m b_i z^{-i} X(z),$$

$$\begin{aligned}
 H(z) &= \frac{Y(z)}{X(z)} \\
 &= \frac{\sum_{i=0}^m b_i z^{-i}}{\sum_{i=0}^n a_i z^{-i}}.
 \end{aligned}$$

- We can multiply numerator and denominator by z^n to get it in terms of positive powers of z .

EXAMPLE: Find the output of a system described by difference equation

$$y[k] + 0.5y[k-1] = x[k],$$

where $x[k] = 1[k]$ and $y[-1] = 6$.

- There are initial conditions, so we use Delay Theorem 2,

$$Y(z) + 0.5 [z^{-1}Y(z) + y[-1]] = X(z)$$

$$Y(z)[1 + 0.5z^{-1}] = \frac{z}{z-1} - 3$$

$$Y(z) = \frac{z - 3(z-1)}{(z-1)(1 + 0.5z^{-1})}$$

$$= \frac{3z - 2z^2}{(z-1)(z+0.5)}$$

$$= \frac{-(8/3)z}{z+0.5} + \frac{(2/3)z}{z-1}$$

$$y[k] = \frac{2}{3} \cdot 1[k] - \frac{8}{3}(-0.5)^k 1[k].$$

2.5: Time multiplication and the z -transform

Multiplication by k and k^2 .

- What happens when we multiply a time sequence $x[k]$ by k or k^2 ?

- Recall:

$$X(z) = \sum_{k=0}^{\infty} x[k]z^{-k}.$$

- Then,

$$\begin{aligned} \frac{d}{dz}X(z) &= \sum_{k=0}^{\infty} -kx[k]z^{-k-1} \\ &= -z^{-1} \underbrace{\sum_{k=0}^{\infty} kx[k]z^{-k}}_{Y(z)}. \end{aligned}$$

- Thus,

$$kx[k] \iff -z \frac{d}{dz}X(z).$$

- Also

$$k^2x[k] \iff z \frac{d}{dz}X(z) + z^2 \frac{d^2}{dz^2}X(z).$$

EXAMPLE: Recall that $a^k 1[k] \leftrightarrow \frac{z}{z-a}$.

$$\frac{d}{dz} \left(\frac{z}{z-a} \right) = \left[\frac{-z}{(z-a)^2} + \frac{1}{z-a} \right] = \frac{-a}{(z-a)^2}$$

$$\text{so, } ka^k 1[k] \leftrightarrow \frac{az}{(z-a)^2}$$

$$\text{and, } k1[k] \leftrightarrow \frac{z}{(z-1)^2}. \quad (\text{ramp})$$

Multiplication by a^k :

- Multiply a time sequence $x[k]$ by a^k (a complex).

$$\begin{aligned} a^k x[k] &\leftrightarrow \sum_{k=0}^{\infty} a^k x[k] z^{-k} \\ &= \sum_{k=0}^{\infty} x[k] \left(\frac{z}{a}\right)^{-k} \\ &= X(z/a). \end{aligned}$$

Multiplication by $\cos[\omega k]$ and $\sin[\omega k]$.

- Multiply a time sequence $x[k]$ by $\cos[\omega k]$ or $\sin[\omega k]$.
- Recall from Euler's theorem

$$\cos[\omega k] = \frac{e^{j\omega k} + e^{-j\omega k}}{2} \quad \text{and} \quad \sin[\omega k] = \frac{e^{j\omega k} - e^{-j\omega k}}{2j}.$$

- So,

$$\begin{aligned} \cos[\omega k]x[k] &\iff \frac{1}{2}X(e^{-j\omega}z) + \frac{1}{2}X(e^{j\omega}z) \\ \sin[\omega k]x[k] &\iff \frac{1}{2j}X(e^{-j\omega}z) - \frac{1}{2j}X(e^{j\omega}z). \end{aligned}$$

EXAMPLE: Find the z -transform of $x[k] = \cos[\omega k]1[k]$.

- Recall that $1[k] \iff \frac{z}{z-1}$.

$$\begin{aligned} X(z) &= \frac{1}{2} \left(\frac{e^{j\omega}z}{e^{j\omega}z-1} + \frac{e^{-j\omega}z}{e^{-j\omega}z-1} \right) \\ &= \frac{1}{2} \left(\frac{e^{j\omega}z(e^{-j\omega}z-1) + e^{-j\omega}z(e^{j\omega}z-1)}{(e^{j\omega}z-1)(e^{-j\omega}z-1)} \right) \end{aligned}$$

$$= \frac{1}{2} \left(\frac{z^2 - e^{j\omega}z + z^2 - e^{-j\omega}z}{z^2 - (e^{j\omega} + e^{-j\omega})z + 1} \right)$$
$$\cos[\omega k]1[k] \iff \frac{z^2 - \cos[\omega]z}{z^2 - (2 \cos[\omega])z + 1}.$$

■ Also,

$$\sin[\omega k]1[k] \iff \frac{z \sin[\omega]}{z^2 - (2 \cos[\omega])z + 1}.$$

2.6: Convolution, initial/final value and the z -transform

Convolution

- Discrete-time convolution is analogous to continuous-time convolution:

$$(x * v)[k] = \sum_{i=0}^k x[i]v[k-i] \quad k \geq 0$$

$$\text{or,} \quad = \sum_{i=0}^k v[i]x[k-i]. \quad k \geq 0.$$

- Assuming that $x[k]$ and $v[k]$ are zero for negative k ,

$$(x * v)[k] = \sum_{i=0}^{\infty} x[i]v[k-i] \quad k \geq 0$$

$$\iff \sum_{k=0}^{\infty} \left[\sum_{i=0}^{\infty} x[i]v[k-i] \right] z^{-k}$$

$$= \sum_{i=0}^{\infty} \sum_{k=0}^{\infty} x[i]v[k-i]z^{-k}$$

$$= \sum_{i=0}^{\infty} x[i] \left[\sum_{k=0}^{\infty} v[k-i]z^{-k} \right] \quad \bar{k} = k - i$$

$$= \sum_{i=0}^{\infty} x[i] \left[\sum_{\bar{k}=-i}^{\infty} v[\bar{k}]z^{-(\bar{k}+i)} \right]$$

$$= \sum_{i=0}^{\infty} x[i] \left[\sum_{\bar{k}=0}^{\infty} v[\bar{k}]z^{-\bar{k}} \right] z^{-i}$$

$$= X(z)V(z).$$

Initial-value theorem

■ If $x[k] \iff X(z)$, $x[0] = \lim_{z \rightarrow \infty} X(z)$.

■ Proof:

$$X(z) = \sum_{k=0}^{\infty} x[k]z^{-k}$$

$$= x[0] + x[1]z^{-1} + x[2]z^{-2} + \dots$$

as $z \rightarrow \infty$,
$$= x[0] + x[1] \cdot 0 + x[2] \cdot 0 + \dots$$

$$= x[0].$$

■ Also,

$$x[1] = \lim_{z \rightarrow \infty} (zX(z) - zx[0])$$

$$x[2] = \lim_{z \rightarrow \infty} (z^2X(z) - z^2x[0] - zx[1])$$

⋮

Final value theorem

■ Proof not too enlightening:

$$\lim_{k \rightarrow \infty} x[k] = \lim_{z \rightarrow 1} (z - 1)X(z)$$

assuming the limit on the left exists! Conditions for existence: All poles of $X(z)$ must be inside unit circle $|p_i| < 1 \forall i$, except possibly a single pole on the unit circle at $z = 1$. The residue of the pole at $z = 1$ is the final value and is found from above.

EXAMPLE:

■ Suppose $X(z) = \frac{3z^2 - 2z + 4}{z^3 - 2z^2 + 1.5z - 0.5}$.

■ Factoring the denominator:

$$X(z) = \frac{3z^2 - 2z + 4}{(z - 1)(z^2 - z + 0.5)}$$

$$\begin{aligned}x_{ss} &= \left. \frac{3z^2 - 2z + 4}{(z^2 - z + 0.5)} \right|_{z=1} \\ &= \frac{5}{0.5} = 10.\end{aligned}$$

2.7: The inverse z -transform

1.] Power-series method:

- Suppose $X(z) = \frac{z^2 - 1}{z^3 + 2z + 4}$,
- Dividing:

$$\begin{array}{r}
 z^{-1} + 0z^{-2} - 3z^{-3} - 4z^{-4} \dots \\
 \hline
 z^3 + 2z + 4 \) \ z^2 - 1 \\
 \underline{z^2 + 2 + 4z^{-1}} \\
 -3 - 4z^{-1} \\
 \underline{-3 - 6z^{-2} - 12z^{-3}} \\
 -4z^{-1} + 6z^{-2} + 12z^{-3} \\
 \underline{-4z^{-1} - 8z^{-3} - 16z^{-4}} \\
 6z^{-2} + 20z^{-3} + 16z^{-4} \\
 \vdots
 \end{array}$$

- So, $X(z) = z^{-1} - 3z^{-3} - 4z^{-4} \dots$.
- and $x[0] = 0, x[1] = 1, x[2] = 0, x[3] = -3, x[4] = -4, \dots$

2.] Partial-fraction expansion

- Recall that partial-fraction expansion was used to take inverse-Laplace transforms of things like:

$$H(s) = \frac{\text{num}(s)}{\text{den}(s)},$$

where the order of $\text{num}(s) < \text{order of den}(s)$.

- z -transforms almost always have the same order numerator and denominator which requires modifying the technique slightly.
- Let $X(z) = \frac{N(z)}{D(z)}$ and let $D(z)$ have *distinct* poles.

- Do partial-fraction expansion on $\frac{X(z)}{z} = \frac{N(z)}{zD(z)}$.

$$\frac{X(z)}{z} = \frac{r_0}{z} + \frac{r_1}{z - p_1} + \frac{r_2}{z - p_2} + \dots + \frac{r_n}{z - p_n}$$

$$r_0 = \left[z \frac{X(z)}{z} \right]_{z=0} = X(0)$$

$$r_i = \left[(z - p_i) \frac{X(z)}{z} \right]_{z=p_i} \quad i = 1, \dots, n.$$

- So,

$$X(z) = r_0 + \frac{r_1 z}{z - p_1} + \frac{r_2 z}{z - p_2} + \dots + \frac{r_n z}{z - p_n}.$$

- Transforming: $r_0 \iff r_0 \delta[k]$.
- Transforming: $\frac{r_i z}{z - p_i} \iff r_i p_i^k 1[k]$.
- Note:

$$r_1 p_1^k + \bar{r}_1 \bar{p}_1^k = 2 |r_1| |p_1|^k \cos [k \angle p_1 + \angle r_1] \quad k = 0, 1, \dots$$

if r_1 and p_1 complex.

EXAMPLE:

- Suppose

$$\begin{aligned} X(z) &= \frac{z^3 + 1}{z^3 - z^2 - z - 2} \\ &= \frac{z^3 + 1}{(z - 2) \left(z + \frac{1}{2} + j\frac{\sqrt{3}}{2} \right) \left(z + \frac{1}{2} - j\frac{\sqrt{3}}{2} \right)} \\ \frac{X(z)}{z} &= \frac{r_0}{z} + \frac{r_1}{z + \frac{1}{2} + j\frac{\sqrt{3}}{2}} + \frac{\bar{r}_1}{z + \frac{1}{2} - j\frac{\sqrt{3}}{2}} + \frac{r_2}{z - 2}. \end{aligned}$$

- Find the residues:

$$r_0 = X(0) = \frac{-1}{2}$$

$$r_1 = \left[\frac{z^3 + 1}{(z - 2) \left(z + \frac{1}{2} - j\frac{\sqrt{3}}{2} \right) z} \right]_{z = -\frac{1}{2} - j\frac{\sqrt{3}}{2}} = \frac{3}{7} + j\frac{\sqrt{3}}{21}$$

$$r_2 = \left[(z - 2) \frac{X(z)}{z} \right]_{z=2} = \frac{9}{14}$$

- So,

$$x[k] = -\frac{1}{2}\delta[k] + r_1 \left(-\frac{1}{2} - j\frac{\sqrt{3}}{2} \right)^k 1[k] + \bar{r}_1 \left(-\frac{1}{2} + j\frac{\sqrt{3}}{2} \right)^k 1[k] + \frac{9}{14} 2^k 1[k].$$

- Note: $|p_1| = 1$, $\angle p_1 = 4\pi/3$. $|r_1| = 0.436$, $\angle r_1 = 10.89^\circ$.

- So,

$$x[k] = -\frac{1}{2}\delta[k] + 0.873 \cos \left(\frac{4\pi}{3}k + 10.89^\circ \right) 1[k] + \frac{9}{14} (2)^k 1[k].$$

Repeated poles:

- There is a special rule for repeated poles, just as with Laplace-transform partial-fraction expansion.

- Suppose p_1 is repeated “ n ” times.

$$\frac{X(z)}{z} = \frac{r_0}{z} + \frac{r_{1,1}}{z - p_1} + \frac{r_{1,2}}{(z - p_1)^2} + \dots + \frac{r_{1,n}}{(z - p_1)^n} + \text{other poles}$$

- Then, the residues are calculated as

$$r_{1,n} = \left[(z - p_1)^n \frac{X(z)}{z} \right]_{z=p_1}$$

$$r_{1,n-1} = \frac{1}{1!} \left[\frac{d}{dz} \left((z - p_1)^n \frac{X(z)}{z} \right) \right]_{z=p_1}$$

$$r_{1,n-2} = \frac{1}{2!} \left[\frac{d^2}{dz^2} \left((z - p_1)^n \frac{X(z)}{z} \right) \right]_{z=p_1}$$

$$\vdots$$

$$r_{1,1} = \frac{1}{(n-1)!} \left[\frac{d^{n-1}}{dz^{n-1}} \left((z - p_1)^n \frac{X(z)}{z} \right) \right]_{z=p_1}.$$

- MATLAB's residue command automates partial-fraction expansion:

```
num=[ ] % coefficients of numerator.
den=[] % coefficients of denominator.
[r,p,k]=residue(num,[den 0]) % don't forget '1/z'!
```

- Or, there is a somewhat easier way:

```
num=[ ] % coefficients of numerator.
den=[] % coefficients of denominator.
[r,p,k]=residuez(num,den)
```

- Be sure to type `help residue` or `help residuez` if there are repeated roots to understand the format of the results that MATLAB returns.

3.] Inversion formula:

- A third way to invert a z -transform is to use the relationship:

$$x[k] = \frac{1}{2\pi j} \oint_{\Gamma} X(z) z^{k-1} dz$$

where this is a counter-clockwise contour integral over the path Γ in the z -plane. Γ must be in the region of convergence.

- This looks terrible, but the theorem of residues makes it simpler. It reduces to:

$$x[k] = \left[\sum_{\text{all poles of } X(z)z^{k-1}} [\text{residues of } X(z)z^{k-1}] \right] 1[k].$$

- For a simple pole of $X(z)z^{k-1}$ at a , the residue is:

$$[\text{residue}]_{z=a} = (z - a)X(z)z^{k-1} \Big|_{z=a}.$$

- For a repeated pole of $X(z)z^{k-1}$ at a , repeated m times,

$$[\text{residue}]_{z=a} = \frac{1}{(m-1)!} \frac{d^{m-1}}{dz^{m-1}} [(z-a)^m X(z)z^{k-1}]_{z=a}.$$

EXAMPLE:

$$X(z) = \frac{z}{(z-1)(z-2)}.$$

- So,

$$\begin{aligned} x[k] &= \left[\frac{z^k}{z-2} \Big|_{z=1} + \frac{z^k}{z-1} \Big|_{z=2} \right] 1[k] \\ &= [-1 + 2^k] 1[k]. \end{aligned}$$

EXAMPLE:

$$X(z) = \frac{z}{(z-1)^2}.$$

$$\begin{aligned} x[k] &= \left[\frac{1}{(2-1)!} \frac{d^{2-1}}{dz^{2-1}} \left[(z-1)^2 \frac{z}{(z-1)^2} z^{k-1} \right]_{z=1} \right] 1[k] \\ &= \left[\frac{d}{dz} z^k \Big|_{z=1} \right] 1[k] \\ &= k z^{k-1} \Big|_{z=1} 1[k] \\ &= k 1[k]. \end{aligned}$$

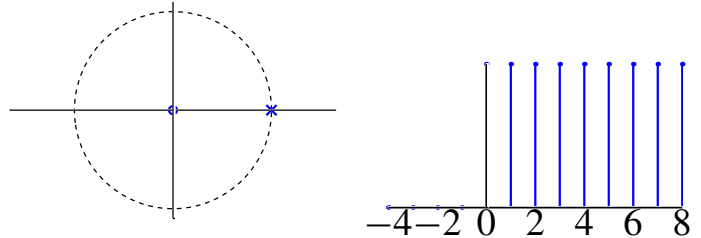
Laplace transform	Time function	z-Transform	Modified z-transform
$E(s)$	$e(t)$	$E(z)$	$E(z, m)$
$\frac{1}{s}$	$1(t)$	$\frac{z}{z-1}$	$\frac{1}{z-1}$
$\frac{1}{s^2}$	$t1(t)$	$\frac{Tz}{(z-1)^2}$	$\frac{mT}{z-1} + \frac{T}{(z-1)^2}$
$\frac{1}{s^3}$	$\frac{t^2}{2}1(t)$	$\frac{T^2z(z+1)}{2(z-1)^3}$	$\frac{T^2}{2} \left[\frac{m^2}{z-1} + \frac{2m+1}{(z-1)^2} + \frac{2}{(z-1)^3} \right]$
$\frac{(k-1)!}{s^k}$	$t^{k-1}1(t)$	$\lim_{a \rightarrow 0} (-1)^{k-1} \frac{\partial^{k-1}}{\partial a^{k-1}} \left[\frac{z}{z-e^{-aT}} \right]$	$\lim_{a \rightarrow 0} (-1)^{k-1} \frac{\partial^{k-1}}{\partial a^{k-1}} \left[\frac{e^{-amT}}{z-e^{-aT}} \right]$
$\frac{1}{s+a}$	$e^{-at}1(t)$	$\frac{z}{z-e^{-aT}}$	$\frac{e^{-amT}}{z-e^{-aT}}$
$\frac{1}{(s+a)^2}$	$te^{-at}1(t)$	$\frac{Tze^{-aT}}{(z-e^{-aT})^2}$	$\frac{T e^{-amT} [e^{-aT} + m(z-e^{-aT})]}{(z-e^{-aT})^2}$
$\frac{(k-1)!}{(s+a)^k}$	$t^k e^{-at}1(t)$	$(-1)^k \frac{\partial^k}{\partial a^k} \left[\frac{z}{z-e^{-aT}} \right]$	$(-1)^k \frac{\partial^k}{\partial a^k} \left[\frac{e^{-amT}}{z-e^{-aT}} \right]$
$\frac{a}{s(s+a)}$	$1 - e^{-at}$	$\frac{z(1-e^{-aT})}{(z-1)(z-e^{-aT})}$	$\frac{1}{z-1} - \frac{e^{-amT}}{z-e^{-aT}}$
$\frac{a}{s^2(s+a)}$	$t - \frac{1-e^{-aT}}{a}$	$z \left[\frac{(aT-1+e^{-aT})z + (a-e^{-aT})z - aT e^{-aT}}{a(z-1)^2(z-e^{-aT})} \right]$	$\frac{T}{(z-1)^2} + \frac{amT-1}{a(z-1)} + \frac{e^{-amT}}{a(z-e^{-aT})}$
$\frac{a^2}{s(s+a)^2}$	$1 - (1+at)e^{-at}$	$\frac{z}{z-1} - \frac{z}{z-e^{-aT}} - \frac{aT e^{-aT} z}{(z-e^{-aT})^2}$	$\frac{1}{z-1} - \left[\frac{1+amT}{z-e^{-aT}} + \frac{aT e^{-aT}}{(z-e^{-aT})^2} \right] e^{-amT}$
$\frac{b-a}{(s+a)(s+b)}$	$e^{-at} - e^{-bt}$	$\frac{(e^{-aT} - e^{-bT})z}{(z-e^{-aT})(z-e^{-bT})}$	$\frac{e^{-amT}}{z-e^{-aT}} - \frac{e^{-bmT}}{z-e^{-bT}}$
$\frac{a}{s^2+a^2}$	$\sin(at)$	$\frac{z \sin(aT)}{z^2-2z \cos(aT)+1}$	$\frac{z \sin(amT) + \sin((1-m)aT)}{z^2-2z \cos(aT)+1}$
$\frac{s}{s^2+a^2}$	$\cos(at)$	$\frac{z(z-\cos(aT))}{z^2-2z \cos(aT)+1}$	$\frac{z \cos(amT) - \cos((1-m)aT)}{z^2-2z \cos(aT)+1}$
$\frac{b}{(s+a)^2+b^2}$	$e^{-at} \sin(bt)$	$\frac{ze^{-aT} \sin(bT)}{z^2-2ze^{-aT} \cos(bT)+e^{-2aT}}$	$\frac{e^{-amT} [z \sin(bmT) + e^{-aT} \sin((1-m)bT)]}{z^2-2ze^{-aT} \cos(bT)+e^{-2aT}}$
$\frac{s+a}{(s+a)^2+b^2}$	$e^{-at} \cos(bt)$	$\frac{z^2-ze^{-aT} \cos(bT)}{z^2-2ze^{-aT} \cos(bT)+e^{-2aT}}$	$\frac{e^{-amT} [z \cos(bmT) + e^{-aT} \sin((1-m)bT)]}{z^2-2ze^{-aT} \cos(bT)+e^{-2aT}}$
$\frac{a^2+b^2}{s[(s+a)^2+b^2]}$	$1 - e^{-at} (\cos(bt) + \frac{a}{b} \sin(bt))$	$\frac{z(Az+B)}{(z-1)(z^2-2ze^{-aT} \cos(bT)+e^{-2aT})}$	$\frac{1}{z-1} - \dots$
$\frac{1}{s(s+a)(s+b)}$	$\frac{1}{ab} + \frac{e^{-at}}{a(a-b)} + \frac{e^{-bt}}{b(b-a)}$	$A = 1 - e^{-aT} (\cos(bT) + \frac{a}{b} \sin(bT))$ $B = e^{-2aT} + e^{-aT} (\frac{a}{b} \sin(bT) - \cos(bT))$	$\frac{e^{-amT} [z \cos(bmT) + e^{-aT} \sin((1-m)bT)]}{z^2-2ze^{-aT} \cos(bT)+e^{-2aT}} + \dots$ $\frac{a}{b} \frac{(e^{-amT} [z \sin(bmT) - e^{-aT} \sin((1-m)bT)])}{z^2-2ze^{-aT} \cos(bT)+e^{-2aT}}$

2.8: Time responses versus pole locations

- We found that the Laplace-transform pole locations determined time response. The same is true of the z -transform.

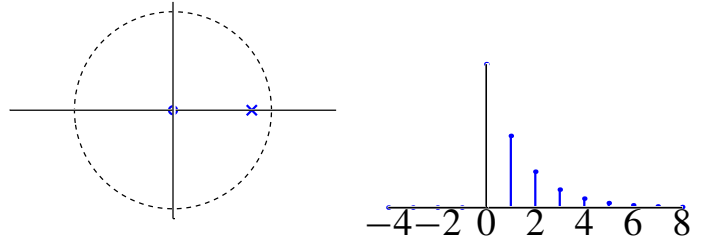
Unit step:

- $x[k] = 1[k]$.
- $X(z) = \frac{z}{z-1}, \quad |z| > 1$.



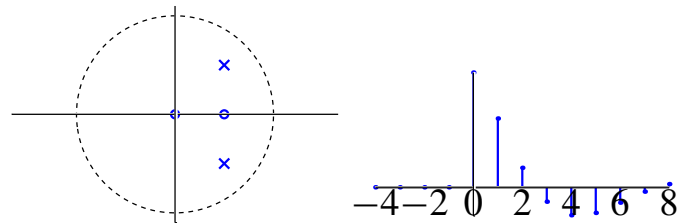
Exponential (geometric):

- $x[k] = a^k 1[k], \quad |a| < 1$.
- $X(z) = \frac{z}{z-a}, \quad |z| > |a|$.



General sinusoid:

- $x[k] = a^k \cos[\omega k] 1[k], \quad |a| < 1$.
- $X(z) = \frac{z(z - a \cos \omega)}{z^2 - 2a(\cos \omega)z + a^2}, \quad |z| > |a|$.



- The radius to the two poles is a .
- The angle to the poles is ω .
- The zero (not at the origin) has the same real part as the two poles.
- Note: If $\omega = 0$, $X(z) = \frac{z}{z-a} \dots$ exponential!
- If $\omega = 0, a = 1$, $X(z) = \frac{z}{z-1} \dots$ step!

- Radius of *pole locations* is the exponential factor, and determines settling time.

1. $|a| > 1$, Growing signal which will not decay.
2. $|a| = 1$, Signal with constant amplitude; either step or cosine.
3. $|a| < 1$, Decaying signal. Small a = fast decay (see below).

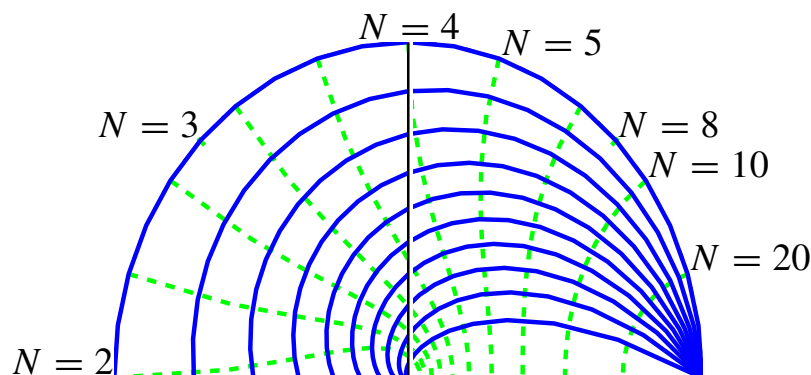
a	\approx duration N
0.9	43
0.8	21
0.6	9
0.4	5

4. $|a| = 0$, Finite-duration response. *e.g.*, $\delta[k - N] \iff z^{-N}$.

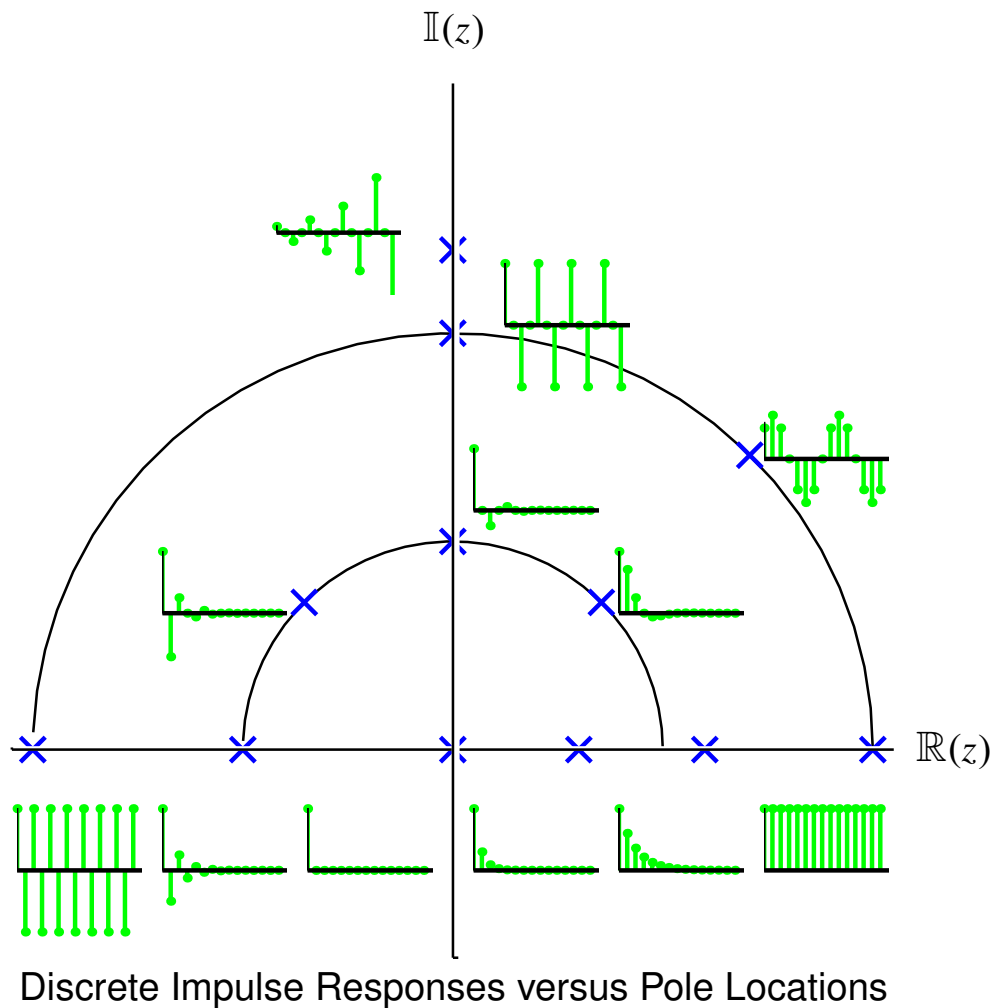
- Angle of pole locations, ω , determines number of samples per oscillation.

- That is, require $\cos[\omega k] = \cos[\omega(k + N)]$.

$$N = \frac{2\pi}{\omega} \Big|_{\text{rad}} = \frac{360}{\omega} \Big|_{\text{deg}} \quad \text{samples/cycle.}$$



- Solid lines are constant damping ratio ζ .
- Dashed lines are constant natural frequency ω_n .



Correspondence with continuous signals

■ Let $x(t) = e^{-at} \cos(bt)1(t)$.

■ Suppose

$$\left. \begin{aligned} a &= 0.3567/T \\ b &= \frac{\pi/4}{T} \end{aligned} \right\} T = \text{sampling period.}$$

■ Then,

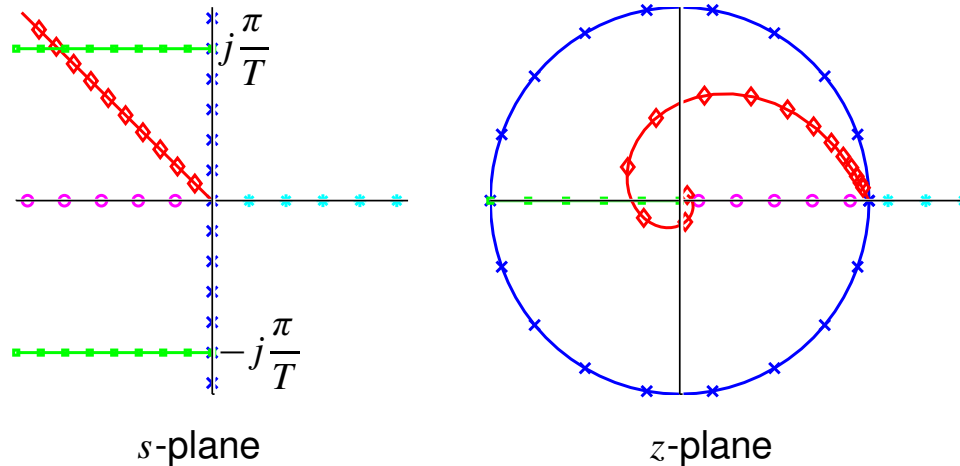
$$\begin{aligned} x[k] &= x(kT) = (e^{-0.3567})^k \cos\left(\frac{\pi k}{4}\right) 1[k] \\ &= 0.7^k \cos\left(\frac{\pi k}{4}\right) 1[k]. \end{aligned}$$

(This is the example used previously).

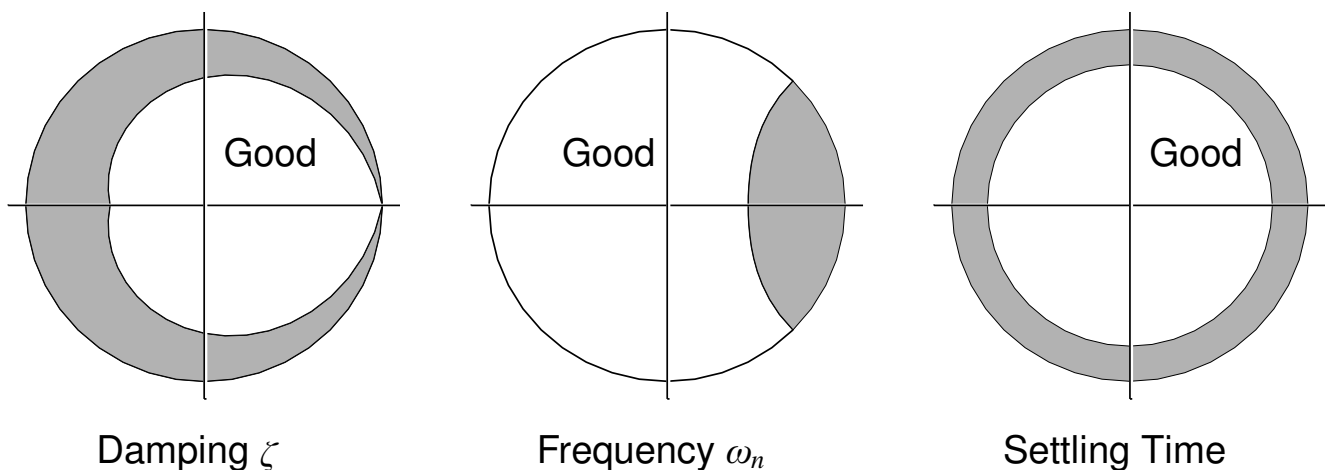
- $X(s)$ has poles at $s_{1,2} = -a + jb$ and $-a - jb$.
- $X(z)$ has poles at radius e^{-aT} angle $\omega = \pm bT$ or at $e^{-aT \pm jbT}$.
 - So, $z_{1,2} = e^{s_1 T}$ and $e^{s_2 T}$.
- In general, *poles* convert between the s -plane and z -plane via

$$z = e^{sT}.$$

EXAMPLE: Some corresponding pole locations:



- $j\omega$ -axis maps to unit circle.
- Constant damping ratio ζ maps to strange spiral.



- Higher-order systems:
 - Pole moving toward $z = 1$, system slows down.
 - Zero moving toward $z = 1$, overshoot.
 - Pole and zero moving close to each other cancel.

Where from here?

- We have now seen the most important aspects of z -transform theory.
- Our next step is to start applying this theory to digital-controller analysis and design.
- We begin by revisiting the concept of emulating an analog controller—what insight and techniques can we gain now that we understand the z transform?
- We then move on to direct digital control analysis and design.

2.9: Emulation in frequency domain

- With a further understanding of the z -transform, we are now in a better position to understand “design-by-emulation”.
- Can we design better $D(z)$ to “emulate” $D(s)$?
 - Qualified “yes”.
 - Method: Numerical integration.

EXAMPLE: Consider

$$\frac{U(s)}{E(s)} = D(s) = \frac{a}{s + a},$$

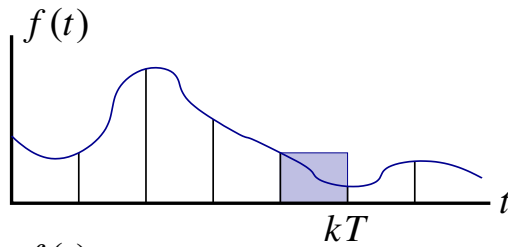
or

$$\dot{u}(t) + au(t) = ae(t).$$

- Rewrite as

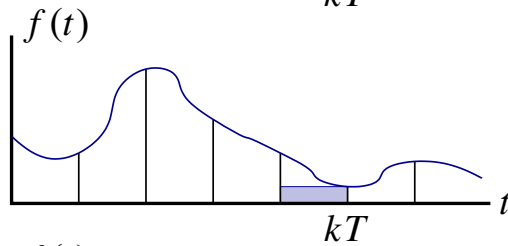
$$\begin{aligned}
 u(t) &= \int_0^t (-au(\tau) + ae(\tau)) \, d\tau \\
 u(kT) &= \int_0^{(k-1)T} (-au(\tau) + ae(\tau)) \, d\tau + \int_{(k-1)T}^{kT} (-au(\tau) + ae(\tau)) \, d\tau \\
 &= u((k-1)T) + \underbrace{\int_{(k-1)T}^{kT} (-au(\tau) + ae(\tau)) \, d\tau}_{\text{area of } -au(\tau) + ae(\tau) \text{ over } kT - T \leq \tau \leq kT}.
 \end{aligned}$$

- Can approximate this area several ways:



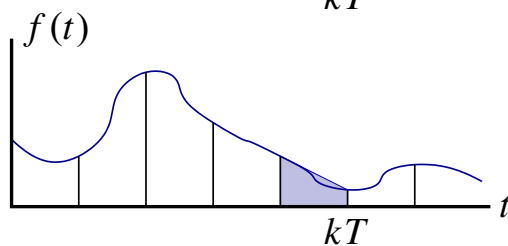
Forward Rectangular Rule

$$\text{Area} \approx f((k-1)T) \cdot T$$



Backward Rectangular Rule

$$\text{Area} \approx f(kT) \cdot T$$



Trapezoid Rule

(a.k.a., Tustin Rule or Bilinear Rule)

$$\text{Area} \approx \frac{f((k-1)T) + f(kT)}{2} \cdot T$$

Method 1: Use the forward-rectangular rule

- Using the *forward rectangular rule* on our example:

$$\begin{aligned} u(kT) &\approx u((k-1)T) + T [-au((k-1)T) + ae((k-1)T)] \\ &= (1 - aT)u((k-1)T) + aTe((k-1)T) \end{aligned}$$

- The transfer function of this is :

$$U(z) = (1 - aT)z^{-1}U(z) + aTz^{-1}E(z)$$

$$U(z)[z - 1 + aT] = aTE(z)$$

$$\frac{U(z)}{E(z)} = \frac{a}{\frac{z-1}{T} + a}$$

- “ s ” has been replaced by “ $\frac{z-1}{T}$ ”. Seen already as Euler’s rule.

Method 2: Use the backward-rectangular rule

- Using the *backward rectangular rule*, we get a different result:

$$u(kT) \approx u((k-1)T) + T[-au(kT) + ae(kT)]$$

$$u(kT)(1 + aT) = u((k-1)T) + aTe(kT)$$

$$u(kT) = \frac{u((k-1)T) + aTe(kT)}{1 + aT}.$$

- Finding the transfer function:

$$(1 + aT)U(z) = z^{-1}U(z) + aTE(z)$$

$$U(z)[1 + aT - z^{-1}] = aTE(z)$$

$$\frac{U(z)}{E(z)} = \frac{aT}{1 + aT - z^{-1}} = \frac{a}{\frac{1-z^{-1}}{T} + a} = \frac{a}{\frac{z-1}{Tz} + a}.$$

- “s” has been replaced by “ $\frac{z-1}{Tz}$ ”.

Method 3: Use the bilinear/Tustin/trapezoid rule

- Using the *bilinear rule*:

$$u(kT) \approx u((k-1)T) + (T/2)[-au((k-1)T) + ae((k-1)T) - au(kT) + ae(kT)]$$

$$u(kT) \left[1 + \frac{aT}{2}\right] = u((k-1)T) \left[1 - \frac{aT}{2}\right] + \frac{aT}{2}[e((k-1)T) + e(kT)].$$

- Finding the transfer function:

$$U(z) \left[1 + \frac{aT}{2}\right] = z^{-1}U(z) \left[1 - \frac{aT}{2}\right] + \frac{aT}{2}E(z)[z^{-1} + 1]$$

$$\frac{U(z)}{E(z)} = \frac{aT(z+1)}{(2+aT)z + aT - 2} = \frac{a}{\frac{2}{T} \left[\frac{z-1}{z+1}\right] + a}.$$

- “s” has been replaced by “ $\frac{2}{T} \left(\frac{z-1}{z+1}\right)$ ”.

2.10: Stability and prewarping

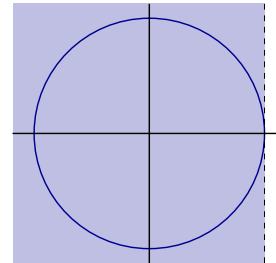
- Each of these transformations maps the s -plane to the z -plane. How do they compare?
- Consider mapping $s = j\omega$ (the stability boundary).

1. Forward rectangular rule:

$$s = \frac{z - 1}{T}$$

$$z = 1 + Ts$$

$$= 1 + Tj\omega.$$



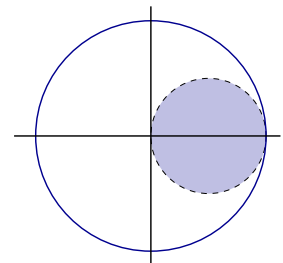
2. Backward rectangular rule:

$$s = \frac{z - 1}{Tz} \implies z(Ts - 1) = -1$$

$$z = \frac{1}{1 - Ts} = \frac{1}{1 - Tj\omega}$$

$$= \frac{1}{2} + \left(\frac{1}{1 - Tj\omega} - \frac{1}{2} \right) = \frac{1}{2} + \frac{1}{2} \left(\frac{1 + Tj\omega}{1 - Tj\omega} \right).$$

- The term in the parenthesis has magnitude 1 for all ω .



3. Bilinear rule:

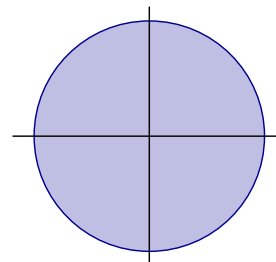
$$s = \frac{2}{T} \left(\frac{z - 1}{z + 1} \right)$$

$$zTs + Ts = 2z - 2$$

$$z(Ts - 2) = -2 - Ts$$

$$z = \frac{2 + Ts}{2 - Ts} = \frac{2 + Tj\omega}{2 - Tj\omega}.$$

- Constant magnitude 1 for all ω .



COMMENTS:

- Rule (1) may possibly map a stable $D(s)$ into an unstable $D(z)$ (!!)
- Rule (2) always maintains stability (can even map an unstable $D(s)$ into a stable $D(z)$ (!) but does not do a good job of mapping frequency response, especially at high frequencies.
- Rule (3) maps stability information *EXACTLY* (important later in the course). The $j\omega$ -axis in the s -plane mapped directly to the unit circle in the z -plane. ω from $0 \dots \infty$ mapped to an angle $0 \dots \pi$. Frequency compression/warping.
- Since the bilinear/Tustin rule is by far the most used, let us examine the frequency warping some more.
- Let frequency in the digital domain be ω_z , and corresponding frequency in the analog domain be ω_s .
- Convert $H(s)$ to $H(z)$.

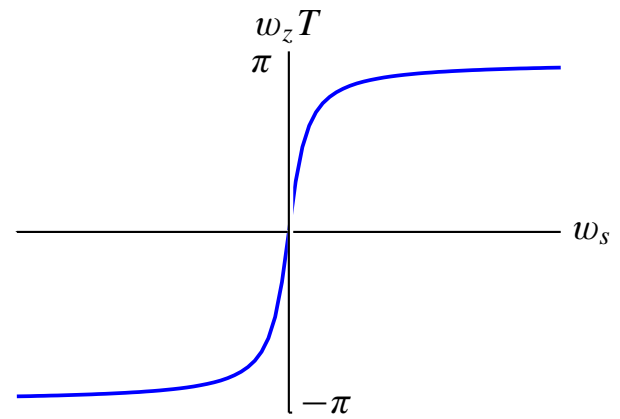
$$H(z) = H(s) \Big|_{s = \frac{2}{T} \frac{z-1}{z+1}}$$

- Let $z = e^{j\omega_z T}$ (a sinusoid of frequency ω_z).

$$\begin{aligned} H(e^{j\omega_z T}) &= H(s) \Big|_{s = \frac{2}{T} \frac{e^{j\omega_z T} - 1}{e^{j\omega_z T} + 1}} \\ &= H(s) \Big|_{s = \frac{2}{T} \left(\frac{e^{j\omega_z T/2} - e^{-j\omega_z T/2}}{e^{j\omega_z T/2} + e^{-j\omega_z T/2}} \right)} \\ &= H(s) \Big|_{s = j \frac{2}{T} \tan\left(\frac{\omega_z T}{2}\right)} \end{aligned}$$

- So, $\omega_s = \frac{2}{T} \tan\left(\frac{\omega_z T}{2}\right)$ and $\omega_z = \frac{2}{T} \tan^{-1}\left(\frac{\omega_s T}{2}\right)$.

- Near $\omega_z = 0$, $\omega_s \approx \omega_z$, but frequencies do not match anywhere else. Important?
- What if $H(s)$ is designed using Bode techniques to have a certain bandwidth? This bandwidth will be warped!
- One solution: Redesign $H(s)$ to take into account eventual warping.
 1. We want our controller to have bandwidth $= \omega_{z1}$.
 2. “Pre-warp” our design spec. $\omega_{s1} = \frac{2}{T} \tan \frac{\omega_{z1} T}{2}$.
 3. Design $H(s)$ to have bandwidth ω_{s1} .
 4. Convert $H(s)$ to $H(z)$ via bilinear transform. Bandwidth $= \omega_{z1}$.
- Note: ω_{z1} is just some critical frequency to be matched. It does not need to be bandwidth.
- Another solution: Force warped frequency axis to match at desired frequency:



$$\text{let } s = \alpha \frac{z - 1}{z + 1}$$

$$H(z) = H(s)|_{s=\alpha \frac{z-1}{z+1}}$$

$$\begin{aligned} H(e^{j\omega_z T}) &= H(s)|_{s=\alpha \frac{e^{j\omega_z T}-1}{e^{j\omega_z T}+1}} \\ &= H(s)|_{s=\alpha \left(\frac{e^{j\omega_z T/2}-e^{-j\omega_z T/2}}{e^{j\omega_z T/2}+e^{-j\omega_z T/2}} \right)} \\ &= H(s)|_{s=j\alpha \tan\left(\frac{\omega_z T}{2}\right)}. \end{aligned}$$

- So, $\omega_s = \alpha \tan(\omega_z T/2)$. Let $\alpha = \frac{\omega_1}{\tan(\omega_1 T/2)}$.

- Then, at frequency $\omega_z = \omega_1$,

$$\omega_s = \frac{\omega_1}{\tan(\omega_1 T/2)} \tan(\omega_1 T/2) = \omega_1.$$

- A match! Drawback: Can match at only one frequency.
- In MATLAB:

```
sysd=c2d(sys,T,'tustin'); % bilinear  
sysd=c2d(sys,T,'prewarp',w1); % w/ freq. match.
```


2.11: Examples and hold-based methods

EXAMPLE: Convert $H(s) = \frac{1}{s^3 + 2s^2 + 2s + 1}$ to digital.

Method 1) $s = \frac{z-1}{T}$.

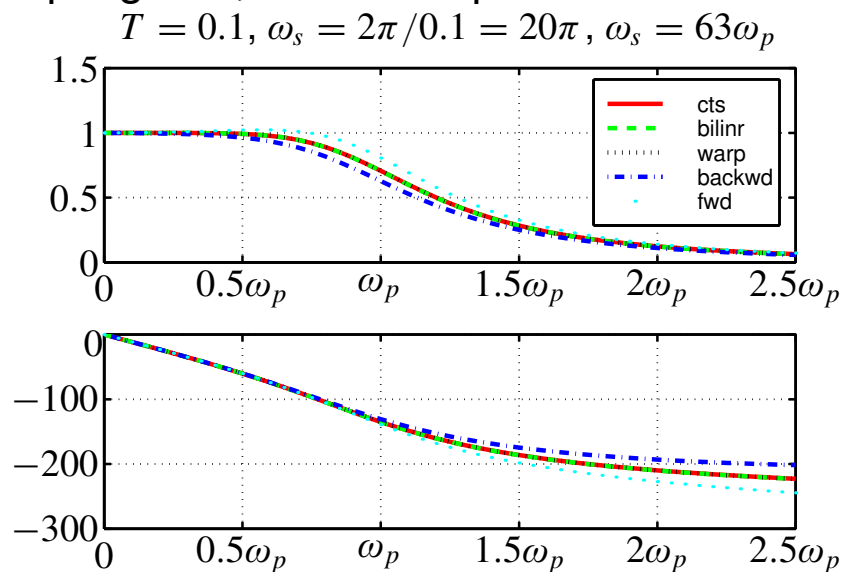
$$\begin{aligned} H(z) &= \frac{1}{\frac{(z-1)^3}{T^3} + 2\frac{(z-1)^2}{T^2} + 2\frac{(z-1)}{T} + 1} \\ &= \frac{T^3}{(z^3 - 3z^2 + 3z - 1) + 2T(z^2 - 2z + 1) + 2T^2(z - 1) + T^3} \\ &= \frac{T^3}{z^3 + z^2(2T - 3) + z(3 - 4T + 2T^2) + (2T - 2T^2 + T^3 - 1)}. \end{aligned}$$

Method 2) $s = \frac{z-1}{Tz}$. Gives: $H(z) = \frac{1}{\frac{(z-1)^3}{(Tz)^3} + 2\frac{(z-1)^2}{(Tz)^2} + 2\frac{(z-1)}{Tz} + 1}$.

- The rest of the math is up to you...

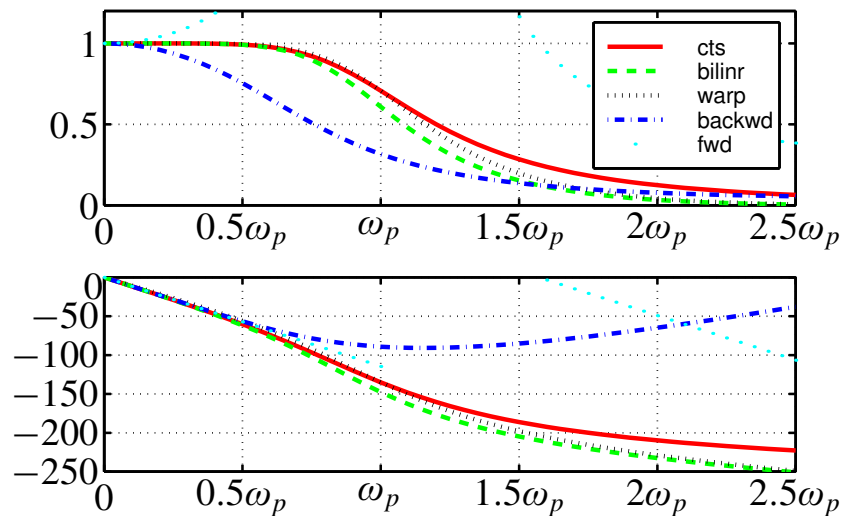
Method 3) $s = \frac{2z-1}{Tz+1}$. Gives: $H(z) = \frac{1}{\frac{8}{T^3} \frac{(z-1)^3}{(z+1)^3} + 2\frac{4}{T^2} \frac{(z-1)^2}{(z+1)^2} + 2\frac{2}{T} \frac{(z-1)}{(z+1)} + 1}$.

- Again, the rest of the math is up to you.
- For a fast sampling rate, these compare as:



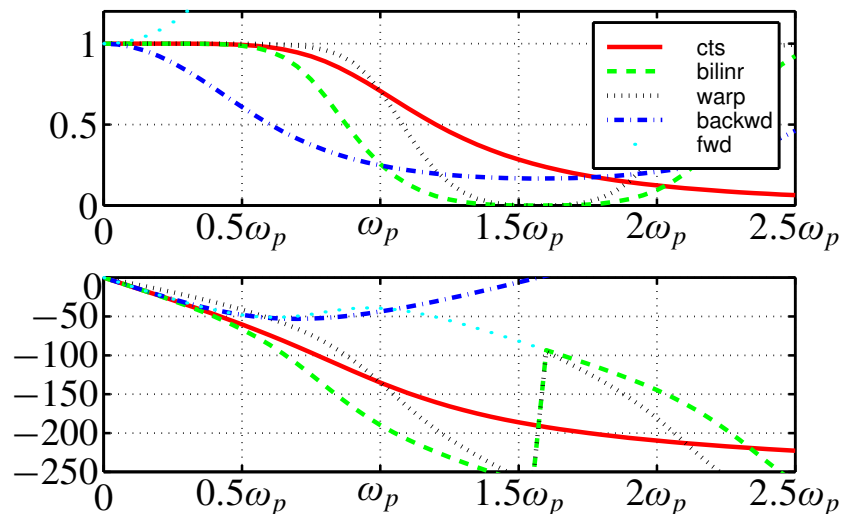
- For a medium sampling rate, these compare as:

$$T = 1, \omega_s = 2\pi/1 = 2\pi, \omega_s = 6.3\omega_p$$



- For a slow sampling rate, these compare as:

$$T = 2, \omega_s = \pi, \omega_s = 3.14\omega_p$$



Zero-pole matching

- There are other methods for emulating continuous-time systems with discrete-time systems.
- For example, we have seen before that if we take the z -transform of a *signal*, the poles of the z -transform are related to the poles of the s -transform via $z = e^{sT}$.

- There was no direct mapping of zeros.
- Now, we consider transforming a system. Rules:
 1. Poles map as $z = e^{sT}$.
 2. Finite zeros map as $z = e^{sT}$.
 3. Zeros at ∞ (high freq.) map to $z = -1$ (high freq).
(but, may want to move one zero from $z = -1$ to ∞ to give processing time: zero at $\infty =$ pole at $0 = 1/z =$ delay).
 4. Match the gain of $H(s)$ and $H(z)$ at dc.

EXAMPLE: First method: move zero at ∞ to $z = -1$:

$$H(s) = \frac{a}{s + a} \quad \rightsquigarrow \text{pole at } -a, \text{ zero at } \infty$$

$$H(z) = k \frac{z + 1}{z - e^{-aT}}.$$

- Match at dc: $H_s(0) = 1$, $H_z(1) = \frac{2k}{1 - e^{-aT}} \dots k = \frac{1 - e^{-aT}}{2}$.

$$H(z) = \frac{(z + 1)(1 - e^{-aT})}{2(z - e^{-aT})}.$$

EXAMPLE: Second method: move zero at ∞ to origin:

$$H(z) = \frac{k}{(z - e^{-aT})}.$$

- Match at dc: $H_s(0) = 1$, $H_z(1) = \frac{k}{1 - e^{-aT}} \dots k = 1 - e^{-aT}$.

$$H(z) = \left(\frac{1 - e^{-aT}}{z - e^{-aT}} \right).$$

```
sysd=c2d(sys,T,'matched');
```

Hold equivalents

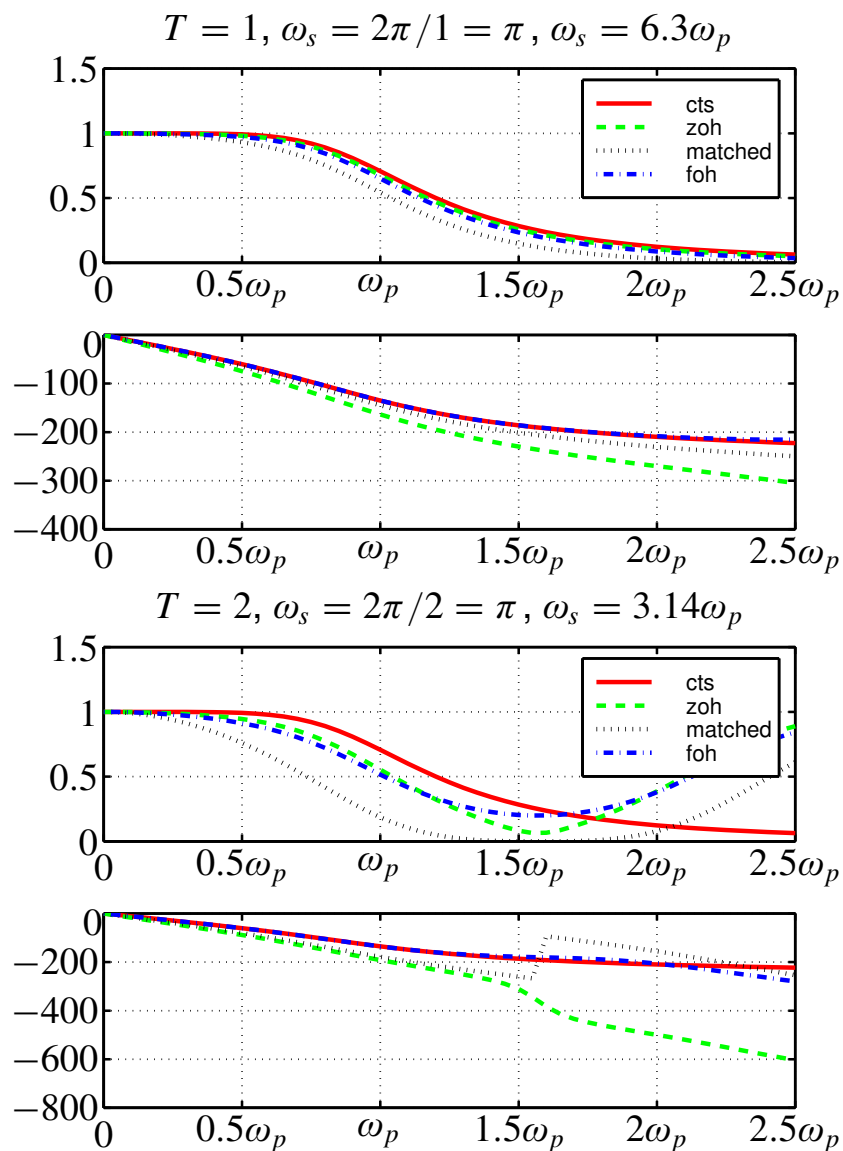
- Two other methods I won't discuss in detail.

- ZOH: $\text{sysd} = \text{c2d}(\text{sys}, T, 'zoh')$ $H(z) = \frac{z-1}{z} \mathcal{Z} \left\{ \frac{H(s)}{s} \right\}$

- FOH: $\text{sysd} = \text{c2d}(\text{sys}, T, 'foh')$ $H(z) = \frac{(z-1)^2}{Tz} \mathcal{Z} \left\{ \frac{H(s)}{s^2} \right\}$

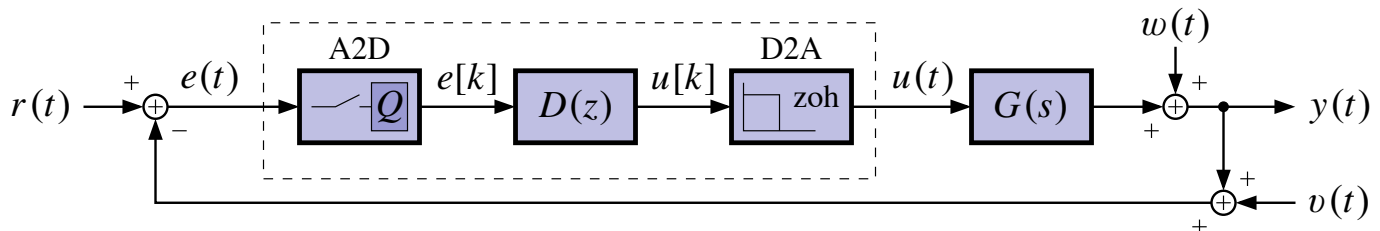
- The $\mathcal{Z}\{\cdot\}$ notation will be explored in detail in Chapter 3.

- System from pg. 2-41 revisited. . .



2.12: Design by emulation

- These emulation techniques approximate an *open-loop* system $D(s)$ with an *open-loop* system $D(z)$.
- What happens when $D(z)$ is placed in a feedback loop?



- We will look at this in more detail later.
- Recall, though, that the biggest difference in performance will be due to the $T/2$ second delay in the ZOH.

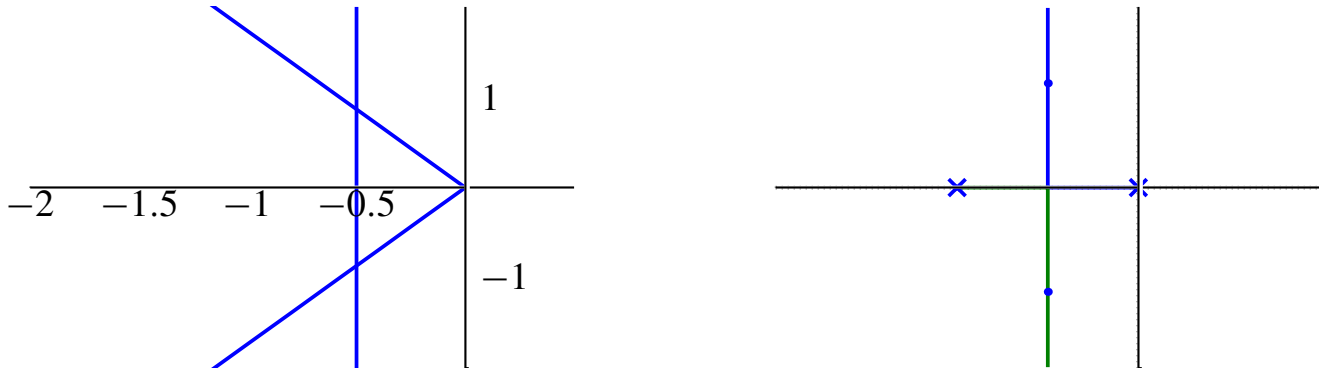
EXAMPLE:

$$G(s) = \frac{1}{s(10s + 1)}.$$

- Specifications for the design:
 1. $M_p < 16\%$.
 2. $t_s < 10$ s to 1%.
 3. e_{ss} for ramp input < 0.01 , slope of ramp = 0.01.
 4. Sampling time to give ≥ 10 samples in rise time.
- Step 1: Design $D(s)$.
 - “1” $\Rightarrow \zeta \geq 0.5$.
 - “2” $\Rightarrow \sigma \geq 0.46$.
 - “3” $\Rightarrow K_v = \frac{0.01}{0.01} = 1.0$.

Choose $D(s)$ to cancel plant pole approximately.

$$D(s) = \frac{10s + 1}{s + 1}.$$



$K = 1 \dots \rightsquigarrow \dots K_v = 1$ and good pole locations.

$\omega_n \approx 1$ for pole locations, so $t_r \approx 1.8$ sec.

“4” $T = t_r/10 = 0.18 \rightsquigarrow$ choose $T = 0.2$.

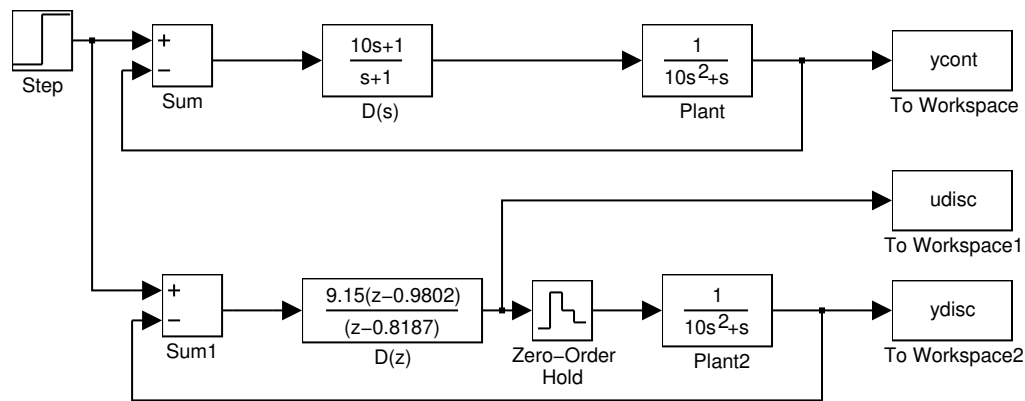
■ Step 2: $D(s) \Rightarrow D(z)$: Use pole-zero matching.

- $D(s)$ has zero at $-1/10 \rightsquigarrow D(z)$ has zero at $z = e^{-0.1T}$.
- $D(s)$ has pole at $-1 \rightsquigarrow D(z)$ has pole at $z = e^{-T}$.
- $D(s)$ has dc-gain 1 $\rightsquigarrow D(z)$ has dc-gain 1.

$$D(z) = K \frac{(z - 0.9802)}{(z - 0.8187)}.$$

- DC-gain of $D(z) = \lim_{z \rightarrow 1} D(z) = K \frac{(1 - 0.9802)}{(1 - 0.8187)} \dots K = 9.15$.

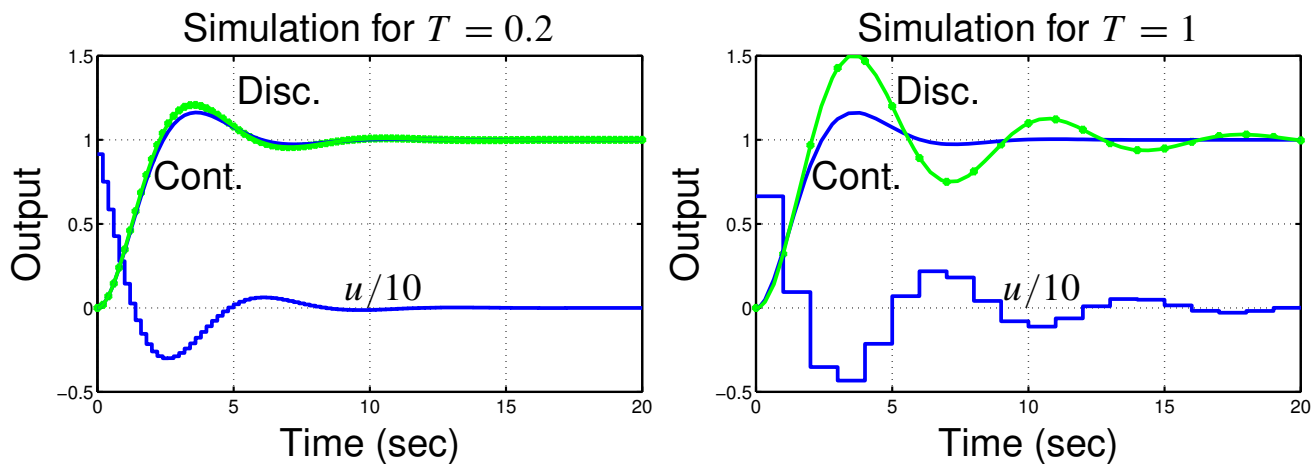
$$D(z) = 9.15 \frac{(z - 0.9802)}{(z - 0.8187)}.$$



- Try again, but with 1.0 sec sampling period.

$$D(z) = 6.64 \frac{(z - 0.9048)}{(z - 0.3679)}$$

- Much more overshoot, poorer damping.



Where to from here?

- We have completed our look at design by emulation.
- We now move in the direction of direct digital-control design.
- First, we need to spend some time understanding the D2A and A2D operations thoroughly.