

4CM00: Control Engineering

Digital filters

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Where innovation starts

- ▶ Derive model / measurement of the plant
- ▶ Design a controller
- ▶ Implement controller in real-time environment...

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- ▶ Design a controller
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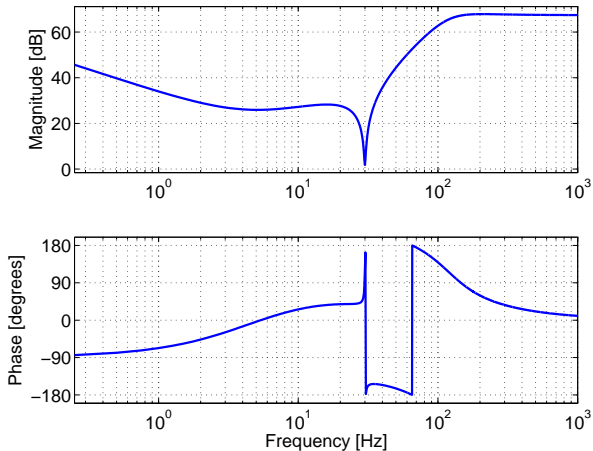
unexpected problems,
maybe even instability!



WHY ??

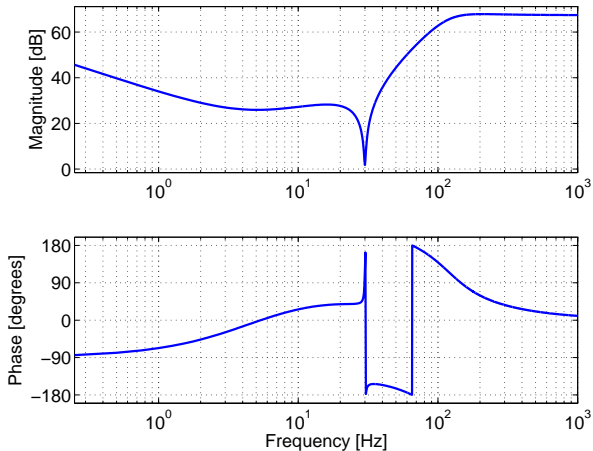
Designed controller $C(s)$:

Bode diagram

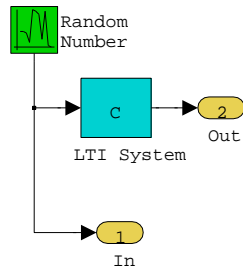


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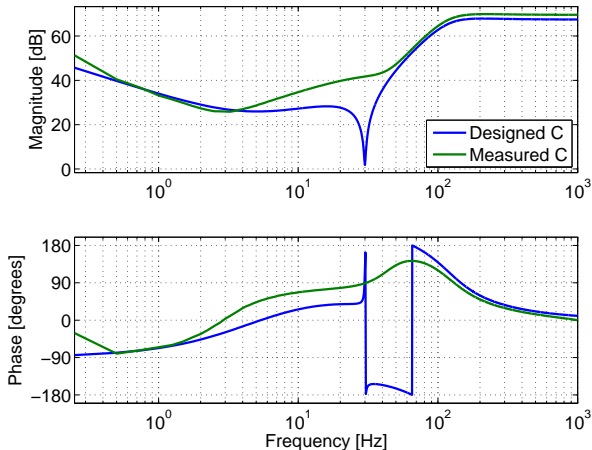


Measurement:

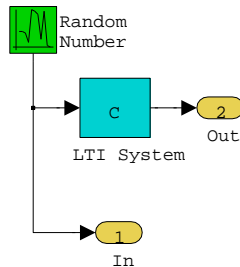


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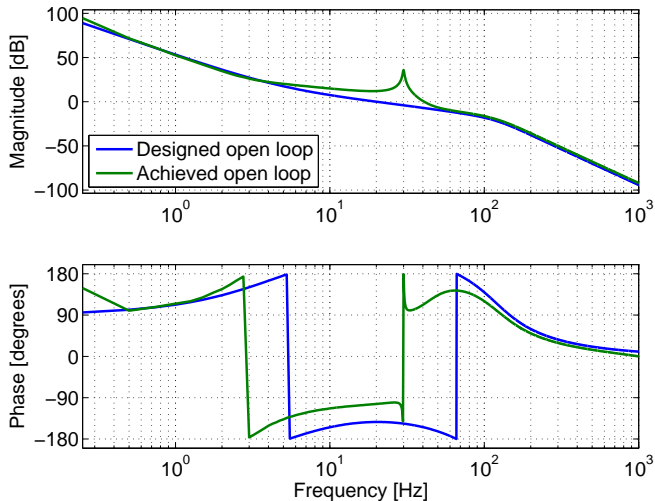
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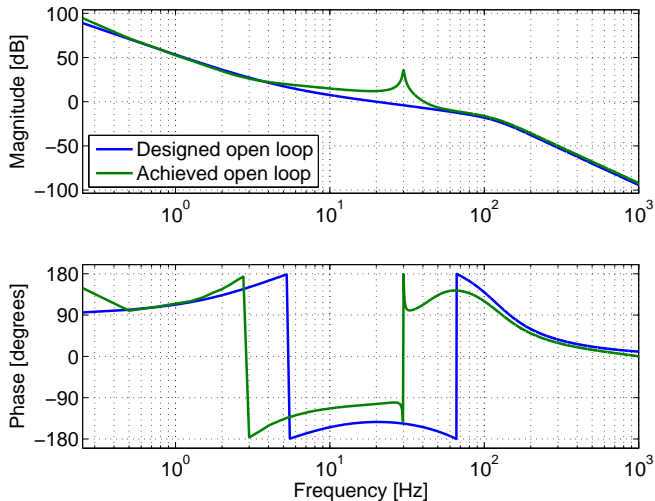
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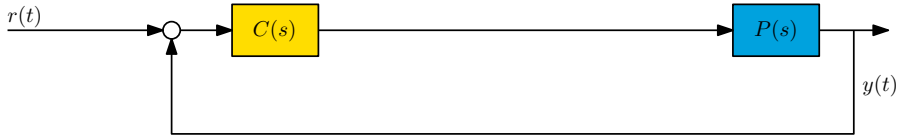


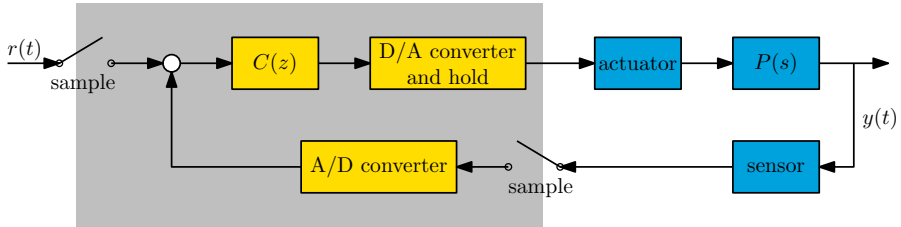
Bode diagram

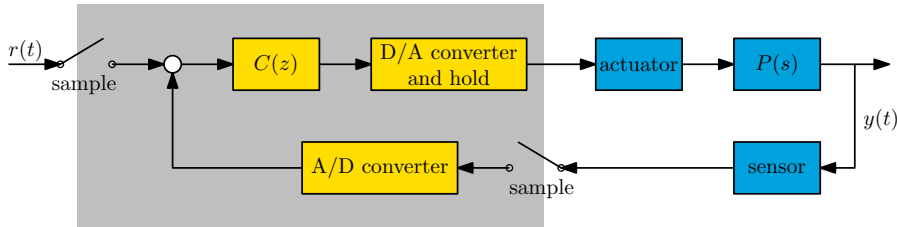


Achieved closed loop system is **unstable!**

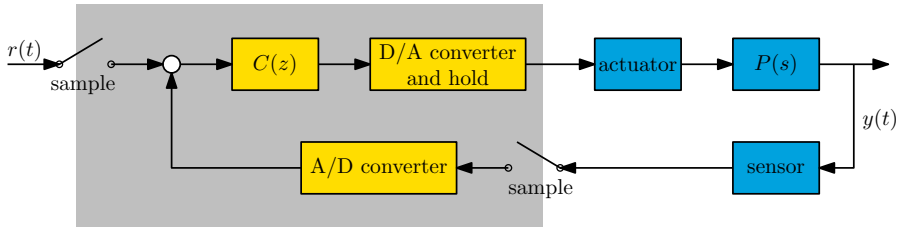
Real-time implementation







- ▶ Measurement $y(t)$ is sampled every T seconds: $y(t) \rightarrow y(kT)$
- ▶ Analog value of $y(kT)$ is digitized
 - e.g. by an encoder
 - e.g. by a 10, 12 or 16 bits A/D converter
- ▶ Controller $C(z)$ computes new output $u(kT)$ based on measurements $y(kT)$, $y((k-1)T)$, $y((k-2)T)$, etc.
- ▶ D/A converter creates analog continuous signal by holding the output $u(kT)$ until the next sample



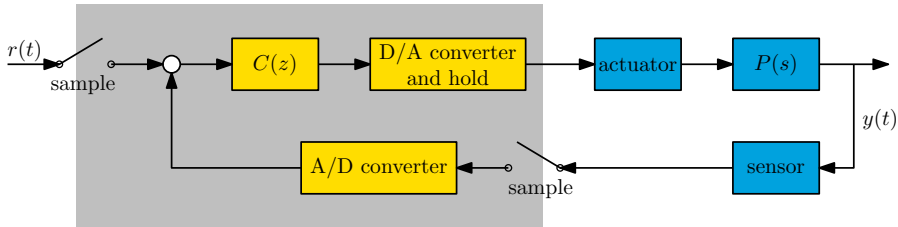
Continuous time:

Controller $C(s)$ computes $u(t)$ based on continuous time signals $e(t)$, $\dot{e}(t)$, $\ddot{e}(t)$, \dots , $\int e(t)$, $\iint e(t)$, \dots

Discrete time:

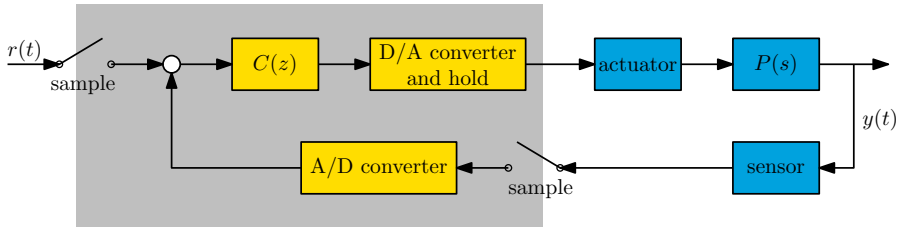
Controller $C(z)$ computes $u(k)$ based on sampled signals $e(k)$, $e(k-1)$, $e(k-2)$, \dots , $u(k-1)$, $u(k-2)$, \dots

Consequence: $C(s) \neq C(z)$!!!



How to make correct controller $C(z)$?

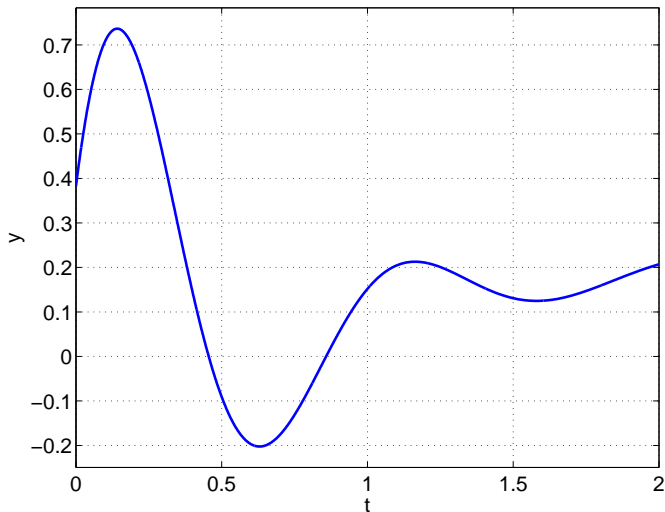
- ▶ Discrete filter design
 - Design $C(z)$ from scratch in z-domain



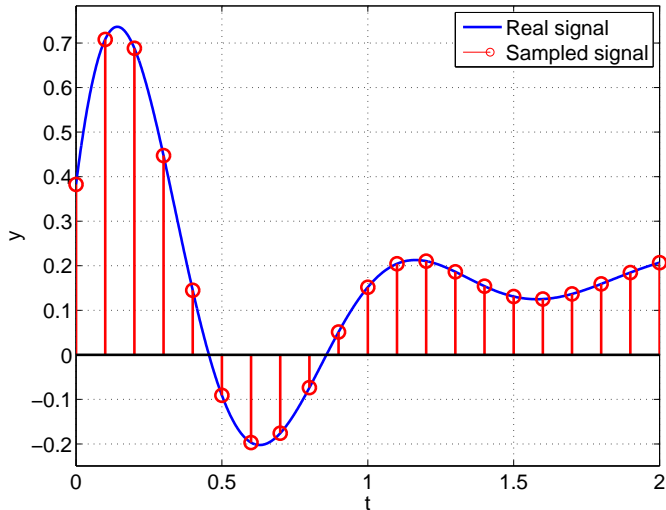
How to make correct controller $C(z)$?

- ▶ Discrete filter design
 - Design $C(z)$ from scratch in z -domain
- ▶ Emulation: discretize continuous time controller $C(s)$
 - Various discretization methods possible
 - Accuracy depends on method and sample time

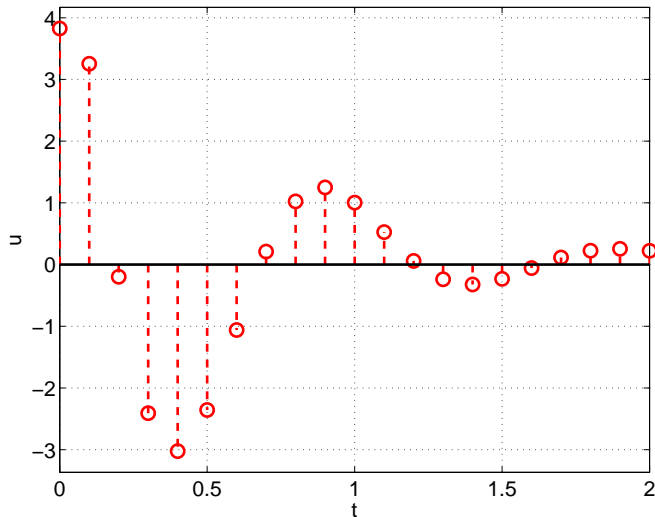
Measured plant output $y(t)$:



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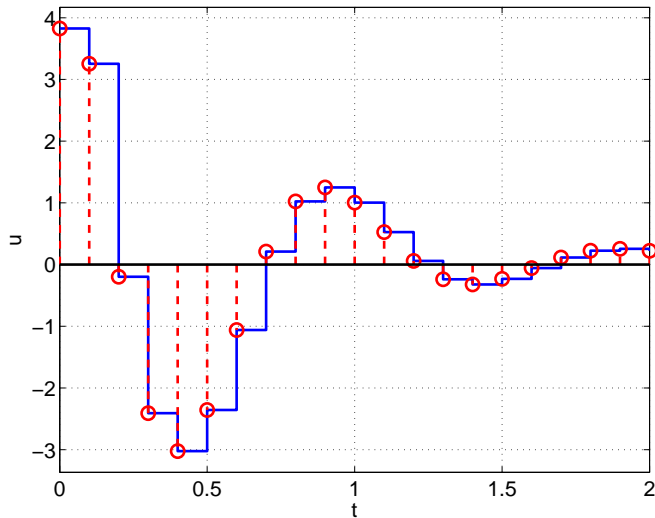


Calculated and applied controller output $u(t)$:



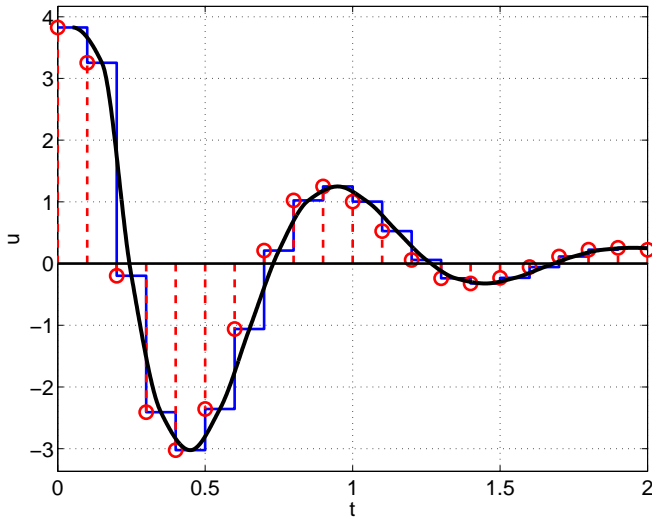
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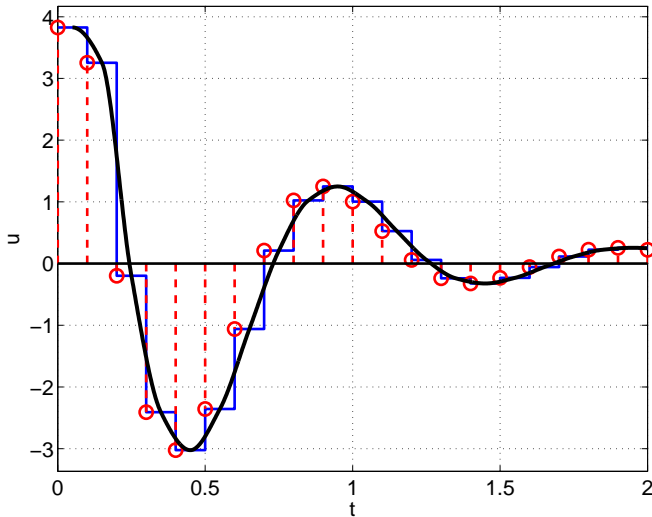
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- Applied signal

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- ▶ Calculated signal
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- ▶ Equivalent signal

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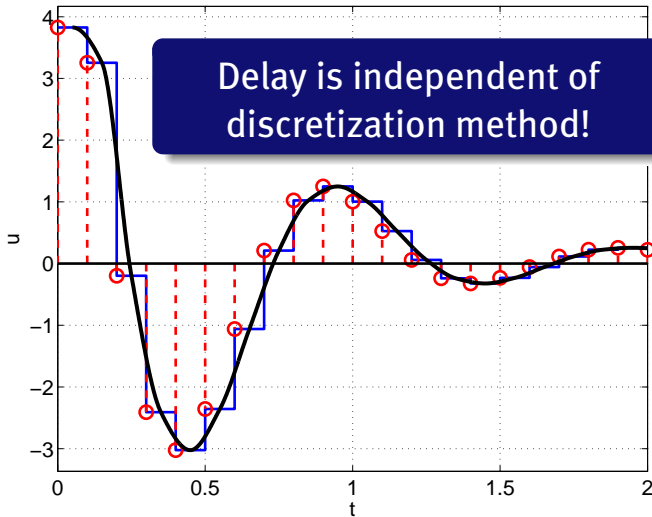


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delay: $\frac{1}{2}T$

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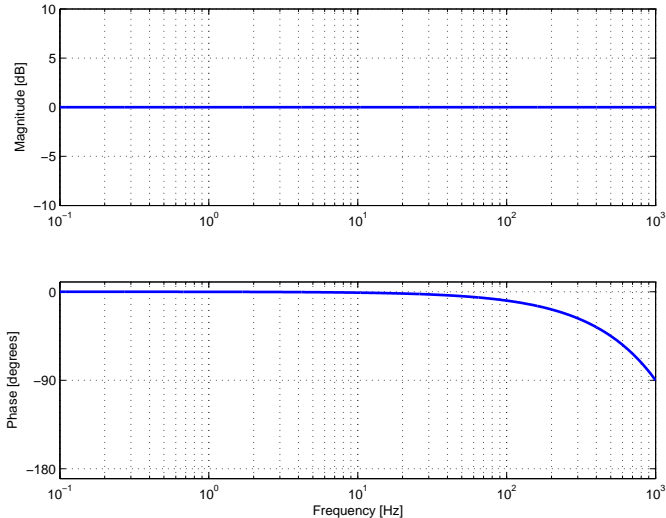


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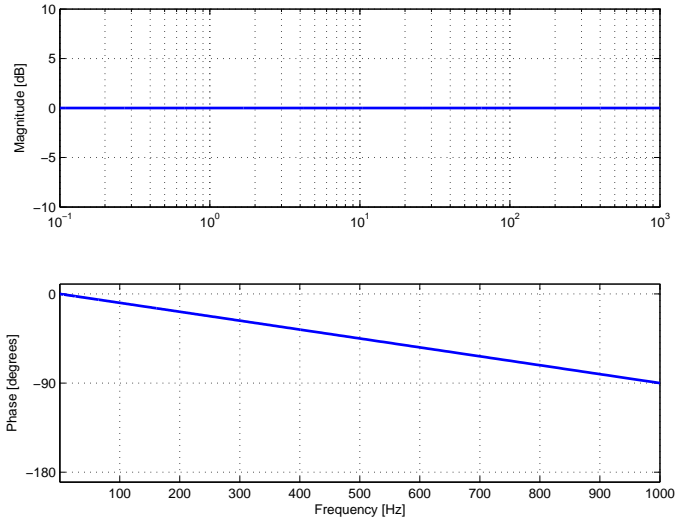


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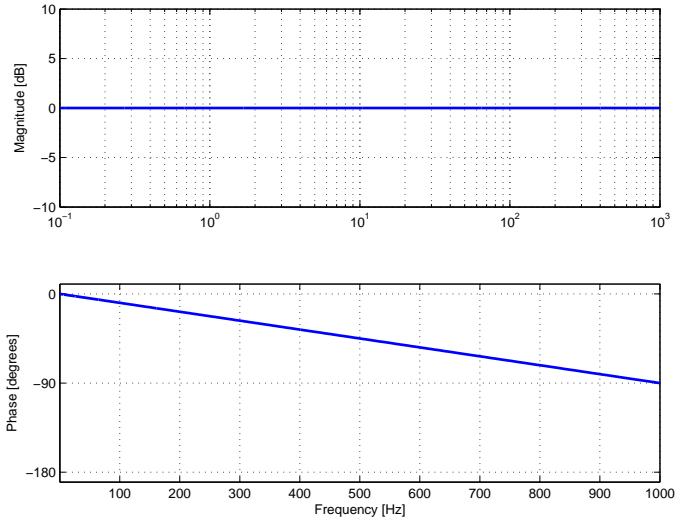
Phase delay is frequency dependent!



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Phase delay is frequency dependent! $\Rightarrow H_d(j\omega) = e^{-j\omega T_d}$



Introduction to discrete time systems

Discrete time systems produce sampled outputs based on sampled inputs.

$$\begin{array}{lll} \text{Continuous:} & \leftrightarrow & \text{Discrete:} \\ \text{time } t & \leftrightarrow & \text{time step } k \\ \text{signal } x(t) & \leftrightarrow & \text{sample } x(k) \\ \dot{x}(t) = f(x(t)) & \leftrightarrow & x(k+1) = f(x(k), x(k-1), \dots) \end{array}$$

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Past, present and future values of $x(k)$ are denoted by the **shift operator** z :

$$\begin{aligned}x(k+1) &= zx(k) \\ x(k-1) &= z^{-1}x(k) \\ x(k-N) &= z^{-N}x(k)\end{aligned}$$

Shift operator z is discrete counterpart of Laplace variable s :

$$\dot{x}(s) = sx(s) \leftrightarrow x(k+1) = zx(k)$$

The z transform can be used to derive I/O models.

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Example:

$$y(k) = -a_1 y(k-1) + b_0 u(k) + b_1 u(k-1)$$

$$y(k) = -a_1 z^{-1} y(k) + b_0 u(k) + b_1 z^{-1} u(k)$$

$$(1 + a_1 z^{-1}) y(k) = (b_0 + b_1 z^{-1}) u(k)$$

$$H(z) = \frac{y(k)}{u(k)} = \frac{b_0 + b_1 z^{-1}}{1 + a_1 z^{-1}}$$

$$\text{or equivalently: } H(z) = \frac{b_0 z + b_1}{z + a_1}.$$

Backwards derivation:

- ▶ derive the difference equation of the transfer function

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Answer:

$$\begin{aligned}\frac{y(k)}{u(k)} &= \frac{z}{z - 1} \\ (z - 1)y(k) &= zu(k) \\ y(k + 1) - y(k) &= u(k + 1) \\ y(k + 1) &= u(k + 1) + y(k)\end{aligned}$$

or equivalently: $y(k) = u(k) + y(k - 1)$.

	<u>Continuous time</u>	<u>Discrete time</u>
Transfer function	$H(s) = \frac{\text{num}(s)}{\text{den}(s)}$	$H(z) = \frac{\text{num}(z)}{\text{den}(z)}$
Poles λ_i	$\text{den}(s = \lambda_i) = 0$	$\text{den}(z = \lambda_i) = 0$
Stable if	$\text{Re}(\lambda_i) < 0 \quad \forall \lambda_i$	$ \lambda_i < 1 \quad \forall \lambda_i$
Stability boundary	imaginary axis: $s = j\omega$	unit disc: $z = \cos \omega + j \sin \omega = e^{j\omega}$

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Discrete time frequency ω :

Frequency ω is normalized, such that $\omega \in [-\pi, \pi]$.

This corresponds to a sampling time $T = 1$.

Actual frequency $\omega_s = \frac{\omega}{T_s}$, where T_s is actual sample time.

Example 1:

$$y(k + 1) = ay(k)$$

Consider a time sequence:

$$y(k + 2) = ay(k + 1) = a^2y(k)$$

$$y(k + 3) = ay(k + 2) = a^3y(k)$$

$$y(k + N) = a^Ny(k)$$

Note: $y(k + N) \rightarrow 0$ for $N \rightarrow \infty$ if and only if $a^N \rightarrow 0$, i.e. $|a| < 1$.

Example 2:

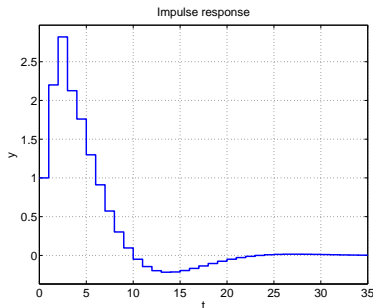
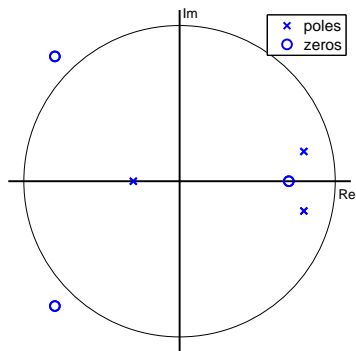
$$H(z) = \frac{z^3 + 0.9z^2 + 0.16z - 0.9}{z^3 - 1.3z^2 + 0.2z + 0.2}$$

$$\text{Poles: } \lambda^3 - 1.3\lambda^2 + 0.2\lambda + 0.2 = 0 \quad \Rightarrow \quad \lambda = \begin{bmatrix} 0.80 + 0.19j \\ 0.80 - 0.19j \\ -0.30 \end{bmatrix}$$

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Discrete frequency response

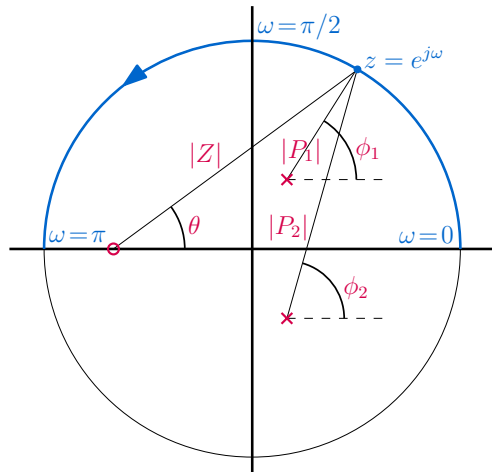
19/39

Continuous: Substitute $s = j\omega$ into $H(s)$ to obtain $H(j\omega)$

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Example:

$$H(z) = \frac{z + b}{(z + a_1)(z + a_2)}$$

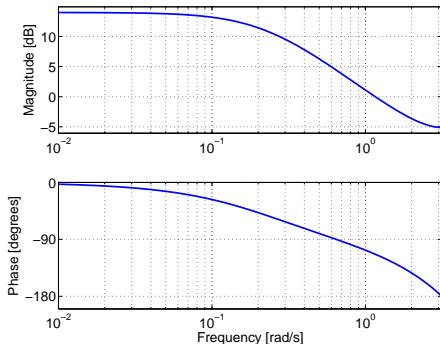
Magnitude and phase:

$$\begin{aligned} |H(j\omega)| &= \frac{|e^{j\omega} + b|}{|e^{j\omega} + a_1||e^{j\omega} + a_2|} \\ &= \frac{|Z|}{|P_1| \cdot |P_2|} \\ \angle H(j\omega) &= \theta - \phi_1 - \phi_2 \end{aligned}$$

- ▶ Stable poles and zeros can have 180° phase change

$$H(z) = \frac{1}{z-0.8}$$

Bode diagram

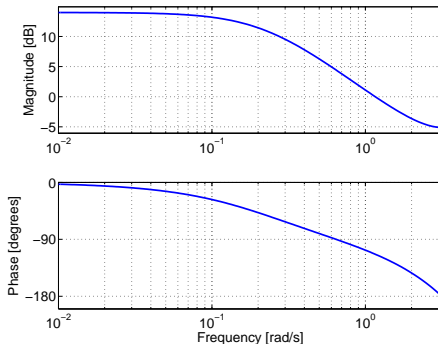


Information up to Nyquist frequency

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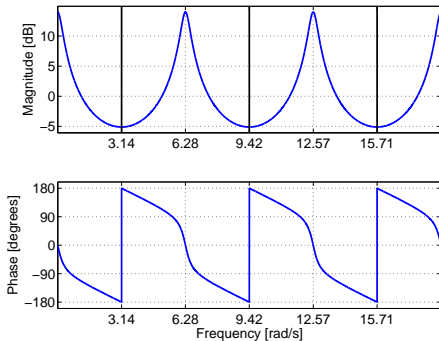


Information up to Nyquist frequency

- ▶ Frequency response repeats itself every 2π rad/s

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Bode diagram



Plotted on linear frequency scale

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Answer: Yes, $\lambda_D = e^{T_s \lambda_C}$

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- ▶ Stable poles / zeros are mapped **inside** the unit disc
 $\text{Re}(\lambda_C) < 0 \Rightarrow |\lambda_D| < 1$
 - ▶ Unstable poles / zeros are mapped **outside** the unit disc
 $\text{Re}(\lambda_C) > 0 \Rightarrow |\lambda_D| > 1$

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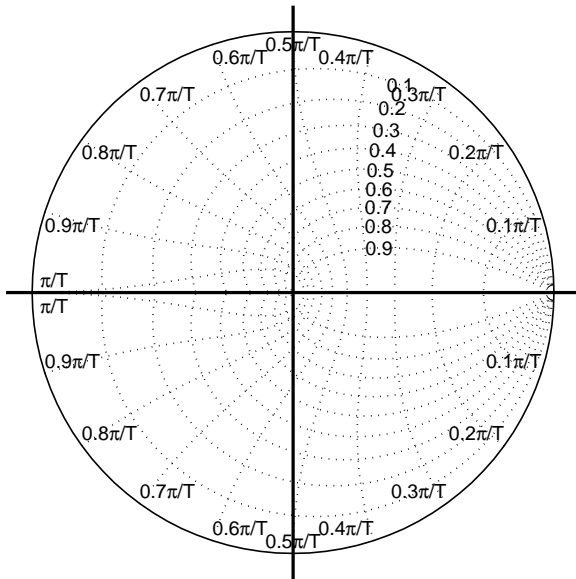
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To find frequency of discrete pole / zero, use inverse relation:

$$f = \left| \frac{\ln(\lambda_D)}{2\pi T_s} \right|$$

E.g. when $f_s = 1000\text{Hz}$ we have that $\lambda_D = 0.8$ is located at $f = 35.5\text{Hz}$.



- ▶ Stability boundary:
 $s = j\omega \Rightarrow |z| = 1$
- ▶ 'Origin':
 $s = 0 \Rightarrow z = +1$
- ▶ 'Infinite' pole:
 $s = -\infty \Rightarrow z = 0$
- ▶ Vertical lines in
s-plane \Rightarrow circles in
z-plane
- ▶ Horizontal lines in
s-plane \Rightarrow radial
lines in z-plane

It is easy to derive the impulse response from the z-transform:

- ▶ write $H(z)$ in the form $H(z) = c_0 + c_1z^{-1} + c_2z^{-2} + c_3z^{-3} + \dots$
- ▶ i.e. eliminate denominator term

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Difference equation then becomes:

$$y(k) = c_0 u(k) + c_1 u(k-1) + c_2 u(k-2) + c_3 u(k-3) + \dots$$

Impulse input: $u(k) = 1$ for $k = 0$, and $u(k) = 0$ for $k \neq 0$.

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- ▶ So coefficient c_i determines output: $y(i) = c_i$

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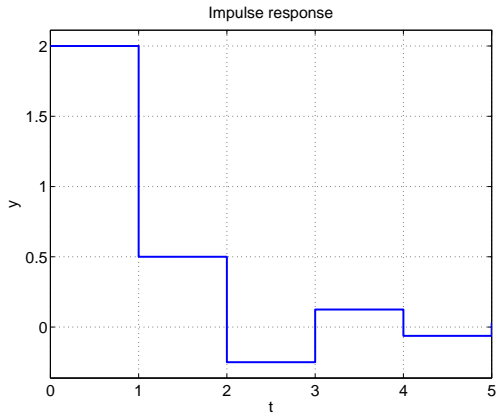
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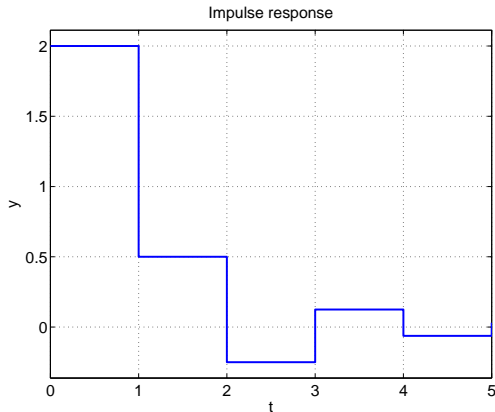
Note: In general, expansion of $H(z)$ is *infinite*. However, in some cases $H(z)$ has a **Finite Impulse Response**. $H(z)$ is then a FIR filter.

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



$$H(z) = \frac{4 + 3z^{-1}}{2 + z^{-1}} = 2 + \frac{1}{2}z^{-1} - \frac{1}{4}z^{-2} + \frac{1}{8}z^{-3} - \dots$$

$$y(k) = 2u(k) + \frac{1}{2}u(k-1) - \frac{1}{4}u(k-2) + \frac{1}{8}u(k-3) - \dots$$



Ordinary causal systems:

- ▶ output only depends on present and past input values:
 $y(k) = f(u(k), u(k-1), u(k-2), \dots)$

Anti-causal systems:

- ▶ output also depends on future input values
- ▶ 'predicts' future
- ▶ do not occur in reality

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How to recognize anti-causal systems?

- ▶ Impulse response of $H(z)$:
contains **positive powers** of shift operator, i.e. z, z^2 , etc.
- ▶ Transfer function $H(z)$:
numerator has higher degree than denominator,
i.e. $H(z)$ has more zeros than poles.

Controller design

Implementation procedure

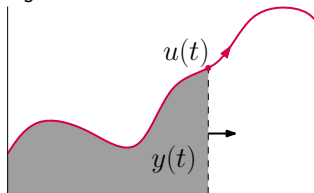
- ▶ Design continuous time controller $C(s)$
- ▶ Discretize the controller
 - Time domain approximation
 - Forward Euler (Zero-order hold)
 - Backward Euler
 - Tustin
 - Frequency domain approximation
 - Tustin with prewarping
 - Pole-zero matching
- ▶ Measure obtained controller $C(z)$ in open loop
 - Compare with design!
 - Adjust controller if necessary
- ▶ Implement $C(z)$ in closed loop

- ▶ Imitate time domain effect of controller $C(s)$
- ▶ Make assumptions on behavior of $u(t)$ between samples:
 - hold: take $u(t)$ constant between samples (Euler)
 - linear: assume linear change from $u(k-1)$ to $u(k)$ (Tustin)

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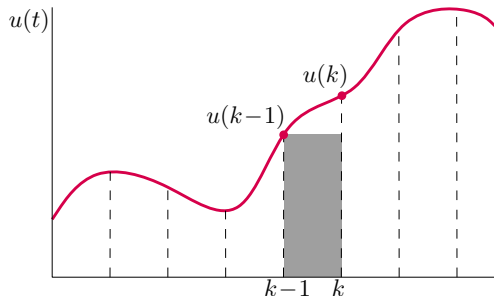
Consider continuous time integrator: $C(s) = \frac{1}{s}$

$$y = \int u \, dt \quad \text{or} \quad \dot{y} = u$$



- ▶ $y(t)$ is surface beneath curve $u(t)$
- ▶ $y(k)$ is previous value $y(k-1)$ + additional surface
- ▶ Sample time: T

Assume $u(t)$ constant from $k-1$ to k .



Forward Euler:

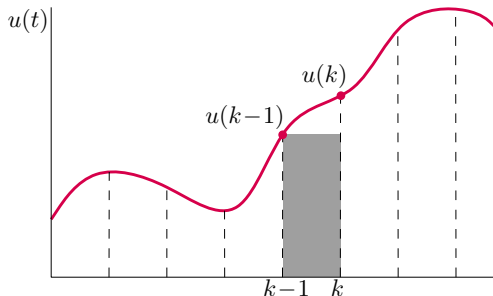
$$y(k) = y(k-1) + Tu(k-1)$$

Approximation of $C(s)$:

$$C(z) = \frac{Tz^{-1}}{1 - z^{-1}} = \frac{T}{z - 1}$$

Forward Euler: $\frac{1}{s} \rightarrow \frac{T}{z-1}$ and $s \rightarrow \frac{z-1}{T}$

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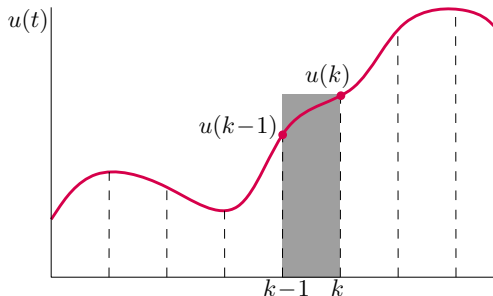
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Example:

$$C(s) = \frac{s+a}{s+b} \Rightarrow C(z) = \frac{z-1+aT}{z-1+bT}$$

Drawback: $C(z)$ is anti-causal for non-proper $C(s)$ (derivative-terms)!

Assume $u(t)$ was constant from k to $k-1$.



Backward Euler:

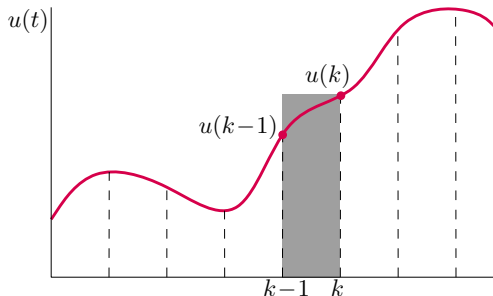
$$y(k) = y(k-1) + Tu(k)$$

Approximation of $C(s)$:

$$C(z) = \frac{T}{1 - z^{-1}} = \frac{Tz}{z - 1}$$

Backward Euler: $\frac{1}{s} \rightarrow \frac{Tz}{z-1}$ and $s \rightarrow \frac{z-1}{Tz}$

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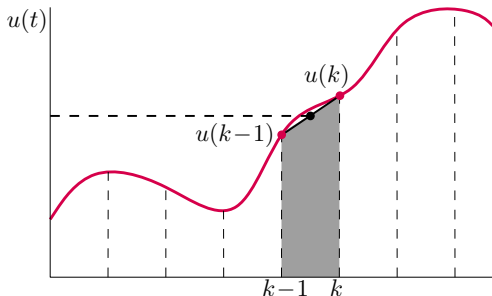
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Example:

$$C(s) = \frac{s + a}{s + b} \Rightarrow C(z) = \frac{(1 + aT)z - 1}{(1 + bT)z - 1}$$

Assume linear change from $u(k-1)$ to $u(k)$



Tustin:

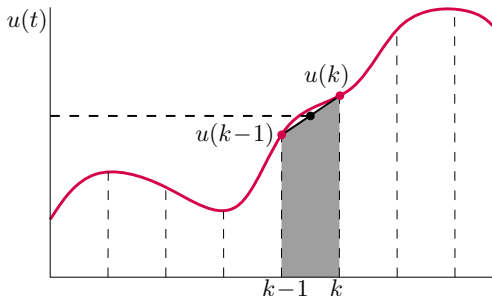
$$y(k) = y(k-1) + T \frac{u(k) + u(k-1)}{2}$$

Approximation of $C(s)$:

$$C(z) = \frac{T(1 + z^{-1})}{2(1 - z^{-1})} = \frac{T(z + 1)}{2(z - 1)}$$

Tustin: $\frac{1}{s} \rightarrow \frac{T(z+1)}{2(z-1)}$ and $s \rightarrow \frac{2(z-1)}{T(z+1)}$

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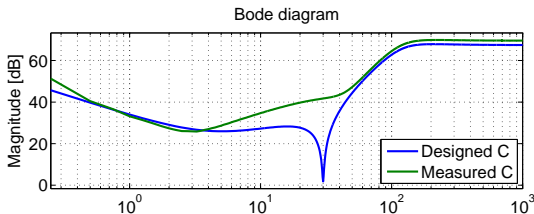
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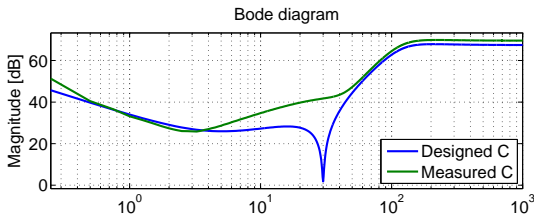
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$$C(s) = \frac{s + a}{s + b} \Rightarrow C(z) = \frac{2(z - 1) + aT(z + 1)}{2(z - 1) + bT(z + 1)}$$

Note: Time domain approximations (especially Euler) do not take frequency response information into account!
Especially ‘fast’ dynamics in $C(s)$ can be ‘missed’ by the discretization, e.g. (anti-)resonances.



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Solution:

- ▶ Include frequency information into discretization:
 - Tustin with prewarping
 - Pole-zero matching

Extension of the Tustin discretization:

- ▶ Prewarping matches frequency responses of $C(z)$ and $C(s)$ at a specified frequency ω_p
- ▶ Combination of time and frequency domain techniques

Tustin substitution changes into:

$$s \rightarrow \frac{\omega_p}{\tan(\omega_p T/2)} \cdot \frac{z - 1}{z + 1}$$

As a consequence:

$$C(s = j\omega_p) = C(z = e^{j\omega_p T})$$

- ▶ Completely frequency based
- ▶ Maps all poles and zeros to the discrete domain

Procedure:

- ▶ Determine poles p_i and zeros q_i of $C(s)$
- ▶ Map p_i and q_i to discrete domain:
 - discrete poles at $z = e^{p_i T}$
 - discrete zeros at $z = e^{q_i T}$
- ▶ Match low-frequency DC-gain of $C(z)$ to $C(s)$

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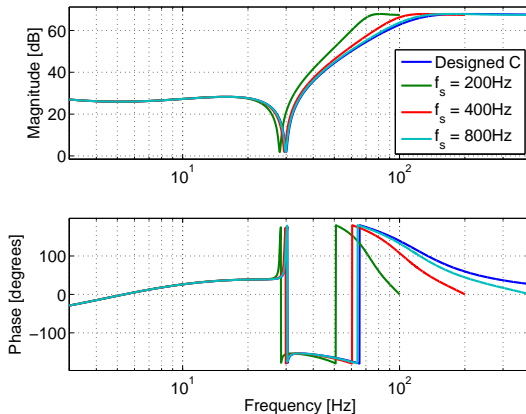
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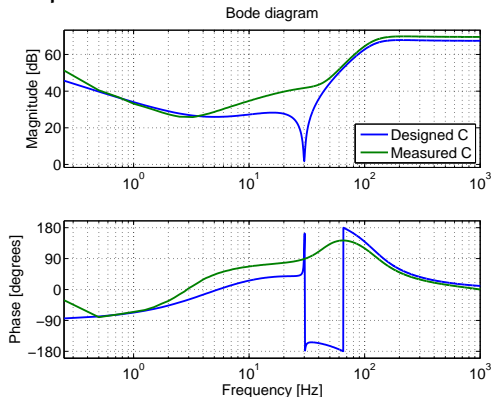
$$C(s) = \frac{s + a}{s + b} \Rightarrow C(z) = \frac{a}{b} \cdot \frac{1 - e^{-bT}}{1 - e^{-aT}} \cdot \frac{z - e^{-aT}}{z - e^{-bT}}$$

For all discretization methods:

- ▶ The smaller the sample time T , the better the approximation
- ▶ In practice, choose sample frequency at least 20 times bandwidth



Back to our original problem:



Simulink chooses its own discretization method for $C(s)$:

- ▶ uses zero-order hold (ZOH)
- ▶ same as forward Euler for small sample times T

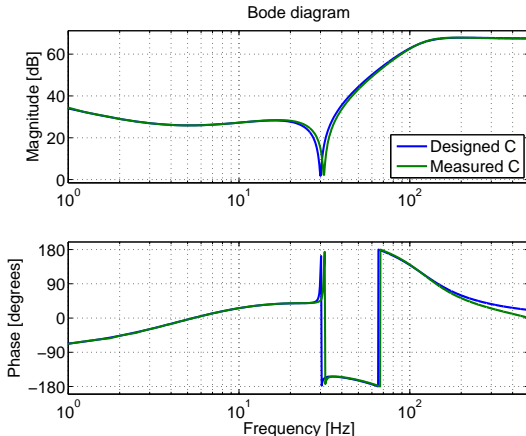
Shapelt comes with the DCtools blockset.

Controller blocks are discretized versions of continuous filters:

- ▶ using Tustin with prewarping
(prewarp frequency differs per block)

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Controller blocks are discretized versions of continuous filters:

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Alternatively, discretize $C(s)$ yourself

- ▶ choose an appropriate method
- ▶ choose a sample frequency
- ▶ compare $C(s)$ and $C(z)$
- ▶ make adjustments in $C(s)$, f_s or the method if necessary
- ▶ implement the obtained $C(z)$

Note: If $C(s)$ and $C(z)$ differ, try to reason how it would alter the open-loop. Will $C(z)$ make it worse or better?

**Always measure your
controller before
implementing!**