

Digital Control
or
The really useful control stuff

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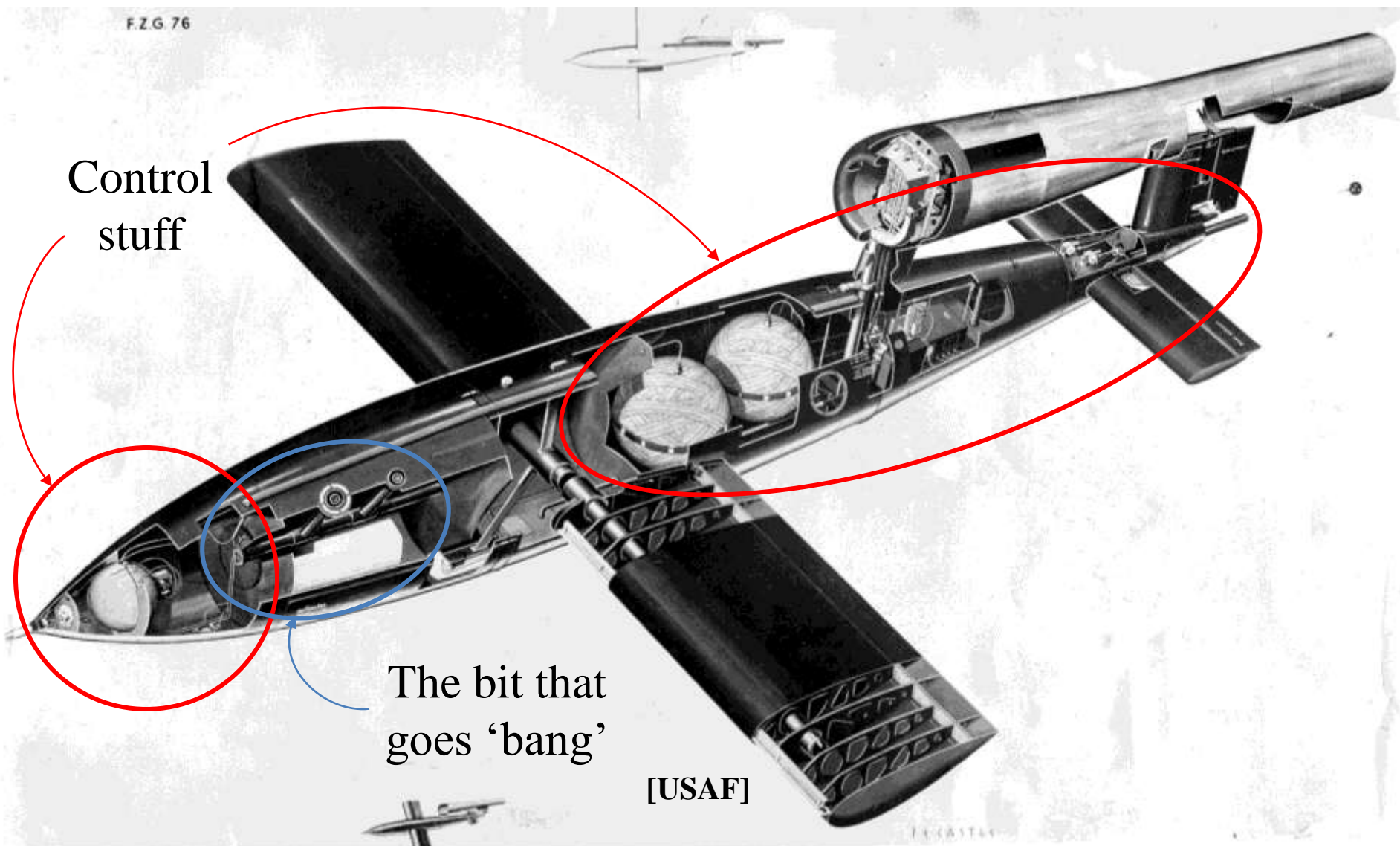
Digital control lolwut?

Once upon a time...

- Electromechanical systems were controlled by electromechanical compensators
 - Mechanical flywheel governors, capacitors, inductors, resistors, relays, valves, solenoids (fun!)
 - But also complex and sensitive!
- Humans developed sophisticated tools for designing reliable analog controllers

Eg. Early UAV flight control

F.Z.G. 76



Control stuff

The bit that goes 'bang'

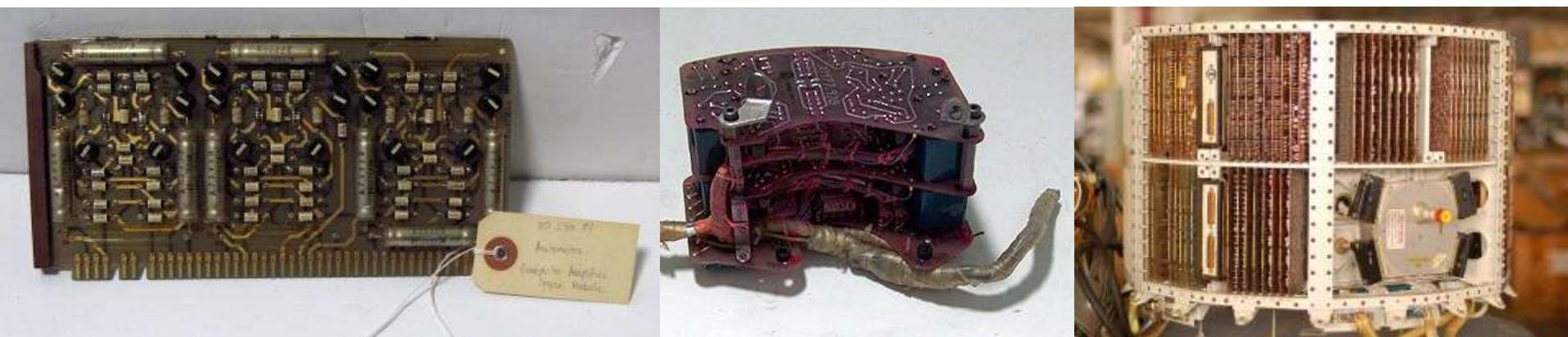
[USAF]

Eg. missile guidance



Computer revolution

- In the 1950s and 60s very smart people developed computerised controllers
- Digital processor implements the control algorithm numerically, rather than in discrete hardware



Many advantages

- Practical improvement over analog control:
 - **Flexible**; reprogrammable to implement different control laws for different systems
 - **Adaptable**; control algorithms can be changed on-line, during operation
 - **Insensitive** to environmental conditions;
(heat, EMI, vibration, etc)
 - **Compact**; handful of components on a PCB
 - **Cheap**

Ok, so how do we do this?

We already know about control, right?

What you already know*

- Signals can be represented by transfer functions in the s-domain
- Roots of a transfer function's denominator (poles) indicate the stability of the system
- Poles move around under feedback control
 - Feedback can stabilise an unstable system

*If you have no idea what I'm talking about, now is the time to mention it.

The good news

Digital control is just like that!

Thanks for coming, see you at the exam!

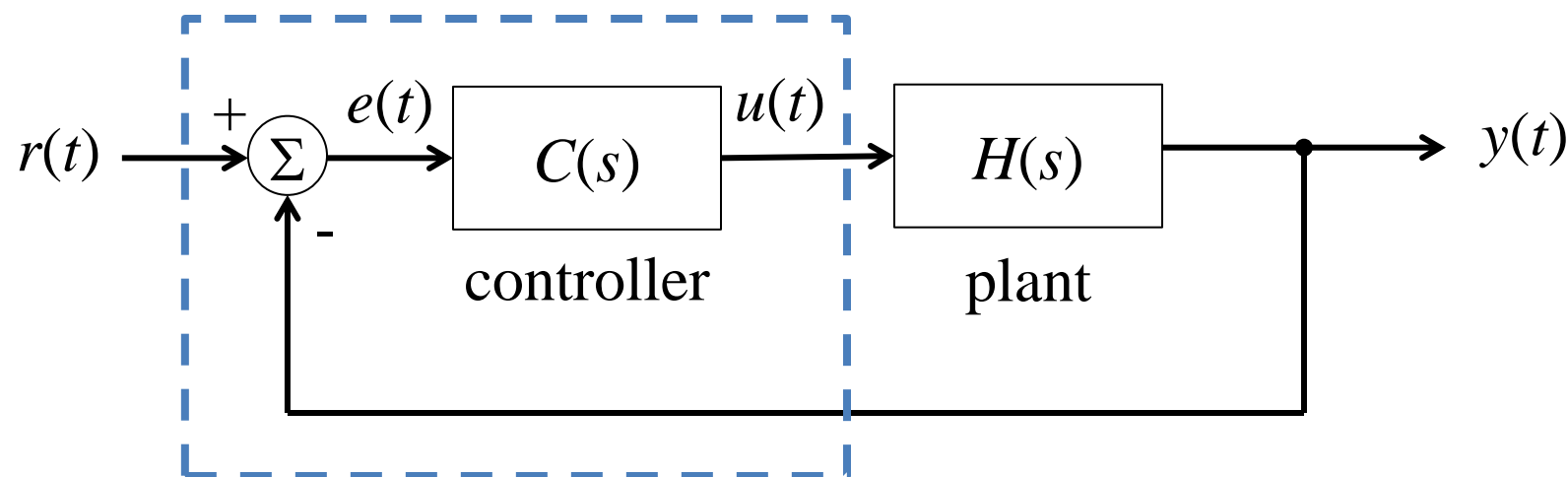
Not so fast

- While there are discrete analogues for every part of continuous systems theory, there are unique and important differences you must be familiar with

Virtually every control system you will ever use will be a computerised digital controller

Archetypical control system

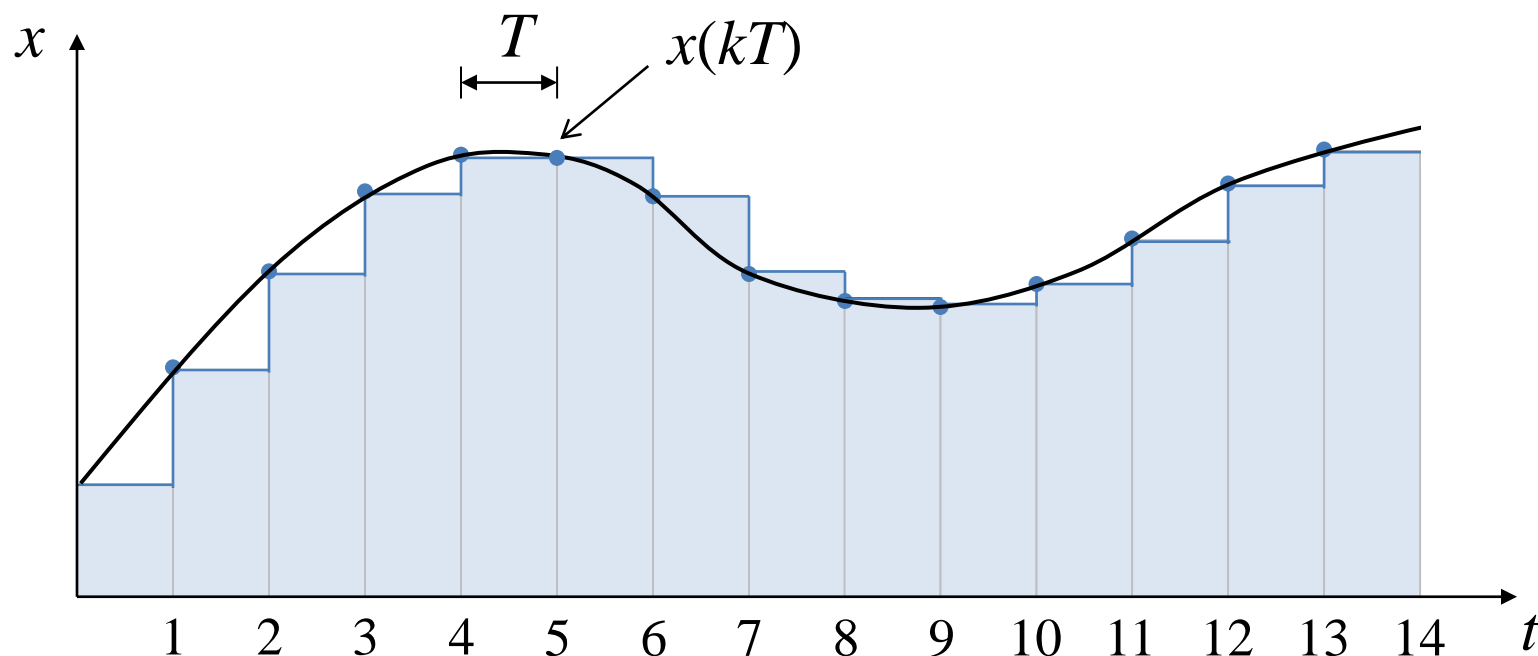
- Consider a continuous control system:



- The functions of the controller can be entirely represented by a discretised computer system

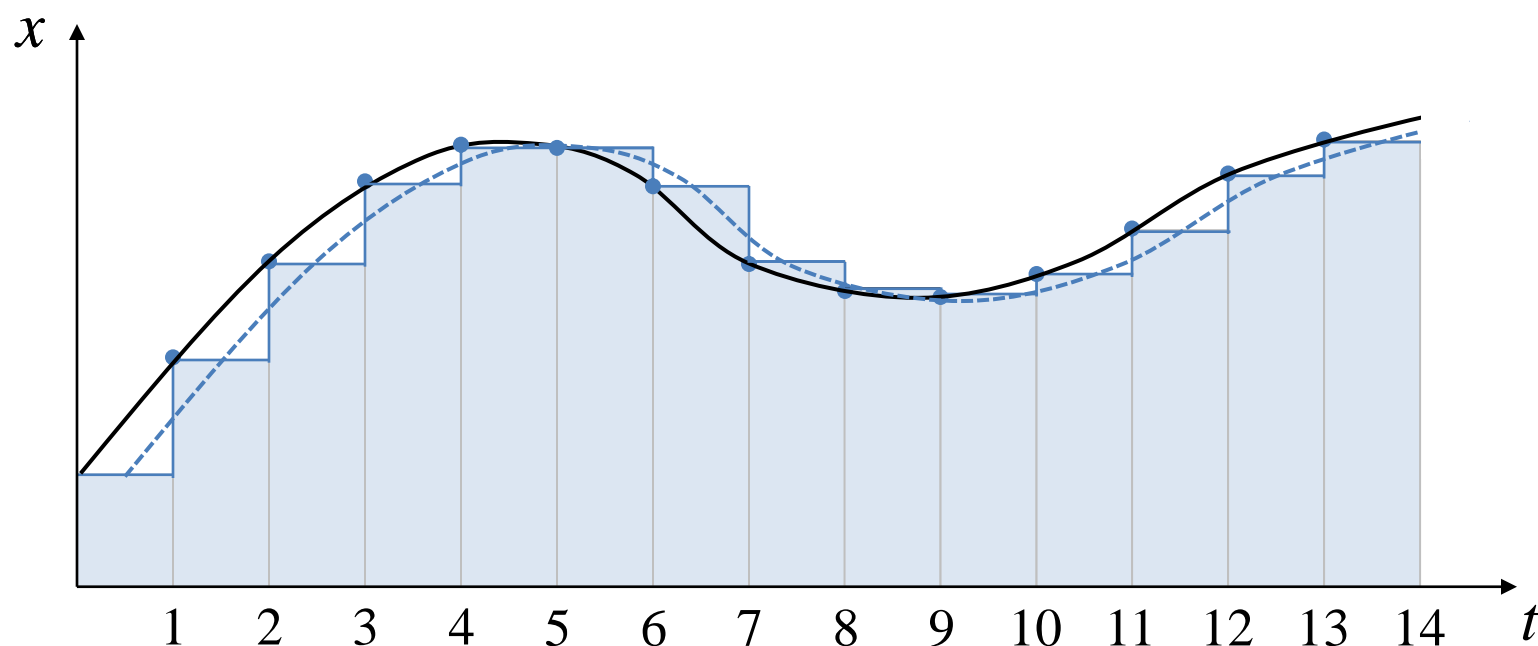
Return to the discrete domain

- Recall that continuous signals can be represented by a series of samples with period T



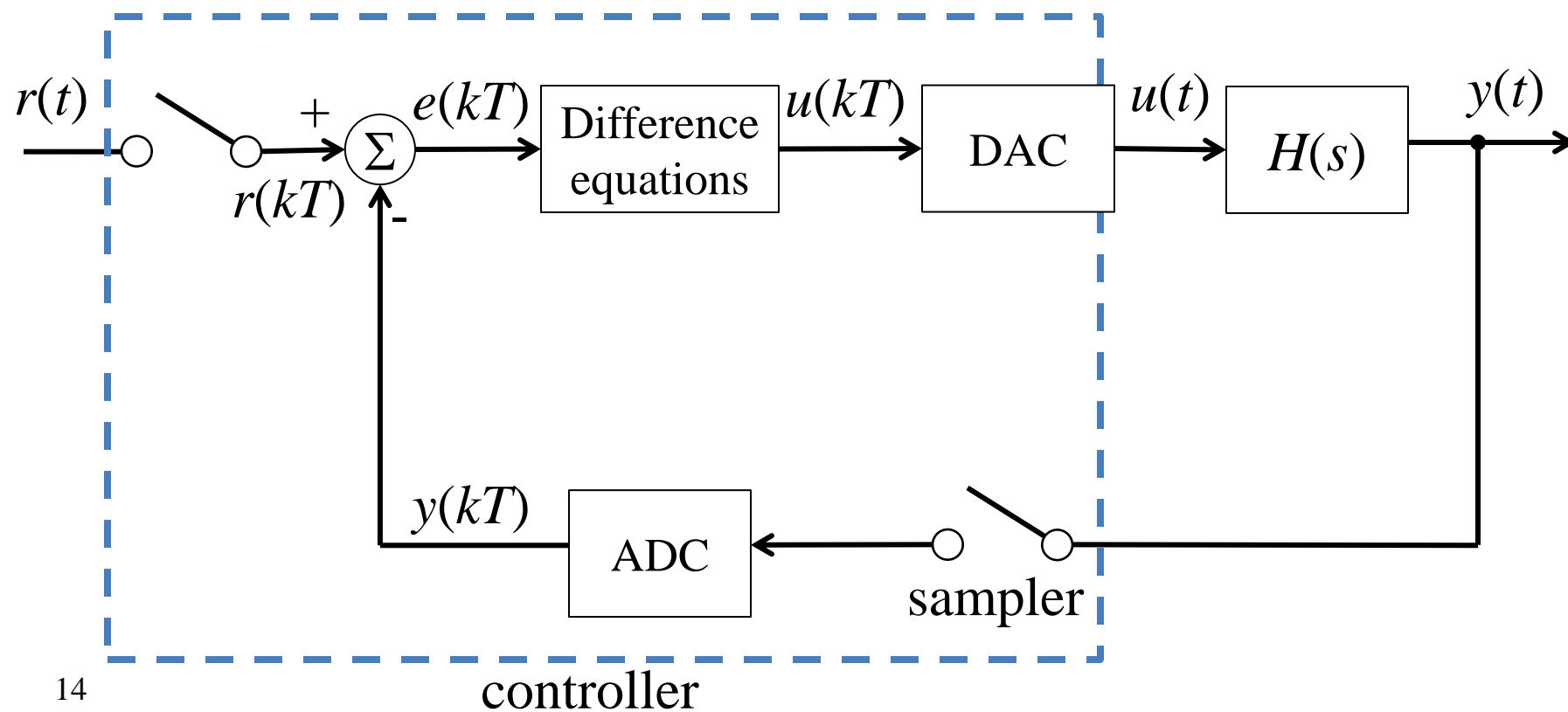
Zero Order Hold

- An output value of a synthesised signal is held constant until the next value is ready
 - This introduces an effective delay of $T/2$



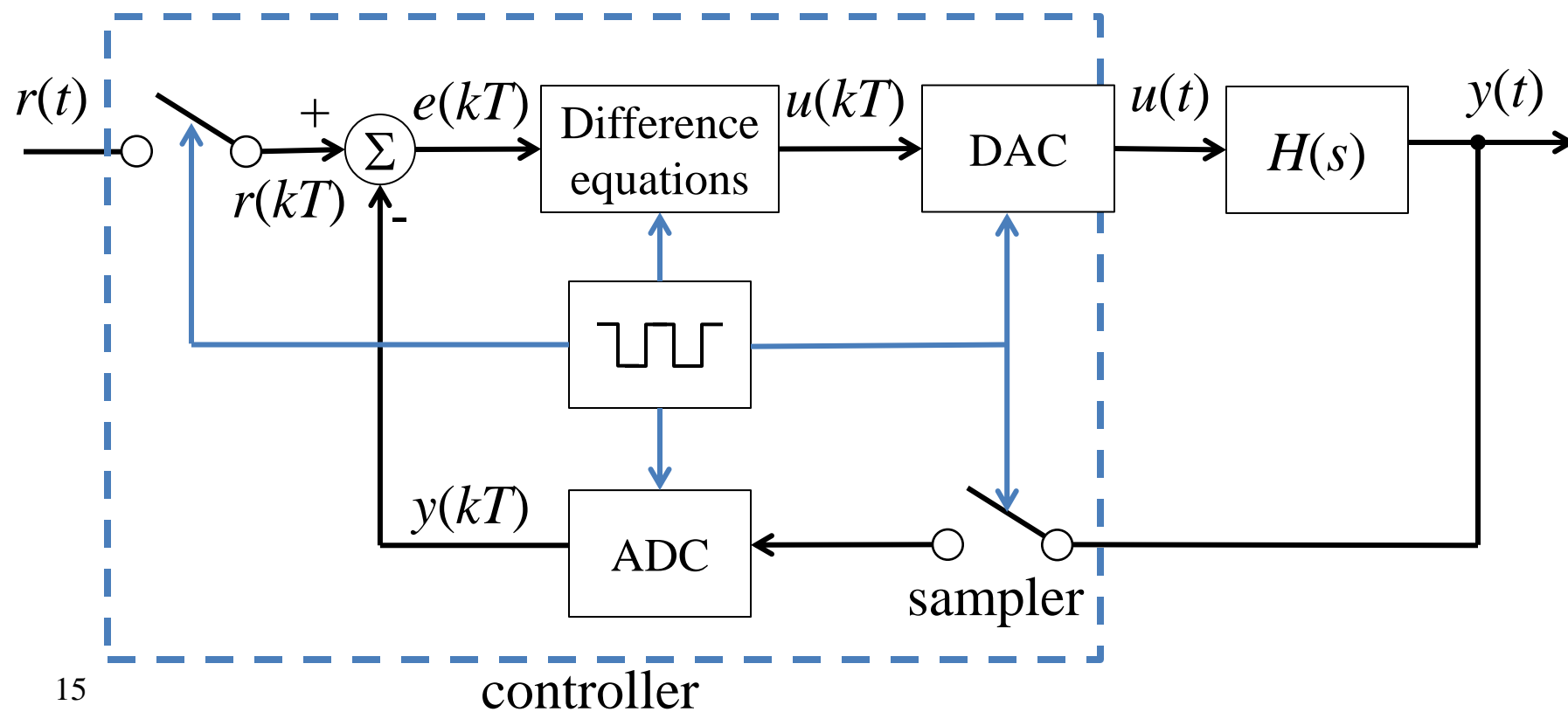
Digitisation

- Continuous signals sampled with period T
- k th control value computed at $t_k = kT$



Digitisation

- Continuous signals sampled with period T
- k th control value computed at $t_k = kT$



Difference equations

- How to represent differential equations in a computer? Difference equations!
- The output of a difference equation system is a function of current and previous values of the input and output:

$$y(t_k) = D(x(t_k), x(t_{k-1}), \dots, x(t_{k-n}), y(t_{k-1}), \dots, y(t_{k-n}))$$

- We can think of x and y as parameterised in k

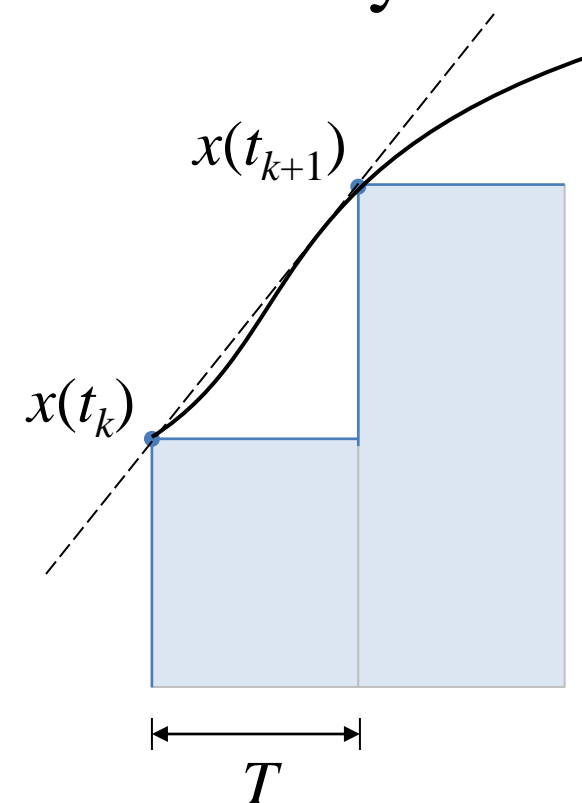
Useful shorthand: $x(t_{k+i}) \equiv x(k+i)$

Euler's method*

- Dynamic systems can be approximated[†] by recognising that:

$$\dot{x} \cong \frac{x(k+1) - x(k)}{T}$$

- As $T \rightarrow 0$, approximation error approaches 0



*Also known as the forward rectangle rule
†Just an approximation – more on this later

An example!

Convert the system $\frac{Y(s)}{X(s)} = \frac{s+2}{s+1}$ into a difference equation with period T , using Euler's method.

1. Rewrite the function as a dynamic system:

$$sY(s) + Y(s) = sX(s) + 2X(s)$$

Apply inverse Laplace transform:

$$\dot{y}(t) + y(t) = \dot{x}(t) + 2x(t)$$

2. Replace continuous signals with their sampled domain equivalents, and differentials with the approximating function

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

An example!

Simplify:

$$y(k + 1) - y(k) + Ty(k) = x(k + 1) - x(k) + 2Tx(k)$$
$$y(k + 1) + (T - 1)y(k) = x(k + 1) + (2T - 1)x(k)$$

$$y(k + 1) = x(k + 1) + (2T - 1)x(k) - (T - 1)y(k)$$

We can implement this in a computer.

Cool, let's try it!

Back to the future

A quick note on causality:

- Calculating the “(k+1)th” value of a signal using

$$y(k + 1) = \underbrace{x(k + 1)}_{\text{future value}} + \underbrace{Ax(k) - By(k)}_{\text{current values}}$$

relies on also knowing the next (future) value of $x(t)$.

(this requires very advanced technology!)

- Real systems always run with a delay:

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

Back to the example!

```
T = 0.02; //period of 50 Hz, a number pulled from thin air
A = 2*T-1; //pre-calculated control constants
B = T-1;

...

while(1)
{
    if(interrupt_flag) //this triggers every 20 ms
    {
        x0 = x; //save previous values
        y0 = y;
        x = update_input(); //get latest x value
        y = x + A*x0 - B*y0; //do the difference equations
        update_output(y); //write out current value
    }
}
```

Great!

We already know how to design compensators, and now we can implement them in a computer.

That means we're done, right?

Not quite

There are unanswered questions:

- How do we analyse more elaborate systems of these difference equation things?
- What happens as you change T ?
 - How would you even choose the right T ?
- What about Nyquist? Or noise?
- How good are those approximations anyway?

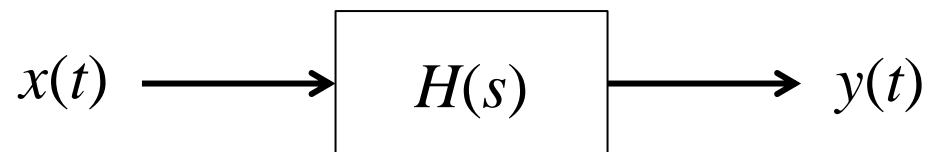
Coping with complexity

- Transfer functions help control complexity
 - Recall the Laplace transform:

$$\mathcal{L}\{f(t)\} = \int_0^{\infty} f(t)e^{-st} dt = F(s)$$

where

$$\mathcal{L}\{\dot{f}(t)\} = sF(s)$$



Is there a something similar for sampled systems?

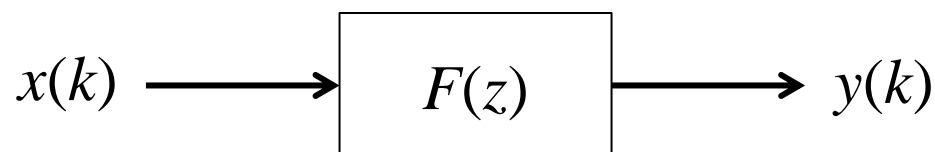
The z -transform

- The discrete equivalent is the z -Transform[†]:

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

and

$$\mathcal{Z}\{f(k-1)\} = z^{-1}F(z)$$



Convenient!

[†]This is not an approximation, but approximations are easier to derive

The z -transform

- Some useful properties
 - **Delay by n samples:** $\mathcal{Z}\{f(k - n)\} = z^{-n}F(z)$
 - **Linear:** $\mathcal{Z}\{af(k) + bg(k)\} = aF(z) + bG(z)$
 - **Convolution:** $\mathcal{Z}\{f(k) * g(k)\} = F(z)G(z)$

So, all those block diagram manipulation tools you know and love will work just the same!

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

Final value theorem

- An important question: what is the steady-state output a stable system at $t = \infty$?
 - For continuous systems, this is found by:

$$\lim_{t \rightarrow \infty} x(t) = \lim_{s \rightarrow 0} sX(s)$$

- The discrete equivalent is:

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

(Provided the system is stable)

An example!

- Back to our difference equation:

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

becomes

$$Y(z) = X(z) + Az^{-1}X(z) - Bz^{-1}Y(z)$$

$$(z + B)Y(z) = (z + A)X(z)$$

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?

Consider the simplest system

- Take a first-order response:

$$f(t) = e^{-at} \Rightarrow \mathcal{L}\{f(t)\} = \frac{1}{s + a}$$

- The discrete version is:

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

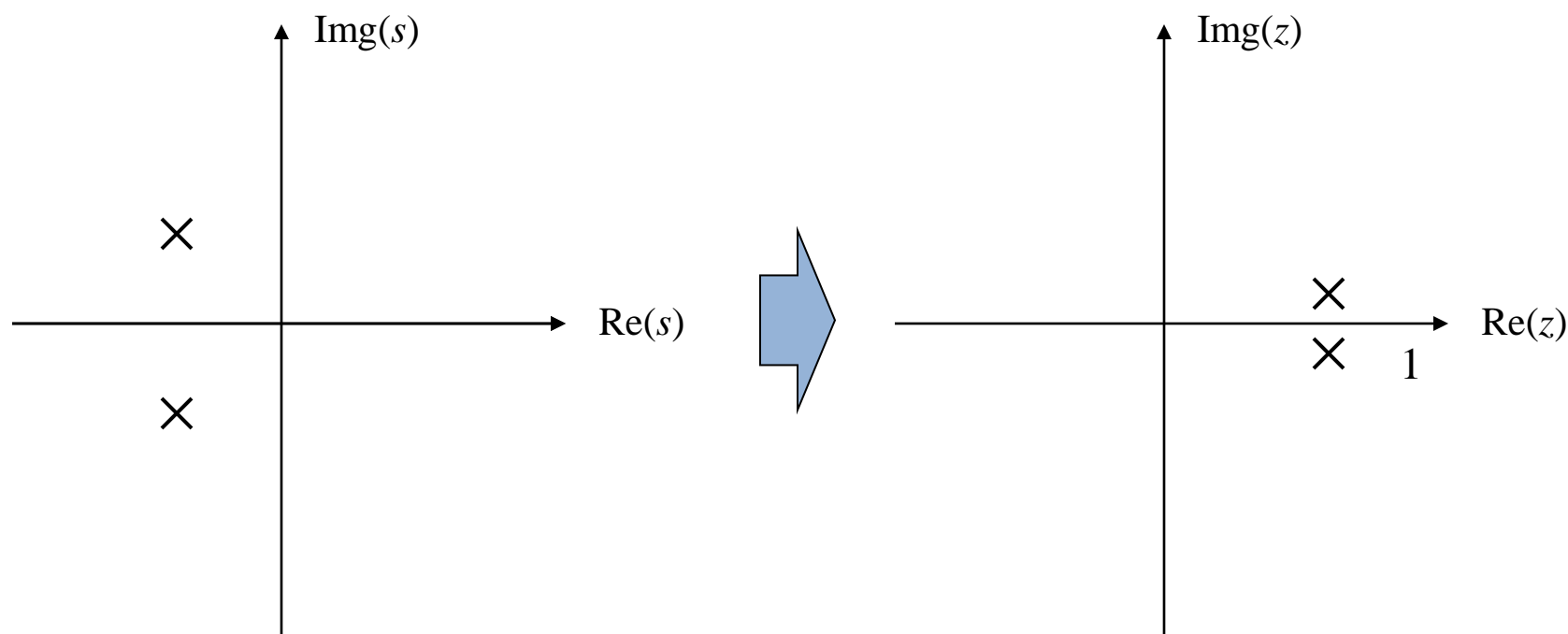
The equivalent system poles are related by

$$z = e^{sT}$$

That sounds somewhat profound... but what does it mean?

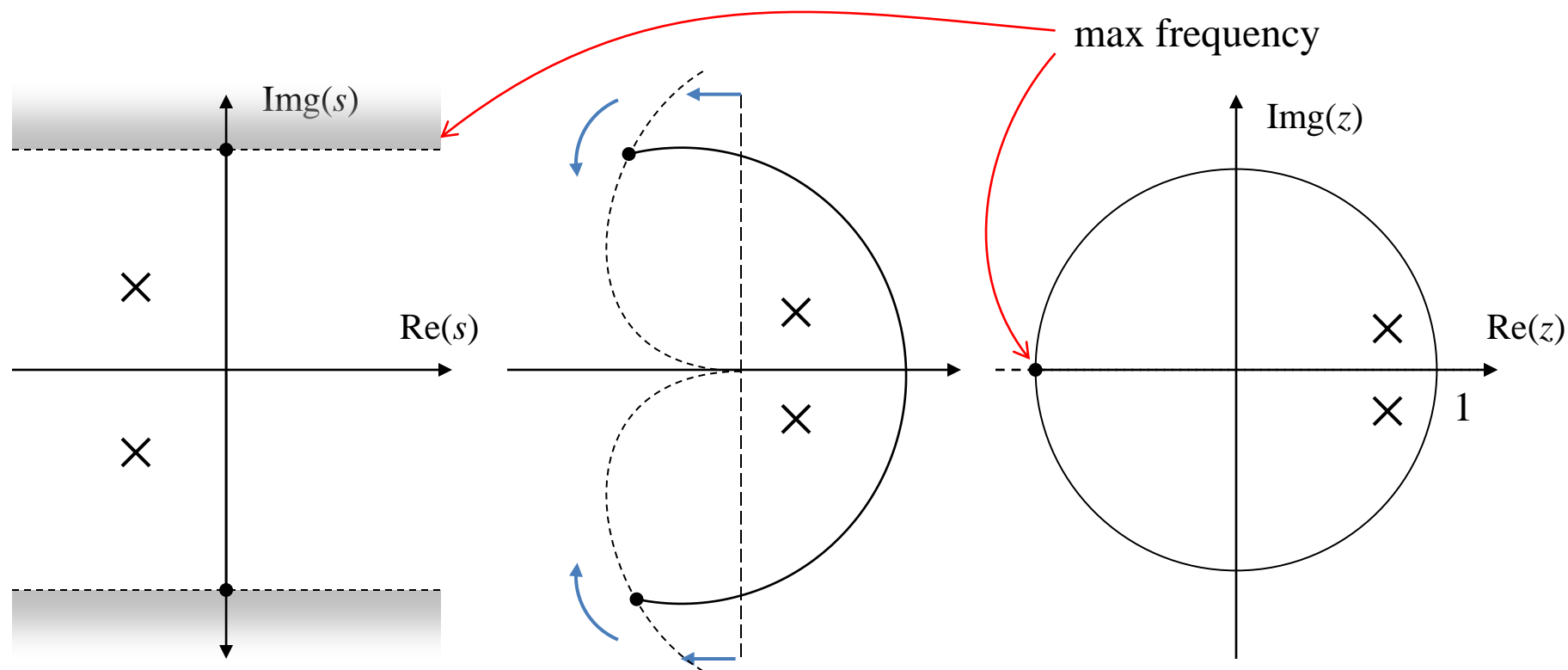
The z -Plane

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



Deep insight #1

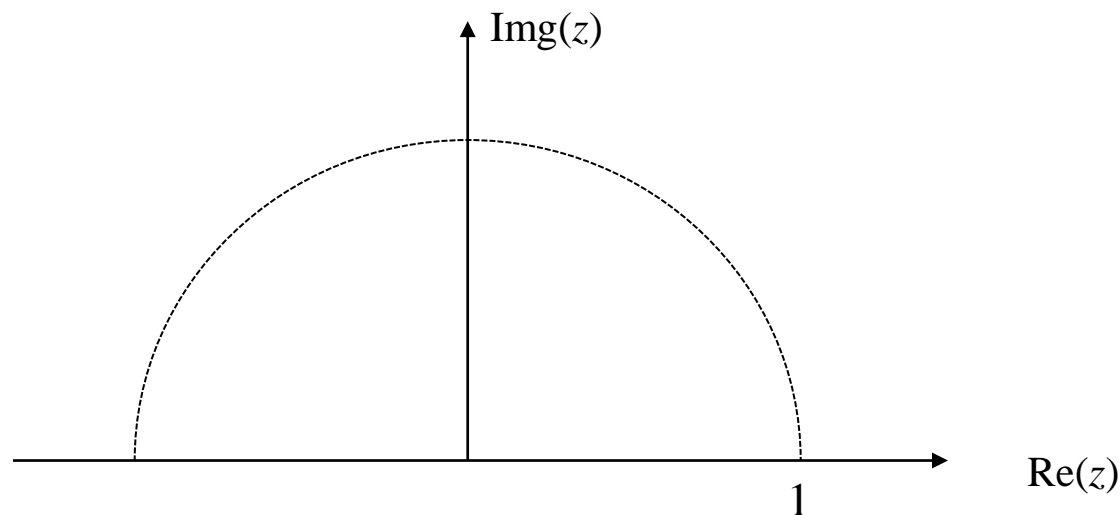
The mapping between continuous and discrete poles and zeros acts like a distortion of the plane



The z -plane

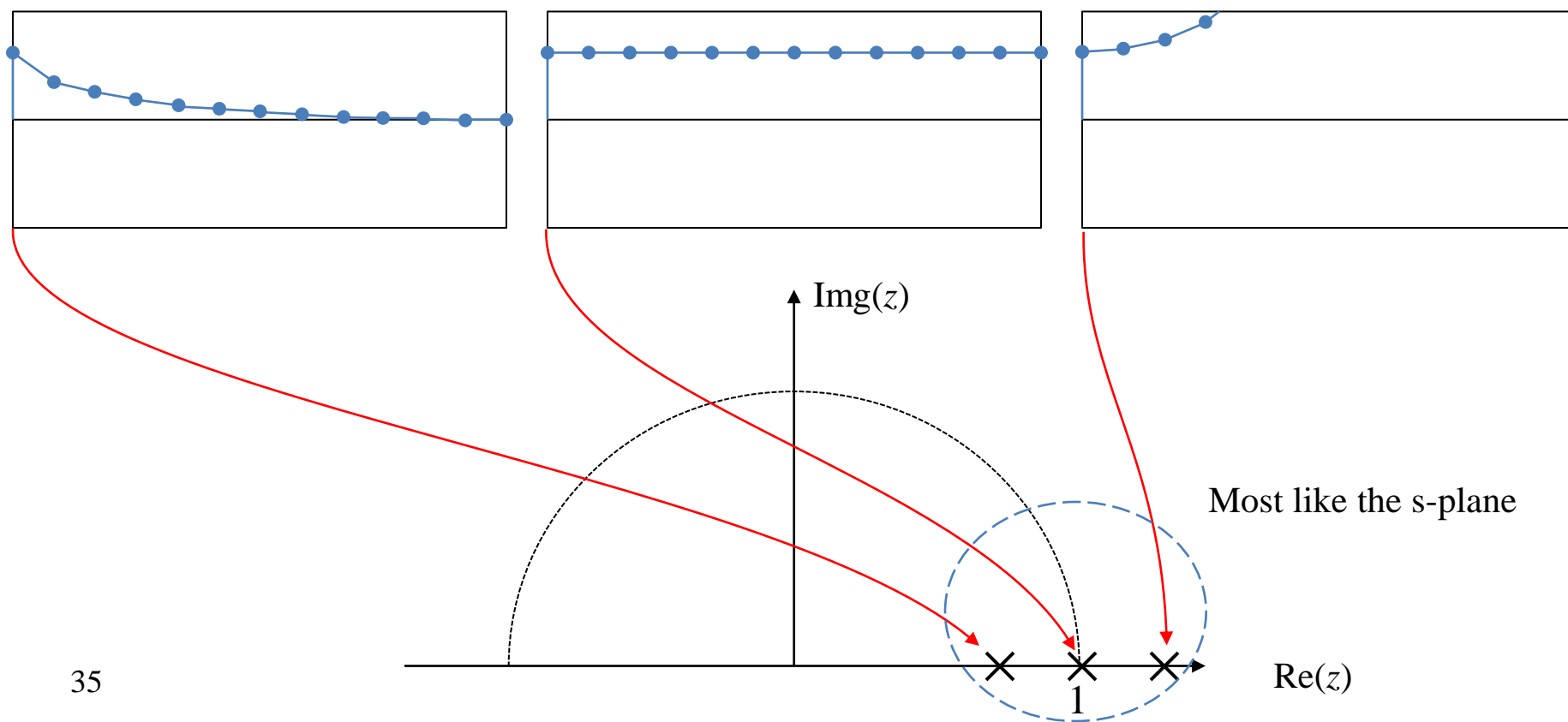
- We can understand system response by pole location in the z -plane

[Adapted from Franklin, Powell and Emami-Naeini]



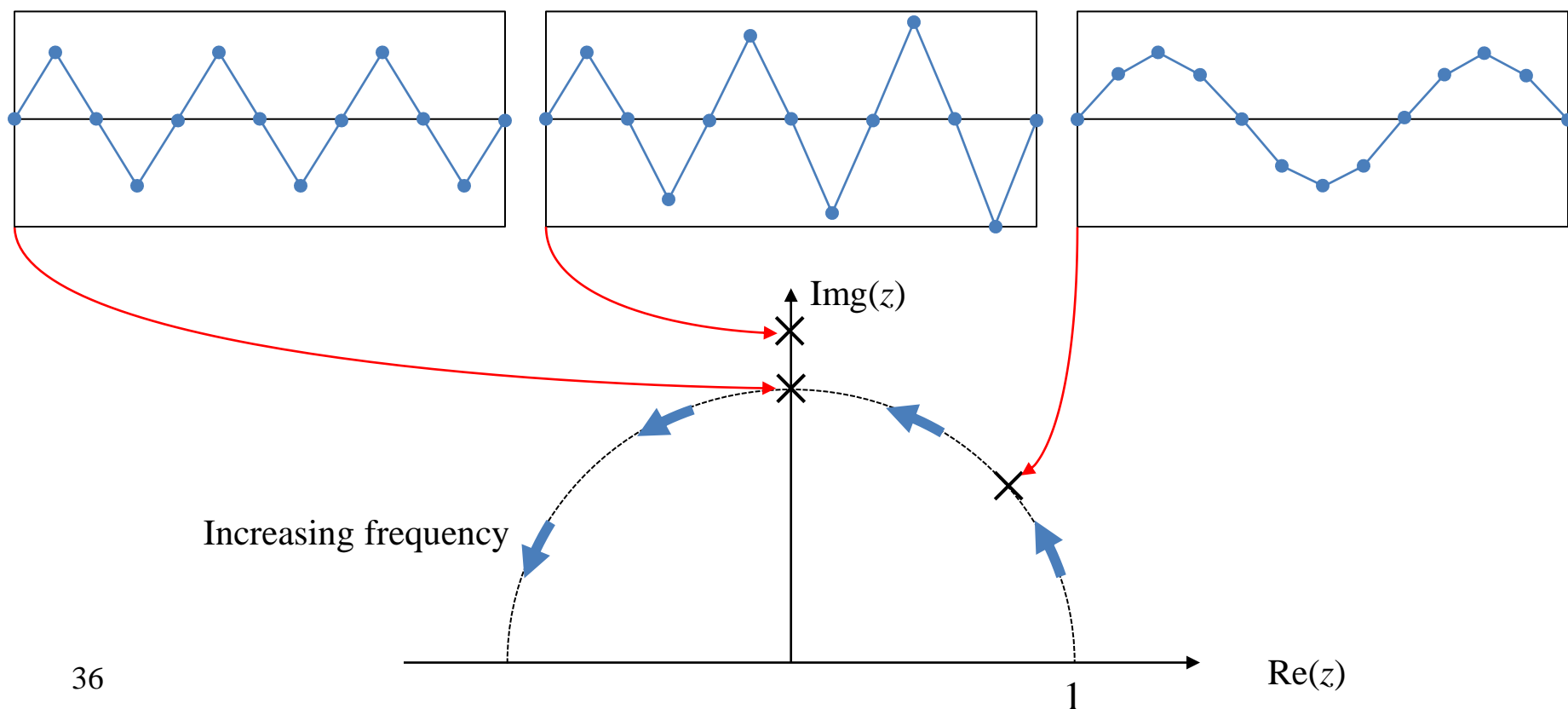
Effect of pole positions

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



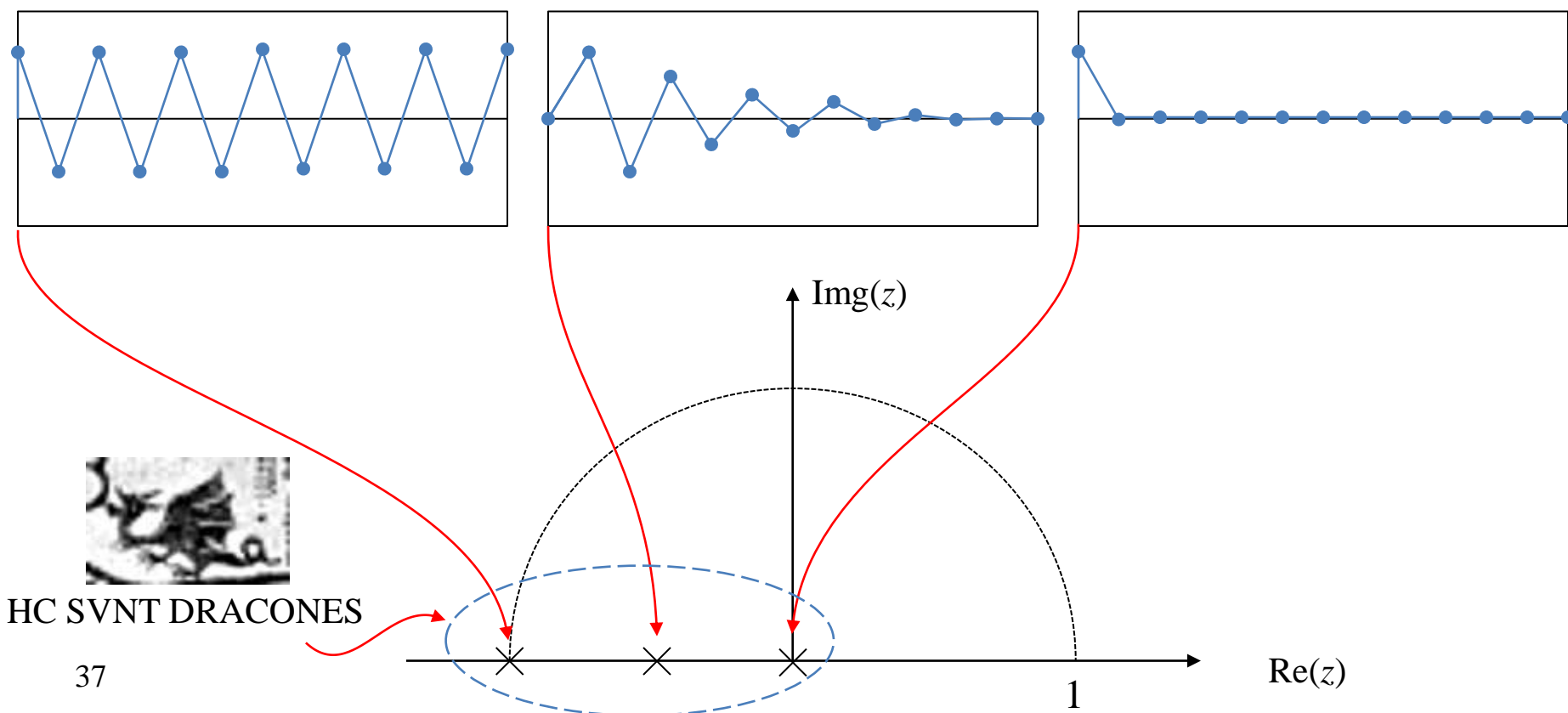
Effect of pole positions

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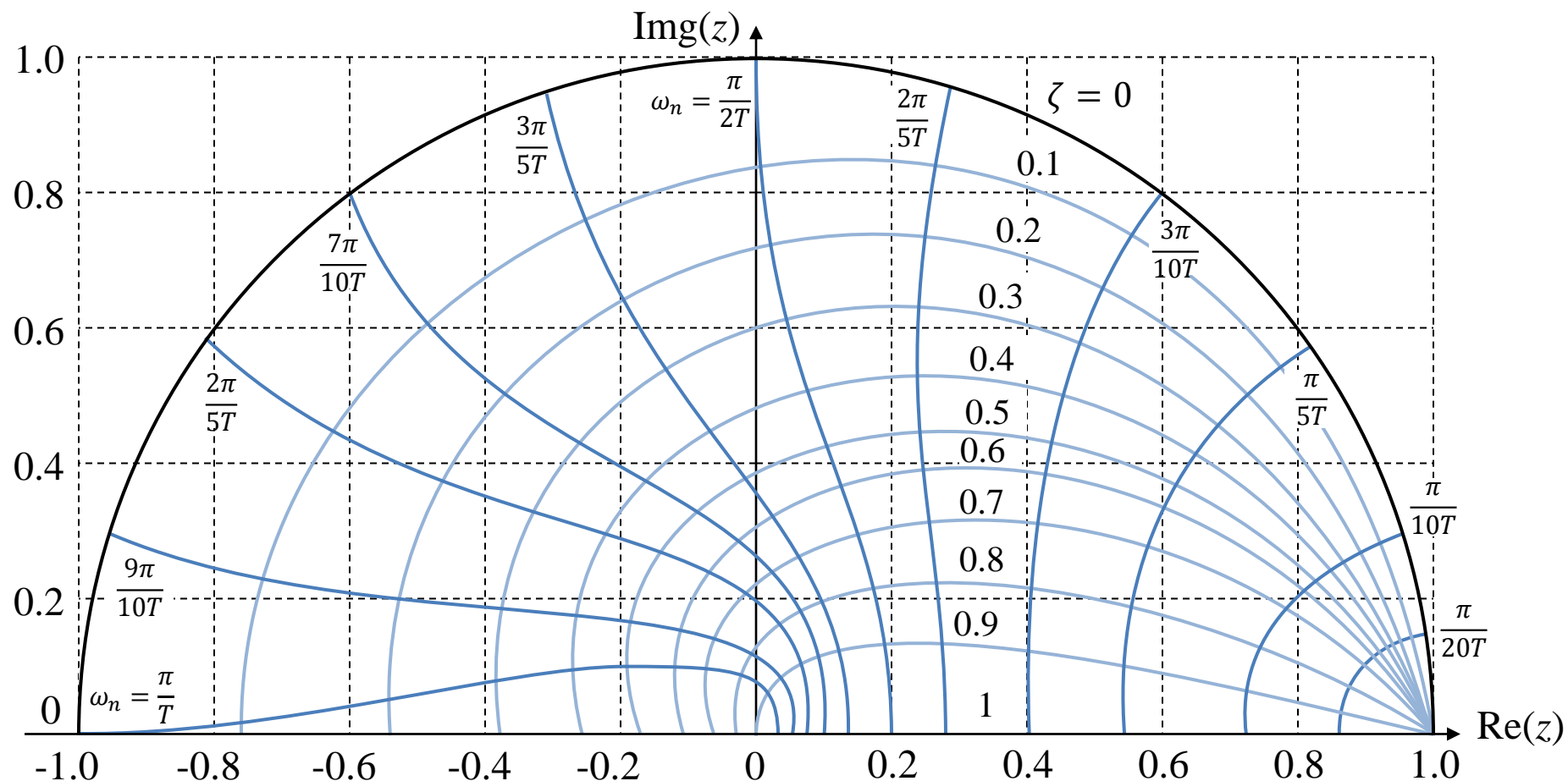
Effect of pole positions

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Damping and natural frequency

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

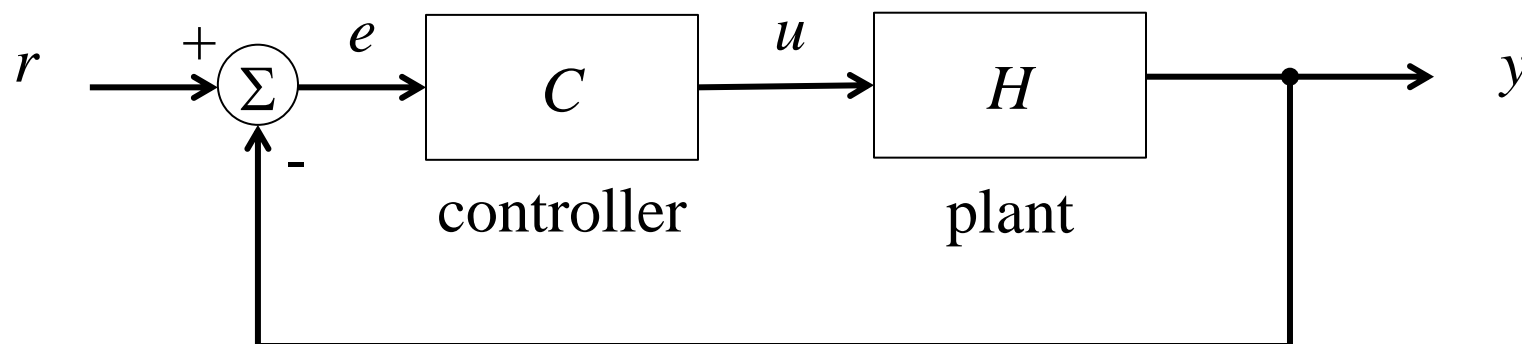


[Adapted from Franklin, Powell and Emami-Naeini]

Quick refresher: the root locus

- The transfer function for a closed-loop system can be easily calculated:

$$y = CH(r - y)$$
$$y + CHy = CHr$$
$$\therefore \frac{y}{r} = \frac{CH}{1 + CH}$$



Quick refresher: the root locus

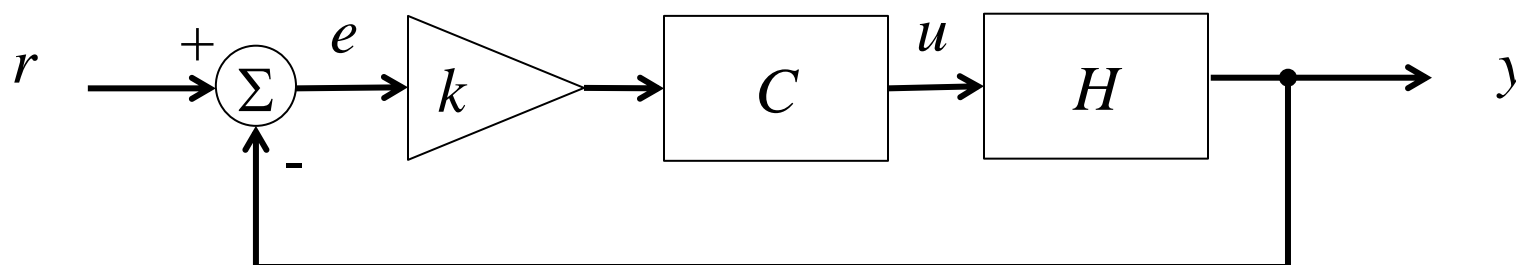
- We often care about the effect of increasing gain of a control compensator design:

$$\frac{y}{r} = \frac{kCH}{1 + kCH}$$

Multiplying by denominator:

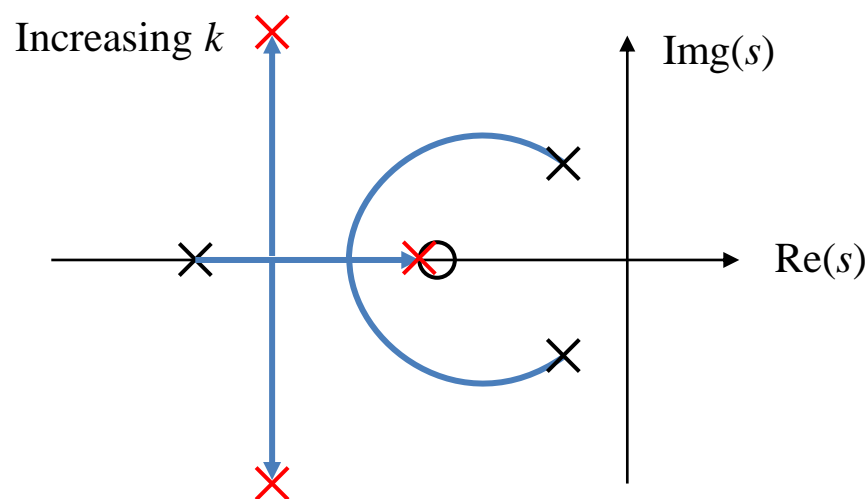
$$\frac{y}{r} = \frac{kC_n H_n}{C_d H_d + kC_n H_n}$$

characteristic polynomial



Quick refresher: the root locus

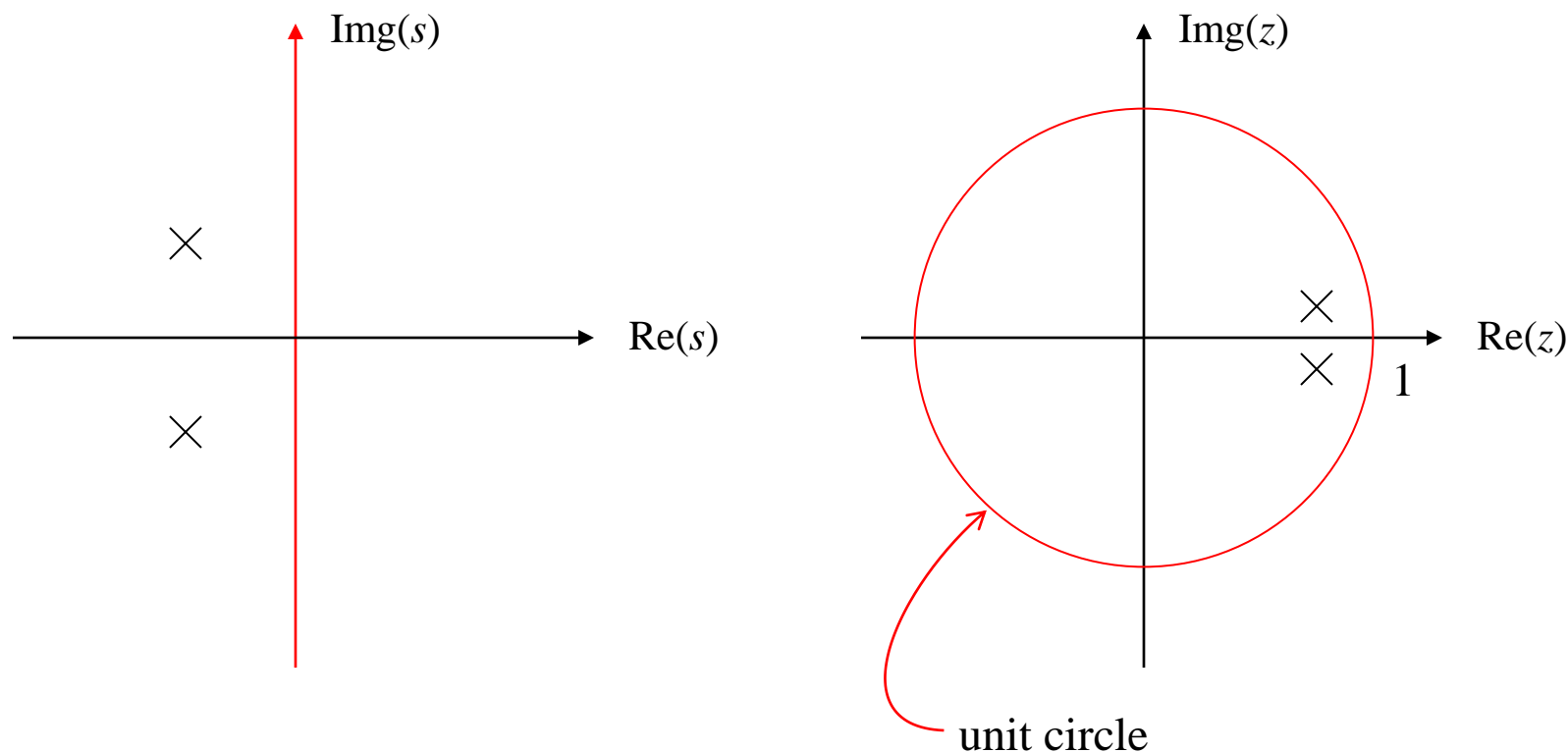
- Pole positions change with increasing gain
 - The trajectory of poles on the pole-zero plot with changing k is called the “root locus”
 - This is sometimes quite complex



(In practice you'd plot these with computers)

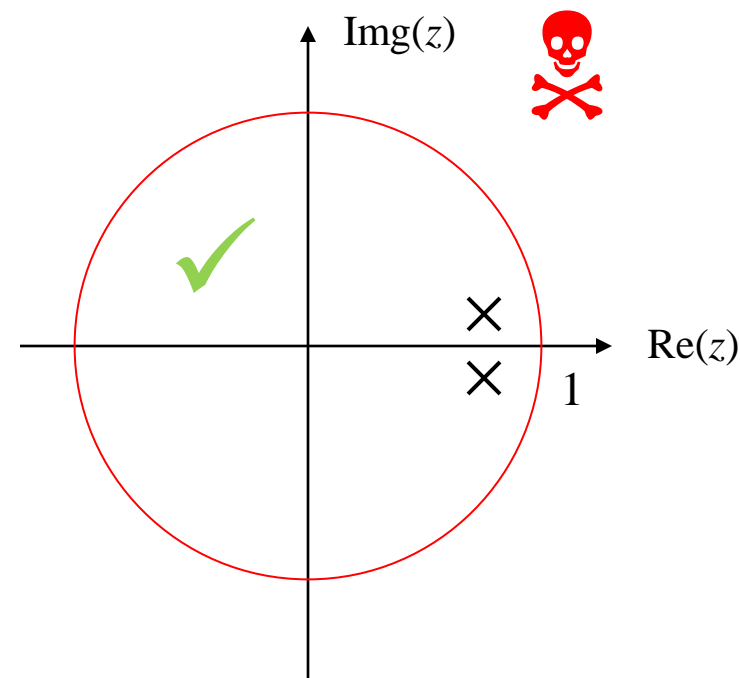
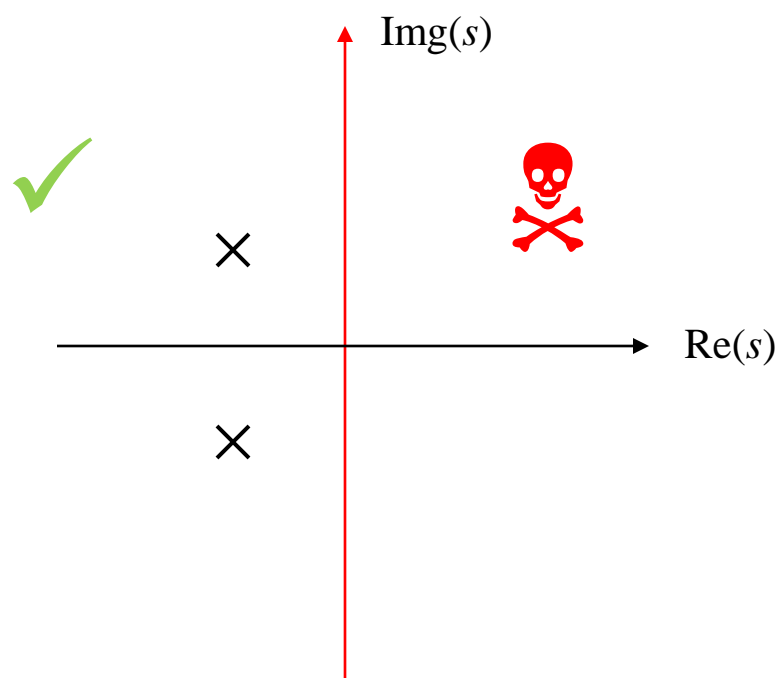
z -plane stability

- In the z -domain, the unit circle is the system stability bound



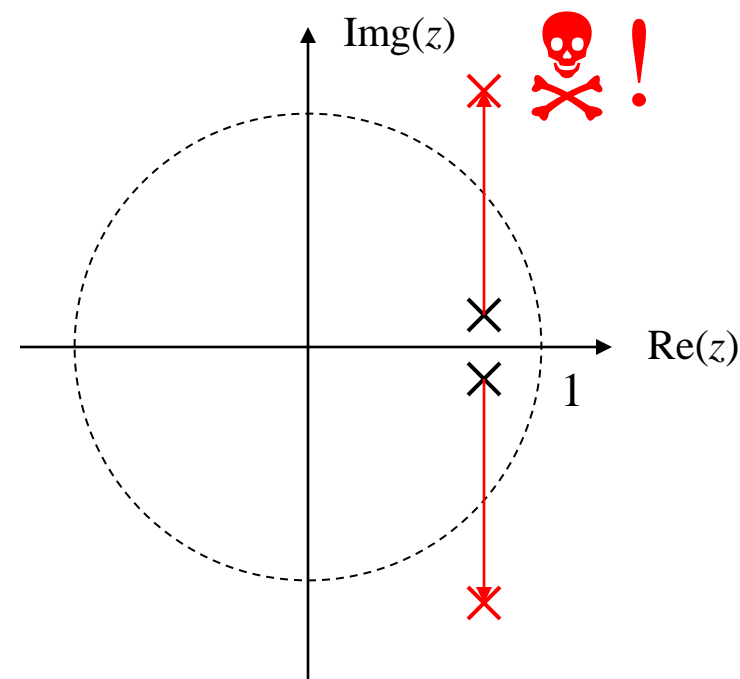
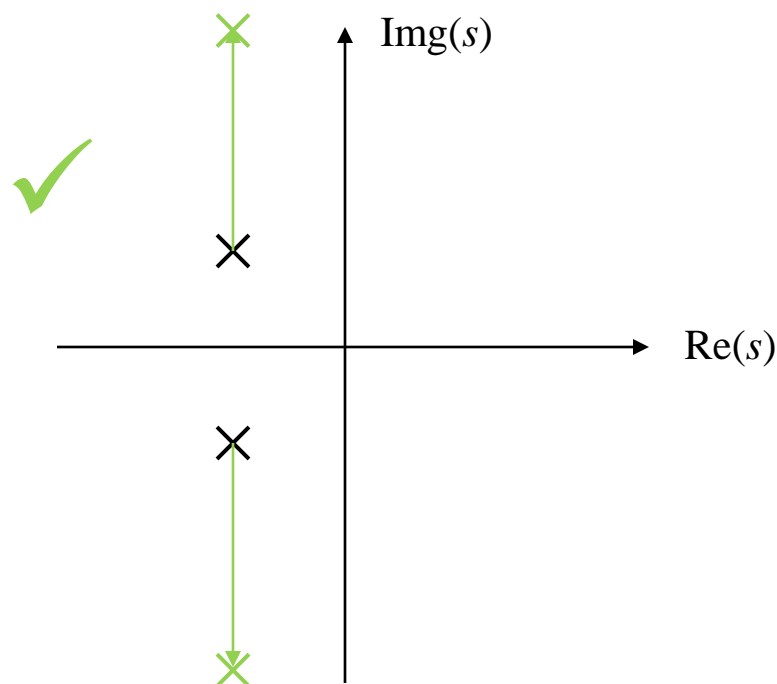
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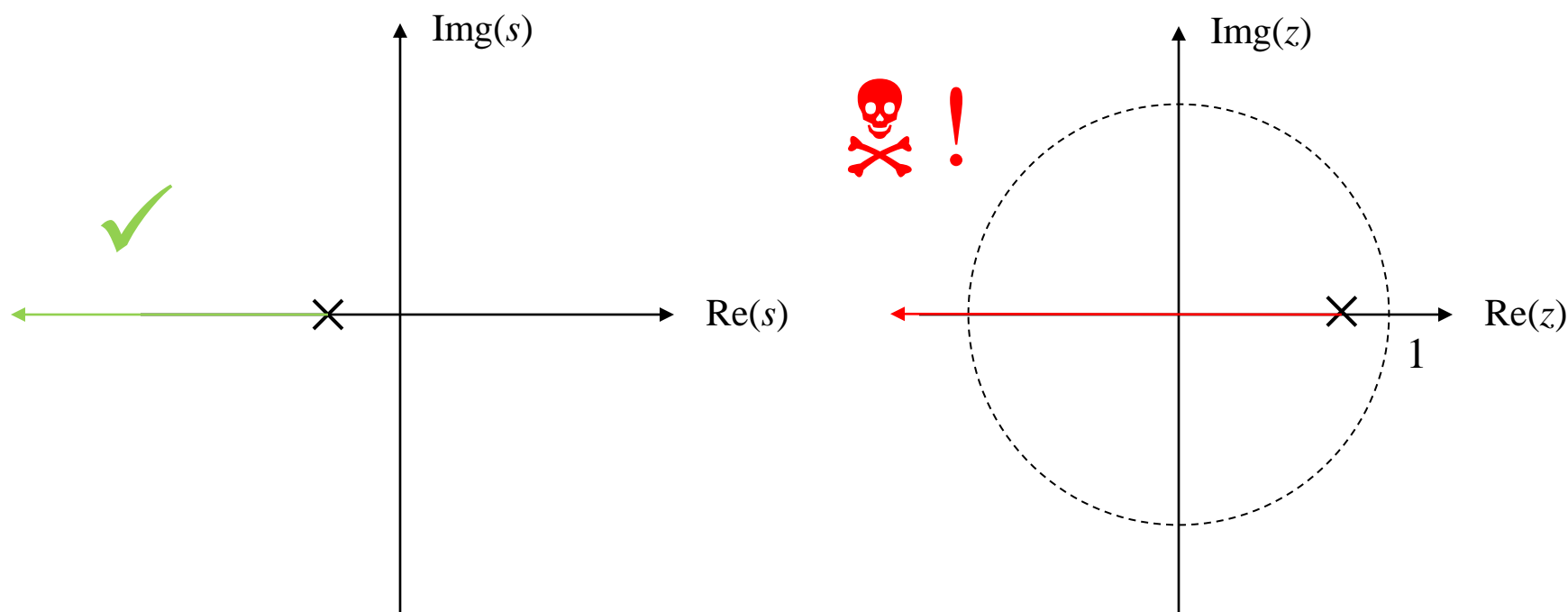
z -plane stability

- The z -plane root-locus in closed loop feedback behaves just like the s -plane:



Deep insight #2

Gains that stabilise continuous systems can actually *destabilise* digital systems!

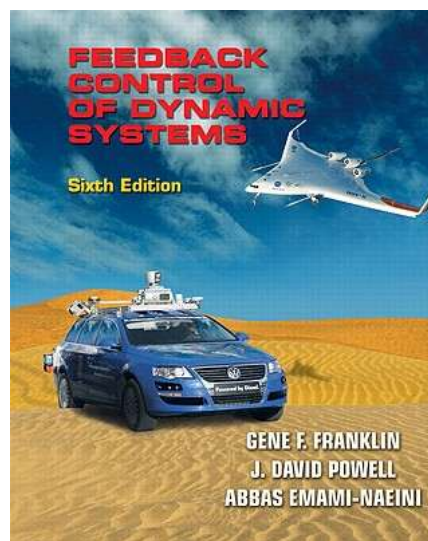


Digital design

Obviously, there's more to this digital control stuff than we thought...

Quick plug*

If you enjoyed the smash-hit “**Signals and Systems**” by Oppenheim and Willsky, you may also enjoy “**Feedback Control of Dynamic Systems**” by Franklin, Powell and Emami-Naeini.



Tune-in next time for...

Design of Digital Compensators

or

How I learned to stop worrying and love the z-transform

Fun fact: W. Hurewicz, the father of sampled data theory, died in 1956 by falling off a ziggurat at a conference outing