Discretization and Implementation of Continuous-time Design

Big picture
Discrete-time frequency response
Discretization of continuous-time design
Aliasing and anti-aliasing

Big picture

why are we learning this:

- nowadays controllers are implemented in discrete-time domain
- implementation media: digital signal processor, field-programmable gate array (FPGA), etc
- either: controller is designed in continuous-time domain and implemented digitally
- or: controller is designed directly in discrete-time domain

Frequency response of LTI SISO digital systems

$$a\sin(\omega T_s k) \longrightarrow G(z) \longrightarrow b\sin(\omega T_s k + \phi)$$
 at steady state

- \triangleright sampling time: T_s
- ϕ ($e^{j\omega T_s}$): phase difference between the output and the input
- $M(e^{j\omega T_s}) = b/a$: magnitude difference

continuous-time frequency response:

$$G(j\omega) = G(s)|_{s=j\omega} = |G(j\omega)| e^{j\angle G(j\omega)}$$

discrete-time frequency response:

$$G\left(e^{j\omega T_{s}}\right) = G(z)|_{z=e^{j\omega T_{s}}} = \left|G\left(e^{j\omega T_{s}}\right)\right| e^{j\angle G\left(e^{j\omega T_{s}}\right)}$$
$$= M\left(e^{j\omega T_{s}}\right) e^{j\phi\left(e^{j\omega T_{s}}\right)}$$

Sampling

sufficient samples must be collected (i.e., fast enough sampling frequency) to recover the frequency of a continuous-time sinusoidal signal (with frequency ω in rad/sec)



Figure: Sampling example (source: Wikipedia.org)

- the sampling frequency = $\frac{2\pi}{T_s}$
- Shannon's sampling theorem: the Nyquist frequency $(\triangleq \frac{\pi}{T_s})$ must satisfy

$$-\frac{\pi}{T_s} < \omega < \frac{\pi}{T_s}$$

bilinear transform

formula:

$$s = \frac{2}{T_s} \frac{z - 1}{z + 1} \qquad z = \frac{1 + \frac{T_s}{2} s}{1 - \frac{T_s}{2} s}$$
 (1)

intuition:

$$z = e^{sT_s} = \frac{e^{sT_s/2}}{e^{-sT_s/2}} \approx \frac{1 + \frac{I_s}{2}s}{1 - \frac{T_s}{2}s}$$

implementation: start with G(s), obtain the discrete implementation

$$G_d(z) = G(s)|_{s = \frac{2}{T_s} \frac{z-1}{z+1}}$$
 (2)

Bilinear transformation maps the closed left half s-plane to the closed unit ball in z-plane

Stability reservation: G(s) stable $\iff G_d(z)$ stable

Bilinear transform is also known as Tustin transform.

Arnold Tustin (16 July 1899 – 9 January 1994):

- British engineer, Professor at University of Birmingham and at Imperial College London
- served in the Royal Engineers in World War I
- worked a lot on electrical machines

frequency mismatch in bilinear transform

$$\frac{2}{T_s} \frac{z-1}{z+1} \bigg|_{z=e^{j\omega T_s}} = \frac{2}{T_s} \frac{e^{j\omega T_s/2} \left(e^{j\omega T_s/2} - e^{-j\omega T_s/2} \right)}{e^{j\omega T_s/2} \left(e^{j\omega T_s/2} + e^{-j\omega T_s/2} \right)} = j \underbrace{\frac{2}{T_s} \tan \left(\frac{\omega T_s}{2} \right)}_{\omega_v}$$

 $G(s)|_{s=j\omega}$ is the true frequency response at ω ; yet bilinear implementation gives,

bilinear transform with prewarping

goal: extend bilinear transformation such that

$$G_d(z)|_{z=e^{j\omega T_s}} = G(s)|_{s=i\omega}$$

at a particular frequency ω_p solution:

$$s = p \frac{z-1}{z+1}, \qquad z = \frac{1+\frac{1}{\rho}s}{1-\frac{1}{\rho}s}, \qquad p = \frac{\omega_p}{\tan\left(\frac{\omega_p T_s}{2}\right)}$$

which gives

$$G_d(z) = G(s)|_{s=\frac{\omega_p}{\tan(\frac{\omega_p T}{2})}^{z-1}}$$

and

$$\frac{\omega_p}{\tan\left(\frac{\omega_p T_s}{2}\right)} \frac{z-1}{z+1} = j \frac{\omega_p}{\tan\left(\frac{\omega_p T_s}{2}\right)} \tan\left(\frac{\omega_p T_s}{2}\right)$$

bilinear transform with prewarping

choosing a prewarping frequency ω_p :

must be below the Nyquist frequency:

$$0<\omega_p<\frac{\pi}{T_s}$$

- lacktriangle standard bilinear transform corresponds to the case where $\omega_p=0$
- ightharpoonup the best choice of ω_p depends on the important features in control design

example choices of ω_p :

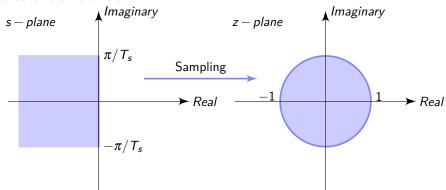
- at the cross-over frequency (which helps preserve phase margin)
- at the frequency of a critical notch for compensating system resonances

Sampling and aliasing

sampling maps the continuous-time frequency

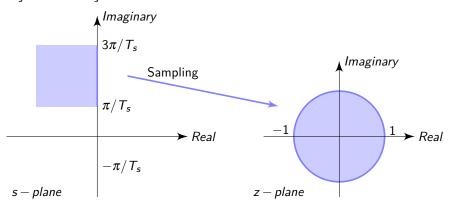
$$-\frac{\pi}{T_s} < \omega < \frac{\pi}{T_s}$$

onto the unit circle



Sampling and aliasing

sampling also maps the continuous-time frequencies $\frac{\pi}{T_s} < \omega < 3\frac{\pi}{T_s}$, $3\frac{\pi}{T_s} < \omega < 5\frac{\pi}{T_s}$, etc, onto the unit circle



Sampling and aliasing

Example (Sampling and Aliasing)

 $T_s=1/60$ sec (Nyquist frequency 30 Hz).

a continuous-time 10-Hz signal [10 Hz \leftrightarrow $2\pi \times 10$ rad/sec $\in (-\pi/T_s,\pi/T_s)$]

$$y_1(t) = \sin(2\pi \times 10t)$$

is sampled to

$$y_1(k) = \sin\left(2\pi \times \frac{10}{60}k\right) = \sin\left(2\pi \times \frac{1}{6}k\right)$$

a 70-Hz signal $[2\pi \times 70 \text{ rad/sec } \in (\pi/T_s, 3\pi/T_s)]$

$$y_2(t) = \sin(2\pi \times 70t)$$

is sampled to

$$y_2(k) = \sin\left(2\pi \times \frac{70}{60}k\right) = \sin\left(2\pi \times \frac{1}{6}k\right) \equiv y_1(k)!$$

Anti-aliasing

need to avoid the negative influence of *aliasing* beyond the Nyquist frequencies

- ▶ sample faster: make π/T_s large; the sampling frequency should be high enough for good control design
- ▶ anti-aliasing: perform a low-pass filter to filter out the signals $|\omega| > \pi/T_s$

Summary

- 1. Big picture
- 2. Discrete-time frequency response
- 3. Approximation of continuous-time controllers
- 4. Sampling and aliasing

Sampling example

continuous-time signal

$$y(t) = \begin{cases} e^{-at}, & t \ge 0 \\ 0, & t < 0 \end{cases}, \ a > 0$$
$$\mathscr{L}\{y(t)\} = \frac{1}{s+a}$$

discrete-time sampled signal

$$y(k) = \begin{cases} e^{-aT_s k}, & k \ge 0\\ 0, & k < 0 \end{cases}$$
$$\mathscr{Z}\{y(k)\} = \frac{1}{1 - z^{-1}e^{-aT_s}}$$

▶ sampling maps the continuous-time pole $s_i = -a$ to the discrete-time pole $z_i = e^{-aT_s}$, via the mapping

$$z_i = e^{s_i T_s}$$