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# Mechatronics

Week 4 Day 1

# Previously

- We distinguished between A, T, D-type elements, and A,T-variables
- We learned modelling multi-domain systems following a five-step procedure to obtain their state-space representation
- We modelled various multi-domain systems involving electrical, mechanical (Newtonian Approach, Vectorial Mechanics), and fluid domains
- We studied modelling mechanical systems via Euler-Lagrange formalism (Analytical Mechanics)



# Today's lecture: Sampling and Discretization

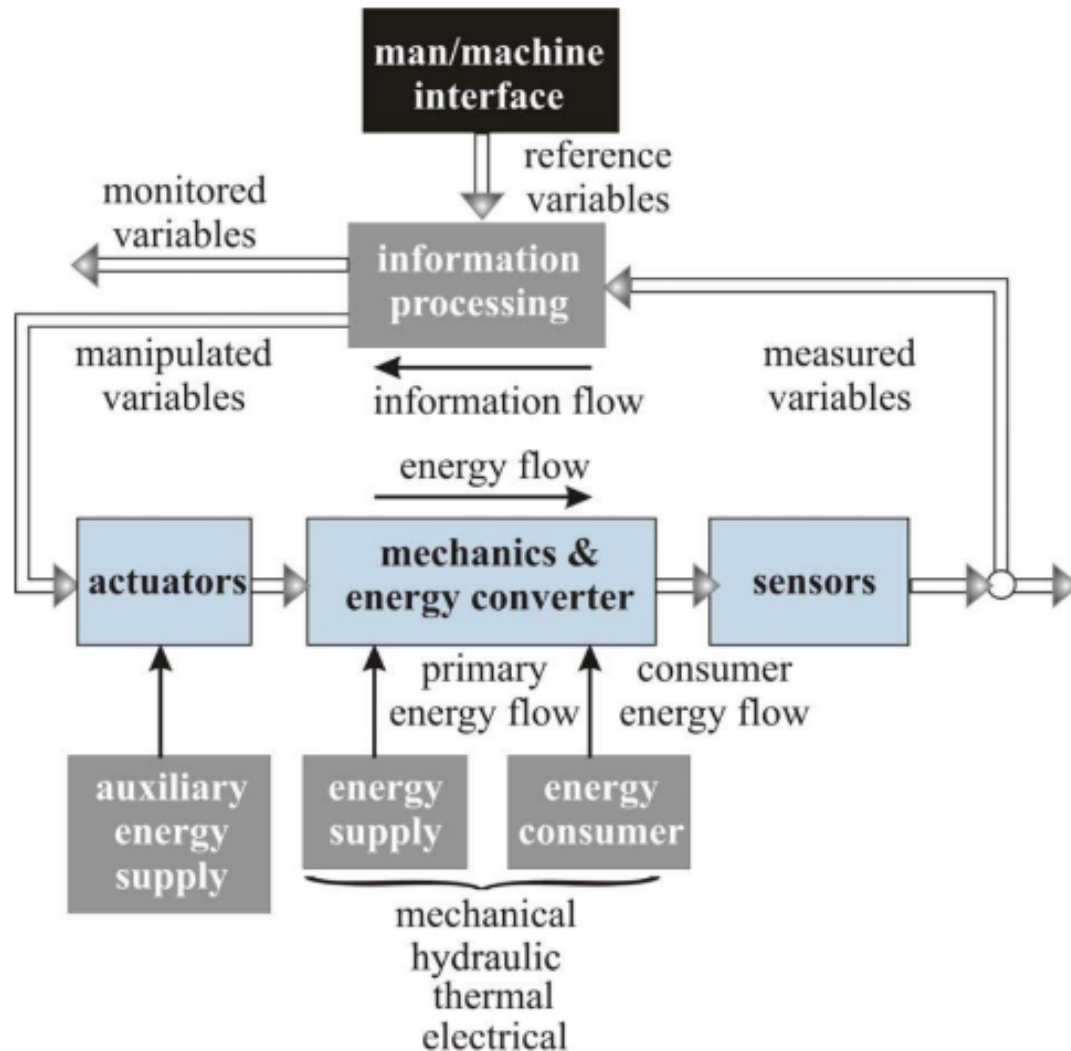


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# Motivation

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Advancements in digital technology have led to a rapid growth in the use of computers for

- data acquisition,
- information,
- processing, and
- control of continuous processes

# Motivation

Continuous-time (**analog**) data **cannot** be **directly handled** by computers.

To use microcontrollers-based information processing data needs to be **transformed**.

Sensor

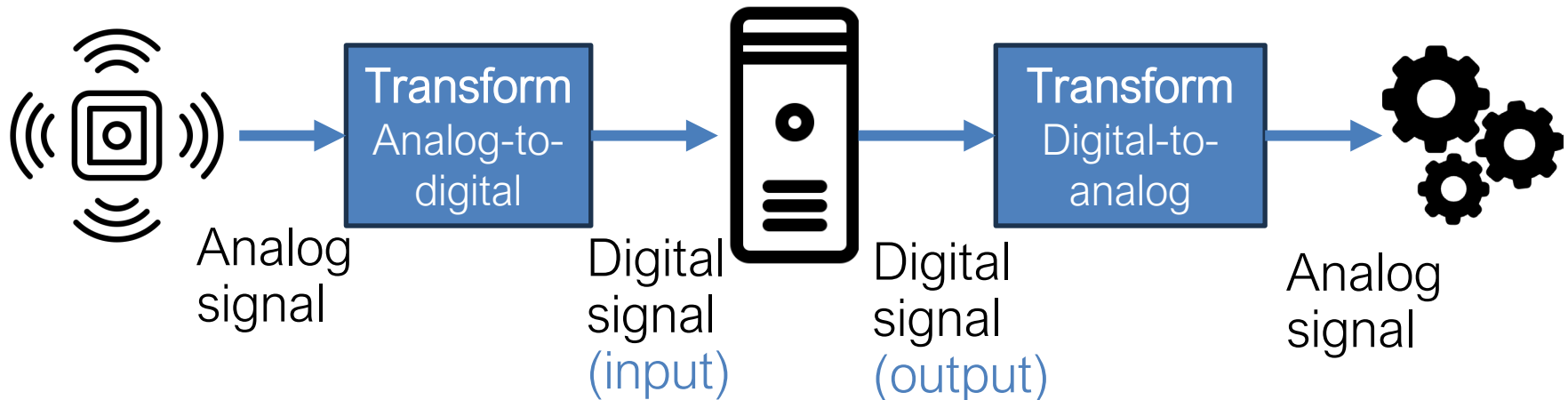
Catches signal

Digital system

Processes+controls

Actuator

Performs actions





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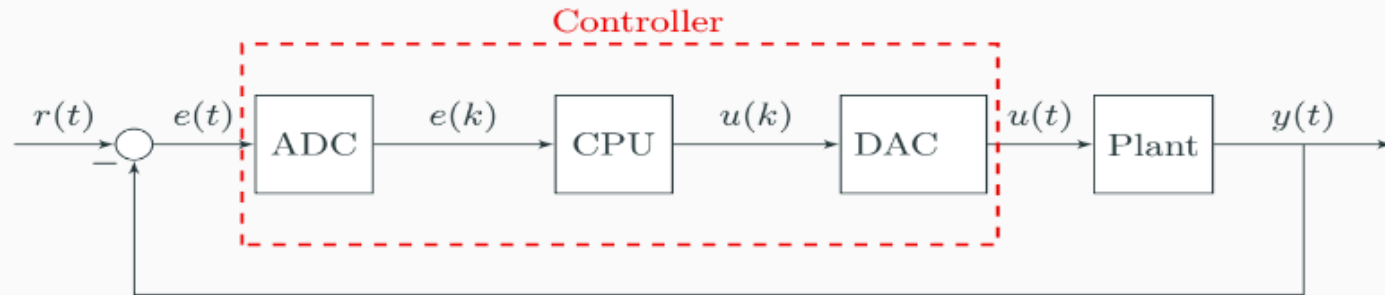
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# Modern Mechatronics Systems

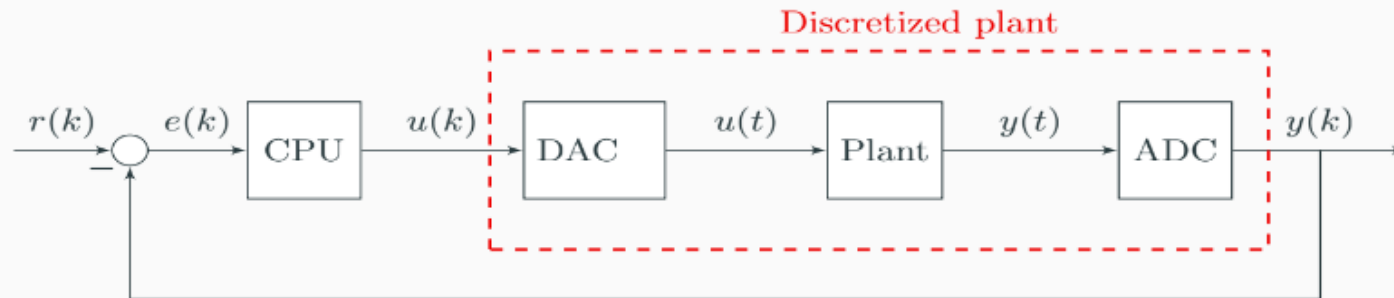
# Modern Mechatronics Systems

Implementing mechatronics systems in practice

## 1. Implementation via discrete controller



## 2. Implementation by discretizing plant







# Learning objectives

After today's lecture, you will be able to

- Familiarize with the **discretization** of continuous-time signals and systems
- Appreciate the significance of choosing appropriate **sample rate** for the discretization
- Obtain the **transfer function** of a **discrete-time** system
- Learn various **approximation methods** and tools to discretize a continuous time system



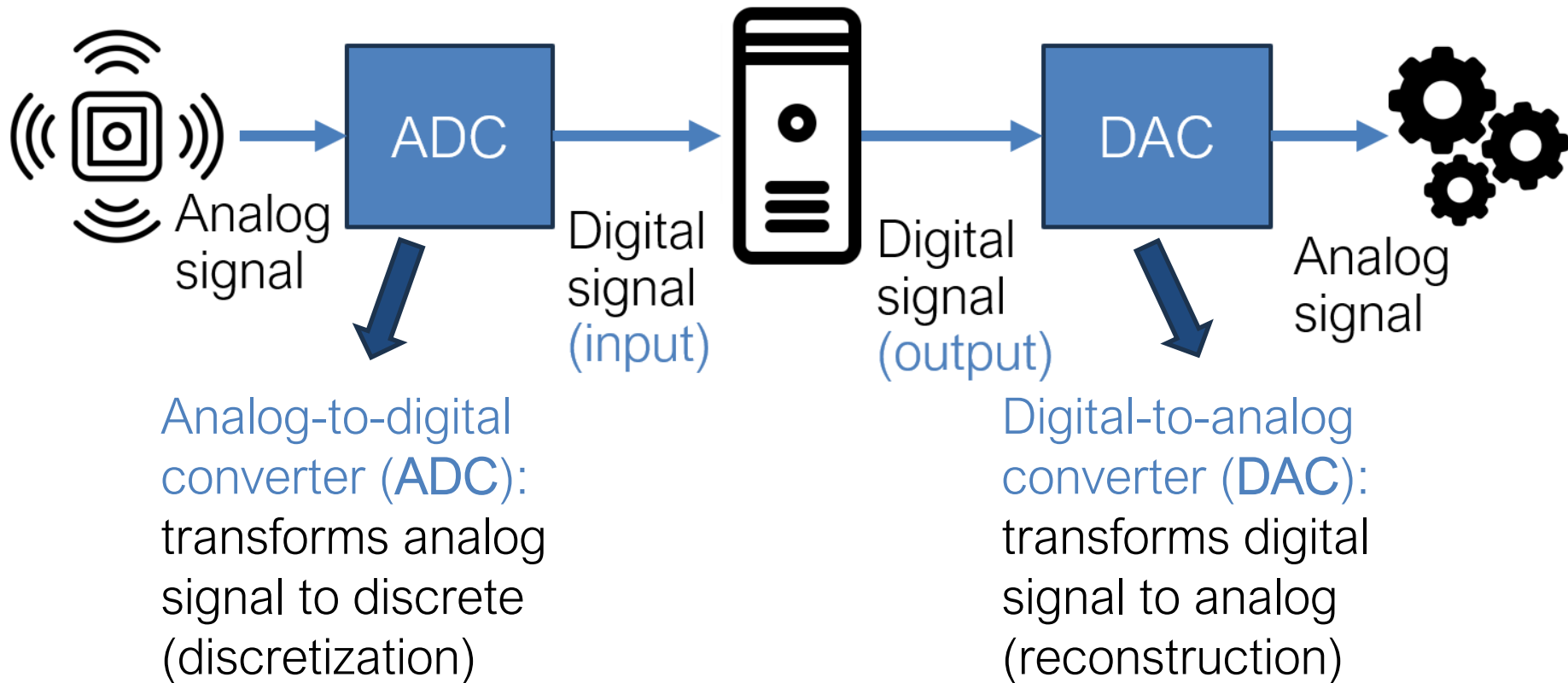
# Discretization and Reconstruction

# Discretization and Reconstruction

Sensor  
Catches signal

Digital system  
Processes+controls

Actuator  
Performs task



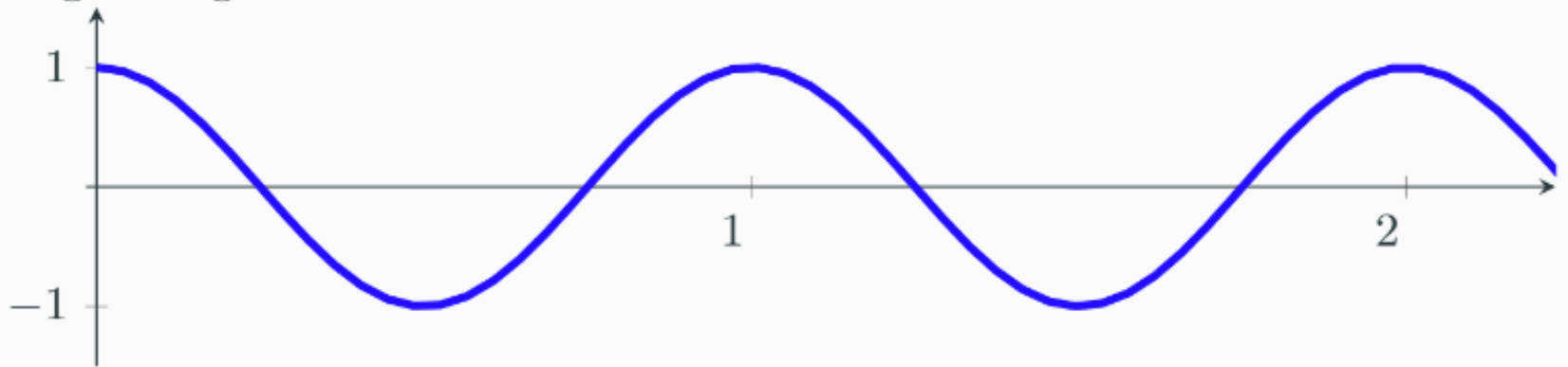


# Sampling theorem



# Nyquist-Shannon sampling theorem

Analog signal



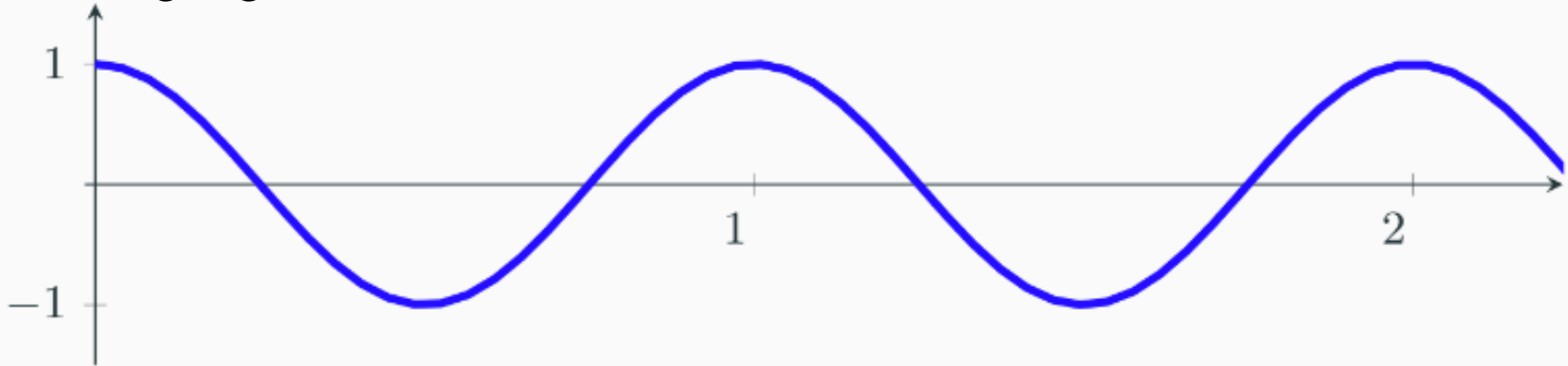
**Frequency ( $\omega$ ):** number of cycles completed in a second

For the given signal,  $\omega = 1Hz$

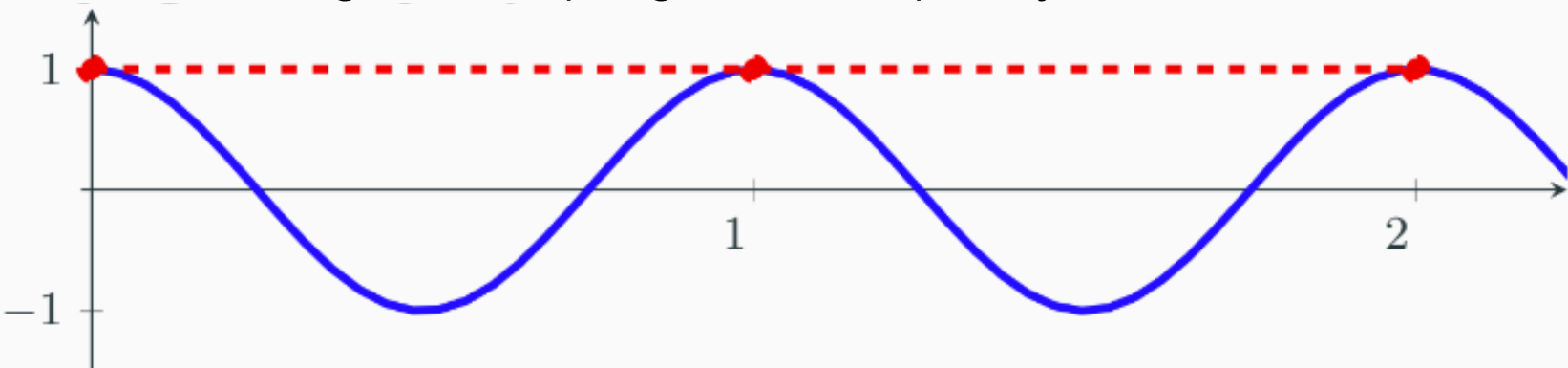


# Nyquist-Shannon sampling theorem

Analog signal

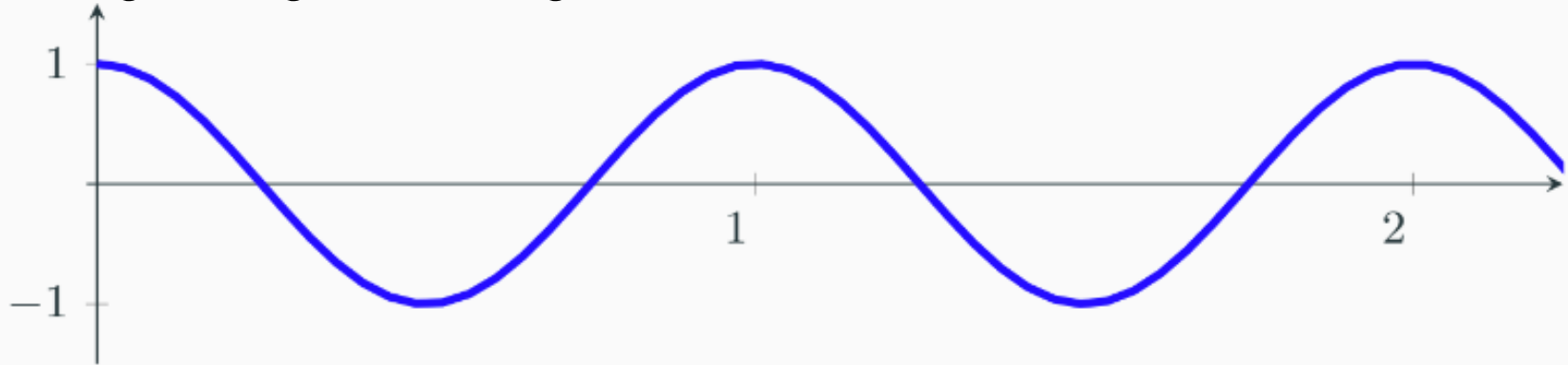


Discretized signal, sampling **one** time per cycle

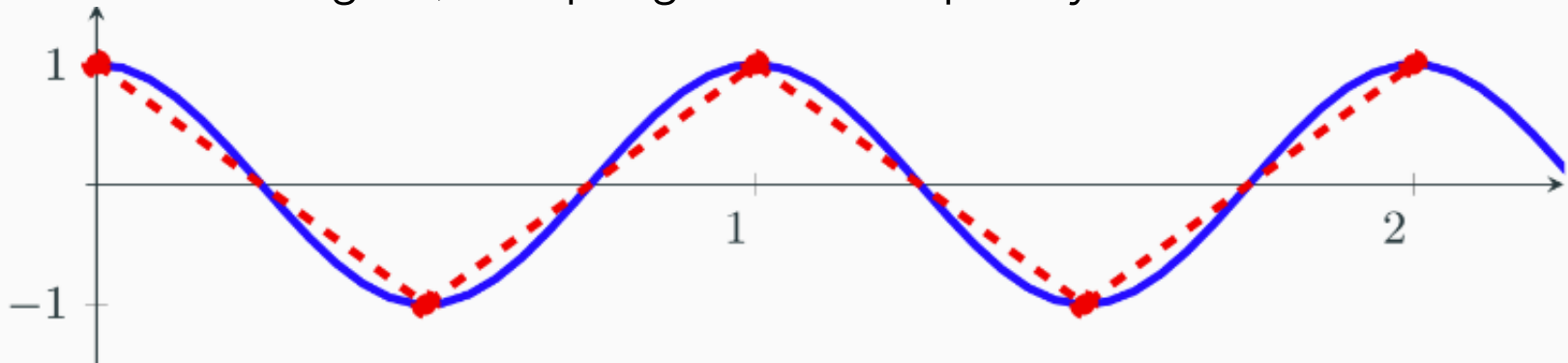


# Nyquist-Shannon sampling theorem

Original signal, analog  $\omega = 1\text{Hz}$

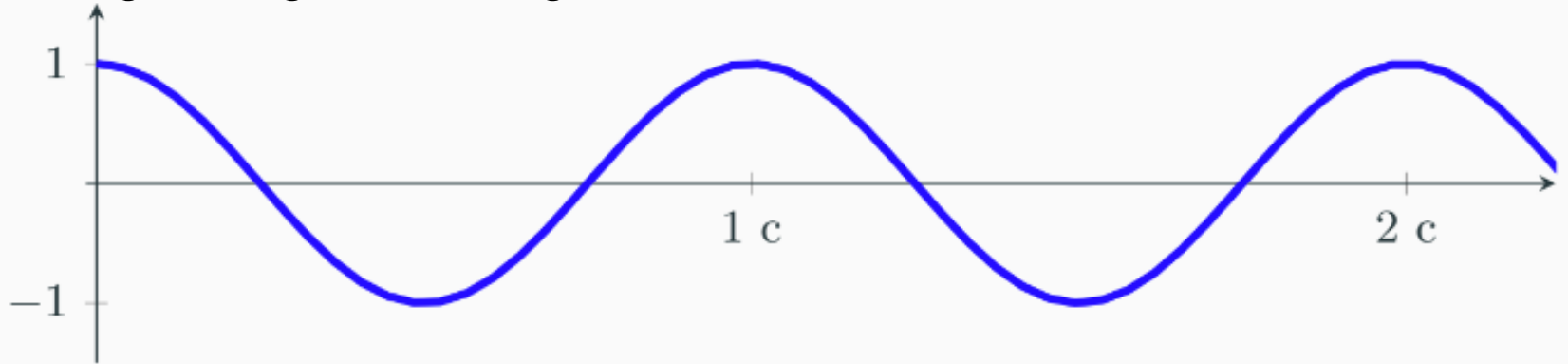


Discretised signal, sampling **two** times per cycle

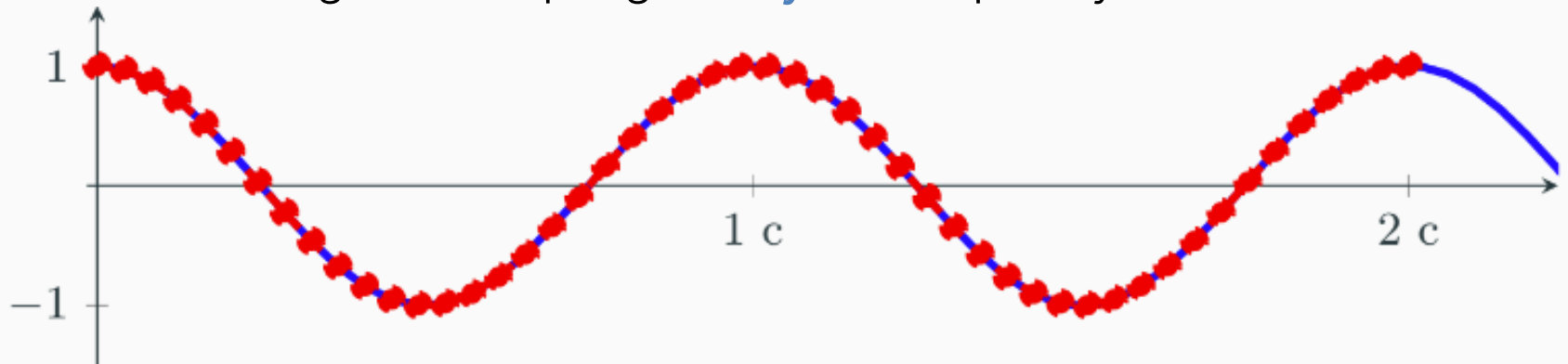


# Nyquist-Shannon sampling theorem

Original signal, analog  $\omega = 1\text{Hz}$



Discretised signal, sampling 'many' times per cycle







# Nyquist-Shannon sampling theorem

A digital signal has a fixed sampling time  $T_s$  per cycle.



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**Caveat:** there is an approximation error as a result of discretization (loss of information)



# Nyquist-Shannon sampling theorem

A digital signal has a fixed sampling time  $T_s$  per cycle.

**Caveat:** there is an approximation error as a result of discretization (loss of information)

Then...

How do we know what sample rate (no of samples per cycle) to use?



# Nyquist-Shannon sampling theorem

## Nyquist-Shannon Theorem

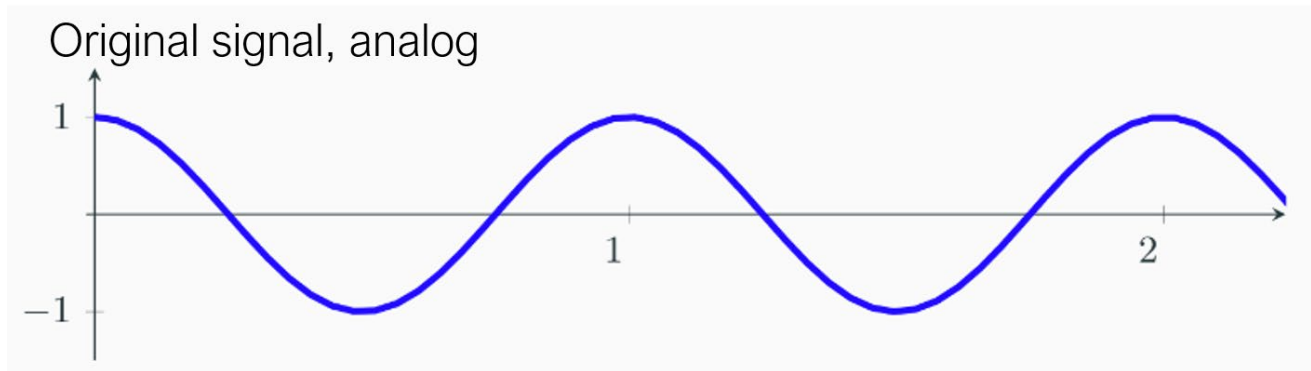
*If a function  $x(t)$  contains no frequencies higher than  $\omega$  Hertz, it is completely determined by giving its ordinates as a series of points spaced  $\frac{1}{2\omega}$  seconds apart.*

*A sufficient sample rate is  $2\omega$  samples/second,  
where  $\omega$  frequency of the signal*



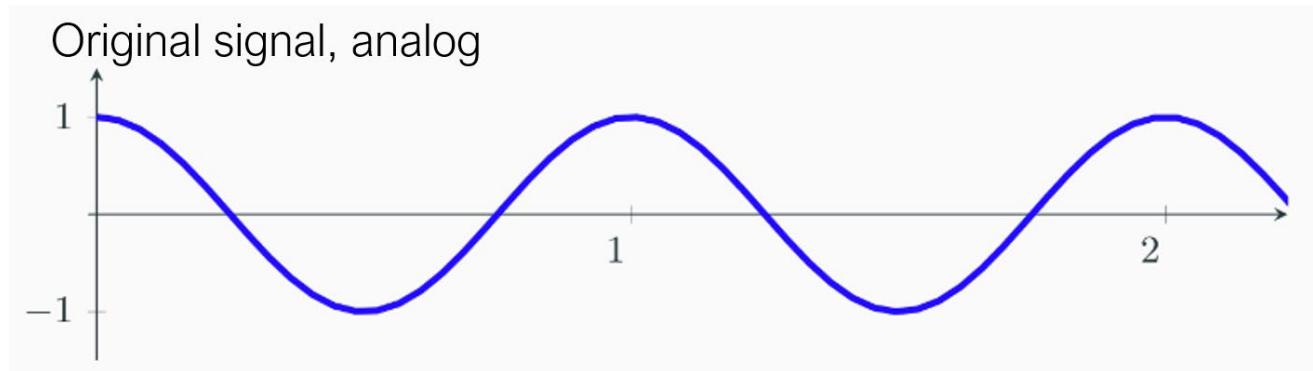
# Nyquist-Shannon sampling theorem

For the example signal with  $\omega = 1\text{Hz}$

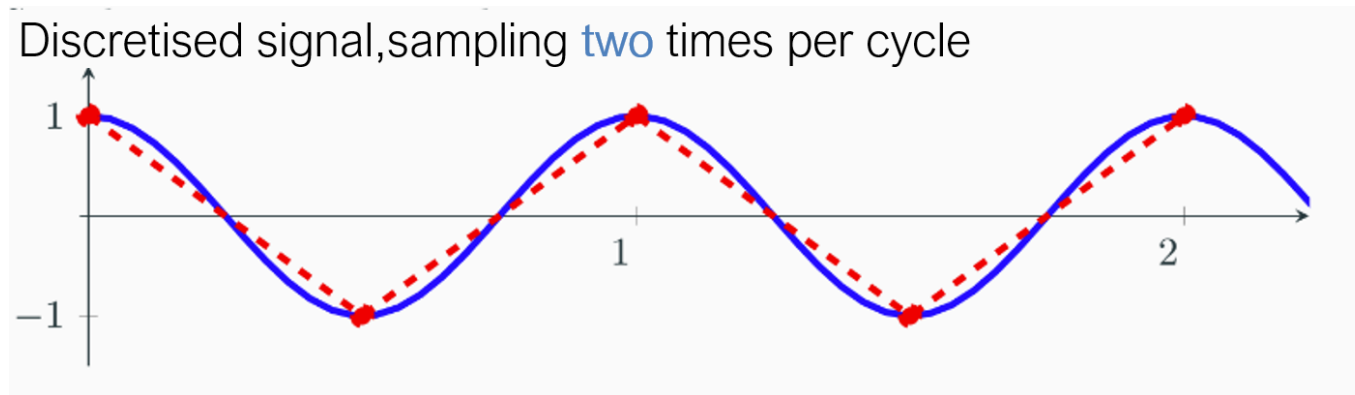


# Nyquist-Shannon sampling theorem

For the example signal with  $\omega = 1\text{Hz}$



A sampling of  $2\omega = 2 \text{ samples/second}$  is deemed *sufficient*





# z-transform



# z-transform

- Similar to Laplace transform for continuous-time signals, for discrete-time framework, we have the **z-transform**





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- z-transform is an important tool to study **discrete time** signals and systems

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- Similar to Laplace transform for continuous-time signals, for discrete-time framework we have the **z-transform**
- z-transform is an important tool to study **discrete time** signals and systems

Given  $x(k) \in \mathbb{R}$ ,  $k = 0, 1, \dots, \infty$ , assuming  $x(k) = 0$  for all  $k < 0$

**Definition (z-transform of  $x(k)$ )**

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

where  $z \in \mathbb{C}$  is the independent variable.

# z-transform

It is possible to relate **Laplace** and **z transforms** with the following methods:

- $z = e^{sT}$  (Impulse Invariant Method)
- $s = \frac{2(z-1)}{T(z+1)}$  (Bilinear Transform Method)

# Example 1: Infinite sequence

Let  $x(k + 1) = ax(k)$  with  $a \in \mathbb{R}$ . This has solution:  $x(k) = a^k x(0)$

Therefore, we have the sequence

$$\{x(k)\} = \{x(0), ax(0), a^2x(0) \dots\},$$

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Then, **the z-transform** of  $x(k)$

$$X(z) = \sum_{k=0}^{\infty} x(k)z^{-k} = x(0) \sum_{k=0}^{\infty} a^k z^{-k} = x(0) \sum_{k=0}^{\infty} \left(\frac{a}{z}\right)^k,$$

which follows the **geometric series formula**

## Recall geometric series formula:

For a sequence  $c_k = c^k$  with  $|c| < 1$ . The sum of the elements in the sequence is

$$\sum_{k=0}^{\infty} c^k = \frac{1}{1-c} \text{ for } |c| < 1$$

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$$X(z) = x(0) \frac{1}{1 - \frac{a}{z}} \text{ when } \left|\frac{a}{z}\right| < 1 \Leftrightarrow X(z) = x(0) \frac{z}{z-a} \text{ when } |z| > |a|.$$

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$|z| > |a|$  is called the **Region of Convergence (ROC)** of the z-transform



# Example 2: Finite sequence

Consider  $\{x(k)\} = \{a, b, c, 0, 0, 0 \dots\}$ , with  $a, b, c \in \mathbb{R}$





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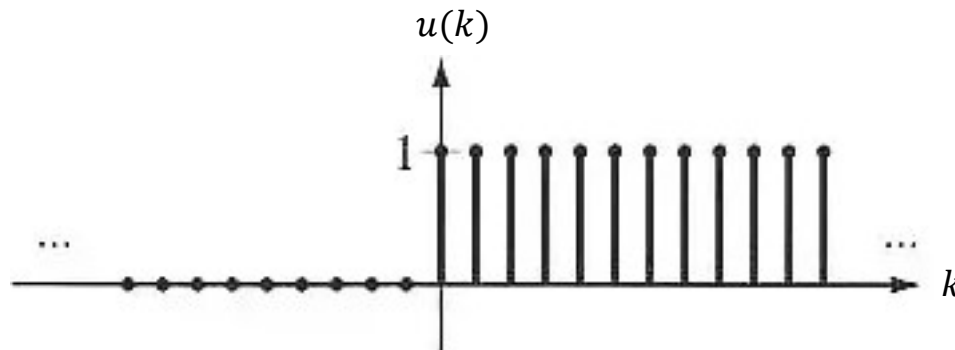
Then, the z-transform of  $x(k)$

$$X(z) = \sum_{k=0}^{\infty} x(k)z^{-k} = x(0)z^{-0} + x(1)z^{-1} + x(2)z^{-2} = a + bz^{-1} + cz^{-2}$$

# Example 3: Unit Step

Consider **unit step function**

$$u(k) = \begin{cases} 0 & k < 0 \\ 1 & k \geq 0 \end{cases}$$





# Example 3: Unit Step

Consider **unit step function**

$$u(k) = \begin{cases} 0 & k < 0 \\ 1 & k \geq 0 \end{cases}$$

Then, **the z-transform** of  $u(k)$

$$U(z) = \sum_{k=0}^{\infty} u(k)z^{-k} = \sum_{k=0}^{\infty} 1z^{-k} = \frac{1}{1 - z^{-1}} = \frac{1}{z - 1}$$

Geometric series

when  $|z| > 1$



# z-transform pairs

$x(k), k \geq 0$	z-Transform $X(z)$	Region of Convergence
$x(k)$	$\sum_{k=0}^{\infty} x(k)z^{-k}$	
$\delta(k)$	1	$ z  > 0$
$au(k)$	$\frac{az}{z-1}$	$ z  > 1$
$ku(k)$	$\frac{z}{(z-1)^2}$	$ z  > 1$
$k^2u(k)$	$\frac{z(z+1)}{(z-1)^3}$	$ z  > 1$
$a^k u(k)$	$\frac{z}{z-a}$	$ z  >  a $
$e^{-ak} u(k)$	$\frac{z}{(z-e^{-a})}$	$ z  > e^{-a}$
$ka^k u(k)$	$\frac{az}{(z-a)^2}$	$ z  >  a $
$\sin(ak)u(k)$	$\frac{z \sin(a)}{z^2 - 2z \cos(a) + 1}$	$ z  > 1$
$\cos(ak)u(k)$	$\frac{z[z - \cos(a)]}{z^2 - 2z \cos(a) + 1}$	$ z  > 1$
$a^k \sin(bk)u(k)$	$\frac{[a \sin(b)]z}{z^2 - [2a \cos(b)]z + a^2}$	$ z  >  a $
$a^k \cos(bk)u(k)$	$\frac{[z - a \cos(b)]z}{z^2 - [2a \cos(b)]z + a^2}$	$ z  >  a $
$e^{-ak} \sin(bk)u(k)$	$\frac{[e^{-a} \sin(b)]z}{z^2 - [2e^{-a} \cos(b)]z + e^{-2a}}$	$ z  > e^{-a}$
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# z-transform properties

## Linearity

$$\mathcal{Z}(a_1 x_1(k) + a_2 x_2(k)) = a_1 X_1(z) + a_2 X_2(z)$$

## Time shift

$$\mathcal{Z}(x(k - k_0)) = z^{-k_0} X(z)$$

## Time reversal

$$\mathcal{Z}(x(-k)) = X(z^{-1})$$

## First difference

$$\mathcal{Z}(x(k) - x(k - 1)) = (1 - z^{-1})X(z)$$



# z-transform properties

$x(n), n \geq 0$	Sequence	Transform	Region of Convergence
	$x[k]$ $x_1[k]$ $x_2[k]$	$X(z)$ $X_1(z)$ $X_2(z)$	$R$ $R_1$ $R_2$
Linearity	$(a_1 x_1(k) + a_2 x_2(k))$	$a_1 X_1(z) + a_2 X_2(z)$	At least intersection of $R_1$ and $R_2$
Time shifting	$x[k - k_0]$	$z^{-n_0} X(z)$	$R$ except possible addition or deletion of origin
Scaling in z-domain	$e^{j\omega_0 k} x[k]$ $z_0^k x[k]$ $a^k x[k]$	$X(e^{-j\omega_0} z)$ $X\left(\frac{z}{z_0}\right)$ $X(a^{-1} z)$	$R$ $z_0 R$ Scaled version of $R$
Time reversal	$x[-k]$	$X(z^{-1})$	Inverted $R$
Conjugation	$x^*[k]$	$X^*(z^*)$	$R$
Convolution	$x_1[k] * x_2[k]$	$X_1(z) X_2(z)$	At least intersection of $R_1$ and $R_2$
First difference	$x[k] - x[k - 1]$	$(1 - z^{-1}) X(z)$	At least intersection of $R$ and $ z  > 0$
Accumulation	$\sum_{k=-\infty}^n x[k]$	$\frac{1}{1 - z^{-1}} X(z)$	At least intersection of $R$ and $ z  > 1$
Differentiation in z-domain	$kx[k]$	$-z \frac{dX(z)}{dz}$	$R$



# Transfer functions

Consider the  $n$ -th order difference equation

$$\begin{aligned} y(k) + a_1 y(k-1) + \dots + a_n y(k-n) \\ = b_0 u(k) + b_1 u(k-1) + \dots + b_n u(k-n) \end{aligned}$$



# Transfer functions

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Applying  $z$ -transformation

$$(1 + a_1 z^{-1} + \dots + a_n z^{-n})Y(z) = (b_0 + b_1 z^{-1} + \dots + b_n z^{-n})U(z)$$



# Transfer functions

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Applying  $z$ -transformation

$$(1 + a_1 z^{-1} + \dots + a_n z^{-n})Y(z) = (b_0 + b_1 z^{-1} + \dots + b_n z^{-n})U(z)$$

Then, the transfer function

$$G(z) = \frac{Y(z)}{U(z)} = \frac{b_0 + b_1 z^{-1} + \dots + b_n z^{-n}}{1 + a_1 z^{-1} + \dots + a_n z^{-n}} = \frac{b_0 z^n + b_1 z^{n-1} + \dots + b_n}{z^n + a_1 z^{n-1} + \dots + a_n}$$



# State space representation and transfer functions

Just as in continuous time

$$x(k + 1) = Ax(k) + Bu(k)$$

$$y(k) = Cx(k) + Du(k)$$

where  $x(k) \in \mathbb{R}^n$ ,  $u(k) \in \mathbb{R}^p$ ,  $y(k) \in \mathbb{R}^m$  is a state space representation of an n-th order difference equation



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Applying **z-transformation**

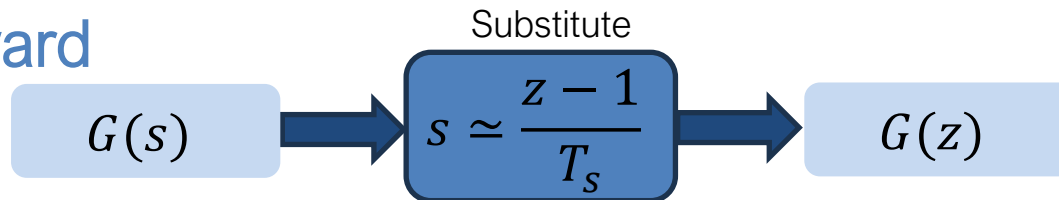
$$\mathbf{G}(z) = \frac{\mathbf{Y}(z)}{\mathbf{U}(z)} = \mathbf{C}(z\mathbf{I} - \mathbf{A})^{-1}\mathbf{B} + \mathbf{D}$$



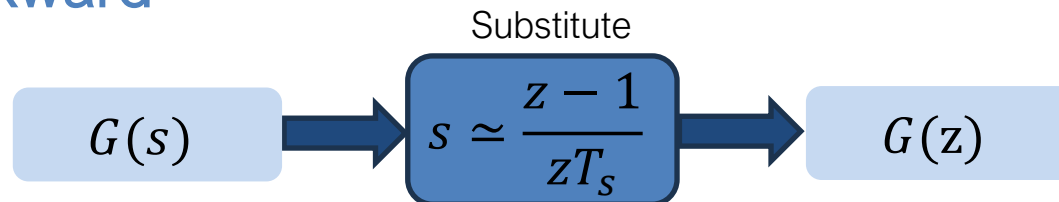
# From continuous to discrete transfer function

Depending on the approximation method, there are **several ways to relate  $s$  and  $z$** . The most important ones are

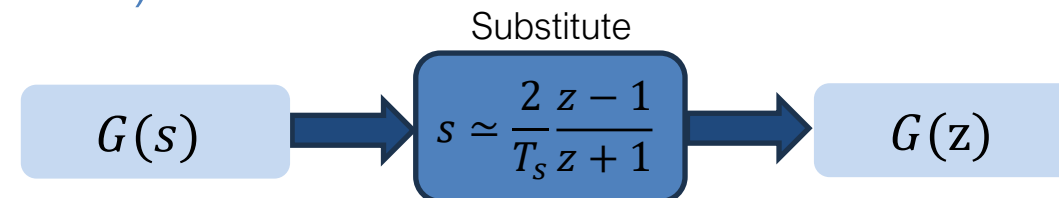
## Euler forward



## Euler backward



## Bilinear (Tustin)

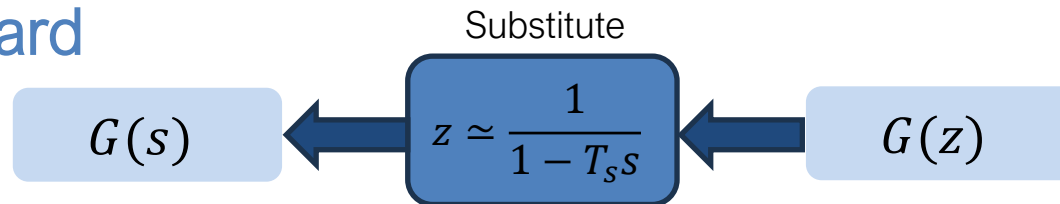




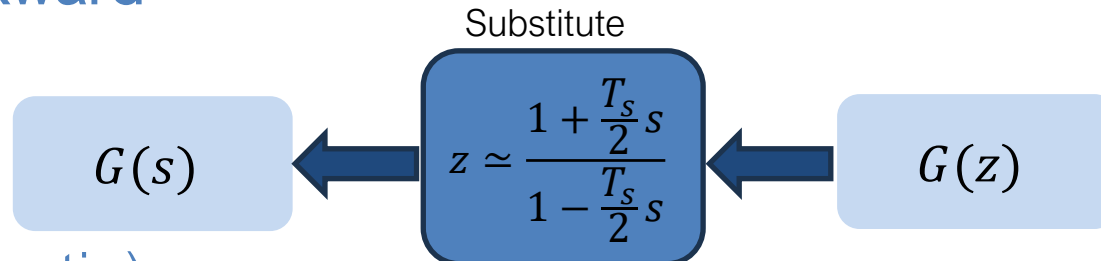
# From discrete to continuous transfer function

The [reverse transformations](#) to go from  $z$  to  $s$  are shown below for the most important methods:

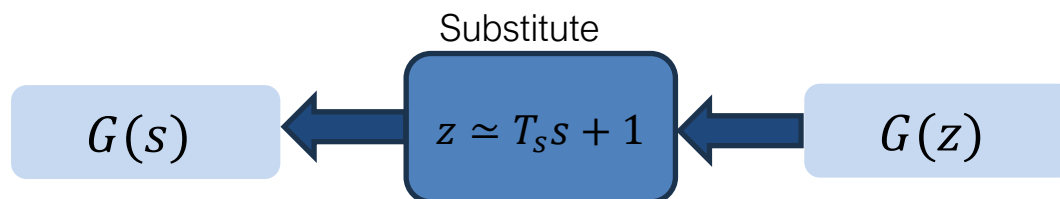
## Euler forward



## Euler backward



## Bilinear (Tustin)



# Example

- Consider the system

$$\frac{dx(t)}{dt} = \underbrace{\begin{bmatrix} -1 & 3 \\ 0 & -2 \end{bmatrix}}_A \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} + \underbrace{\begin{bmatrix} 0 \\ 1 \end{bmatrix}}_B u(t)$$

$$y(t) = \underbrace{\begin{bmatrix} 1 & 0 \end{bmatrix}}_C \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix}, D = 0$$

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- Can we use  $C(z) = C(zI - A)^{-1}B + D$  to obtain discrete time TF?

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- Can we use  $C(z) = C(zI - A)^{-1}B + D$  to obtain a discrete time TF?
- NO!** It's a continuous-time system represented by a differential equation, not a discrete-time system represented by a difference



# Example

- Consider the system

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$$y(t) = \underbrace{\begin{bmatrix} 1 & 0 \end{bmatrix}}_C \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix}, D = 0$$

- Can we use  $C(z) = C(zI - A)^{-1}B + D$  to obtain a discrete time TF?
- NO!** It's a continuous-time system represented by a differential equation, not a discrete-time system represented by a difference
- We instead need to use  $C(s) = C(sI - A)^{-1}B + D$

# Example

- It follows that

$$\begin{aligned} C(s) &= C(sI - A)^{-1}B + D = [1 \quad 0] \begin{bmatrix} s+1 & -3 \\ 0 & s+2 \end{bmatrix}^{-1} \begin{bmatrix} 0 \\ 1 \end{bmatrix} + 0 \\ &= [1 \quad 0] \frac{1}{(s+1)(s+2)} \begin{bmatrix} s+1 & 3 \\ 0 & s+2 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \frac{3}{(s+1)(s+2)} \end{aligned}$$

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- Using Euler forward approximation (for example),  $s \simeq \frac{z-1}{T_s}$ , we get the transfer function in  $z$ :

$$\mathbf{G}(z) = \frac{3T_s^2}{(z + T_s - 1)(z + 2T_s - 1)}$$



# Summary

- Discretization plays important role in information processing part of a Mechatronics system
- Nyquist-Shannon sampling theorem establishes necessary and sufficient conditions to select adequate sample rate
- $z$ -transform, analogous to Laplace, is useful to obtain transfer function of a discrete-time system



# Next lecture:

## Discrete-time systems and control