

Control Systems 2 Cheatsheet

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Version: May 14, 2024

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1 Discrete Time

1.1 Sampling

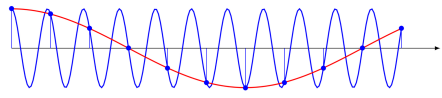
$$T_s: \text{Sampling Time} \quad \omega_s = \frac{2\pi}{T_s}$$

1.2 Aliasing

$$y_1[k] = \cos(\omega k T_s), \quad k = 0, 1, 2$$

$$y_2[k] = \cos((\omega + n \frac{2\pi}{T_s}) k T_s), \quad n = 0, 1, 2$$

$$= \cos(\omega k T_s + 2\pi n k) = y_1[k]$$



1.2.1 Nyquist-Shannon Sampling theorem

$$f_N = \frac{1}{2T_s} [\text{Hz}] \quad \text{or} \quad \omega_N = \frac{\pi}{T_s} \left[\frac{\text{rad}}{\text{s}} \right]$$

No aliasing if $\omega < \omega_N$!

1.3 DT State Space Representation

$$x[k+1] = A_d x[k] + B_d u[k]$$

$$= e^{A^T} x[k] + \left(\int_0^T e^{A(T-\tau)} d\tau \right) B u[k]$$

$$y[k] = C_d x[k] + D_d u[k]$$

$$= C x[k] + D u[k]$$

If A is invertible: $B_d = A^{-1}(A_d - I)B$

1.3.1 Homogeneous response

$$x[0] = x_0, \quad u[k] = 0$$

$$x[k+1] = Ax[k], \Rightarrow x[k] = A^k x_0$$

$$y[k] = Cx[k]$$

$$= CA^k x_0$$

$$A^k = (TAT^{-1})^k = T\Lambda^k T^{-1}$$

$$\lim_{k \rightarrow +\infty} A^k = 0 \Rightarrow |\lambda_i| < 1$$

1.3.2 Forced response

$$x[1] = Bu[0],$$

$$x[2] = ABu[0]Bu[1], \dots,$$

$$x[k] = \sum_{i=0}^{k-1} A^{k-i-1} Bu[i],$$

$$y[k] = \underbrace{CA^k x_0}_{\text{Homogeneous}} + \underbrace{C \sum_{i=0}^{k-1} A^{k-i-1} Bu[i] + Du[k]}_{\text{Forced}}$$

1.4 State Transfer Function

$$u[k] = u_0 z^k = u_0 e^{ksT} = u(kT)$$

$$y[k] = C \sum_{i=0}^{k-1} A^{k-i-1} B u_0 z^i + D u_0 z^k$$

$$= \underbrace{CA^k (x_0 - C(zI - A)^{-1} B u_0)}_{\text{Transient}}$$

$$+ \underbrace{C(zI - A)^{-1} B u_0 z^k + D u_0 z^k}_{\text{Steady-state}}$$

$$\lim_{k \rightarrow +\infty} A^k = 0 \Rightarrow y[k] \approx [C(zI - A)^{-1} B + D] u[k]$$

$$y[k] \approx G(z) u[k], \quad G(z) := C(zI - A)^{-1} B + D$$

1.5 Approximation Methods

1.5.1 Emulation

Exact	$s = \frac{1}{T_s} \cdot \ln(z)$	$z = e^{sT_s}$
Euler forward	$s = \frac{z-1}{T_s}$	$z = s \cdot T_s + 1$
Euler backward	$s = \frac{z-1}{z \cdot T_s}$	$z = \frac{1}{1-s \cdot T_s}$
Tustin	$s = \frac{2}{T_s} \cdot \frac{z-1}{z+1}$	$z = \frac{1+s \cdot \frac{T_s}{2}}{1-s \cdot \frac{T_s}{2}}$

2 System Properties

2.1 Similarity Transformation

$$\begin{cases} x^+ = Ax + Bu \\ y = Cx + Du \end{cases} \Rightarrow \begin{cases} \hat{x}^+ = (T^{-1}AT)\hat{x} + (T^{-1}B)u \\ y = (CT)\hat{x} + Du \end{cases}$$

2.1.1 Modal decomposition

$$\tilde{x}_i(t) = e^{\lambda_i t} \tilde{x}_i(0) \quad x(t) = \sum_{i=1}^n e^{\lambda_i t} \tilde{x}_i(0) v_i$$

2.2 Reachability

$$\mathcal{R} := [A^{n-1}B | \dots | AB | B] \in \mathbb{R}^{n \times n \cdot m} \quad U := \begin{bmatrix} u[0] \\ u[1] \\ \vdots \\ u[n-1] \end{bmatrix}$$

$$\Rightarrow x[n] = \mathcal{R}U$$

The system is reachable if and only if \mathcal{R} has full row rank n

2.3 Observability

$$\mathcal{O} = \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix} \quad Y = \begin{bmatrix} y[0] \\ y[1] \\ \vdots \\ y[n-1] \end{bmatrix} \quad Y = \mathcal{O}x[0]$$

The system is observable if and only if \mathcal{O} has full column rank n

2.4 Controllability

A system is controllable if, for any initial condition x_0 , there exists a control input u that brings the state x to 0 in finite time.

For CT Systems: Controllability = Reachability

For DT Systems: A is invertible \Rightarrow Controllability = Reachability

2.5 Kalman Decomposition

$$x^+ = \begin{bmatrix} \Lambda_{r\bar{o}} & 0 & 0 & 0 \\ 0 & \Lambda_{r\bar{o}} & 0 & 0 \\ 0 & 0 & \Lambda_{\bar{r}\bar{o}} & 0 \\ 0 & 0 & 0 & \Lambda_{\bar{r}\bar{o}} \end{bmatrix} x + \begin{bmatrix} B_{r\bar{o}} \\ B_{r\bar{o}} \\ 0 \\ 0 \end{bmatrix} u,$$

$$y = \begin{bmatrix} 0 & C_{r\bar{o}} & 0 & C_{\bar{r}\bar{o}} \end{bmatrix} x + Du$$

• Stabilizability

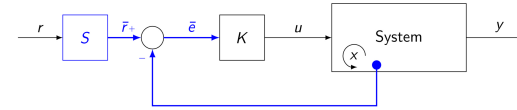
A system is said to be stabilizable if all unstable modes are reachable

• Detectability

A system is said to be detectable if all unstable modes are observable

3 State Feedback

3.1 State Feedback



$$x^+ = (A - BK)x + BKSr$$

$$= (A - BK)x + B\bar{N}r, \quad \bar{N} = KS$$

$$y = Cx + (D = 0)$$

$$G_{yr}^{cl}(s) = C(sI - A + BK)^{-1} BKS$$

$$= C(sI - A - BK)^{-1} B\bar{N}$$

3.1.1 K - Direct Method

$$p_{cl}^* = \sum_{i=1}^n (s + \lambda_i), \quad K = [k_1 \quad k_2 \quad \dots \quad k_n]$$

$$p_{cl} = \det(sI - A + BK) = p_{cl}^*$$

3.1.2 K - Ackermann's Formula

$$K = [0 \quad \dots \quad 0 \quad 1] \mathcal{R}^{-1} p_{cl}^*(A)$$

3.1.3 S

$$\text{No steady-state error: } \Rightarrow G_{yr}^{cl}(0) = 1$$

$$\Rightarrow \bar{N} = -(C(A - BK)^{-1}B)^{-1}, \quad S = K^{-1}\bar{N}$$

3.2 LQR

LQR guarantees:

- phase margin $\geq 60^\circ$
- gain margin $(\frac{1}{2}, +\infty)$

3.2.1 Continuous Time

$$\min_K J(x, u) = \int_0^{+\infty} [x(t)^T Q x(t) + u(t)^T R u(t)] dt,$$

$$\text{s.t.: } \dot{x}(t) = Ax(t) + Bu(t),$$

$$u(t) = -Kx(t)$$

$$\text{soln.: } 0 = A^T P + PA + Q - PBR^{-1}B^T P$$

$$K = R^{-1}B^T P$$

3.2.2 Discrete Time

$$\min_K J(x, u) = \sum_{k=0}^{+\infty} (x[k]^T Q x[k] + u[k]^T R u[k]),$$

$$\text{s.t.: } x[k+1] = Ax[k] + Bu[k],$$

$$u[k] = -Kx[k]$$

$$\text{soln.: } P = A^T P A - (A^T P B)(B^T P B + R)^{-1} (B^T P A) + Q$$

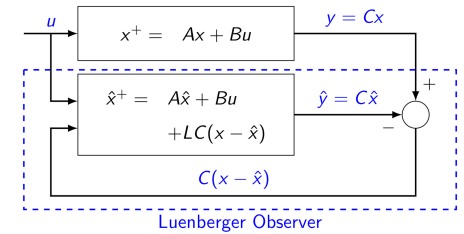
$$K = (R + B^T P B)^{-1} B^T P A$$

3.2.3 LQR Servo

$$\begin{bmatrix} x \\ \epsilon \end{bmatrix}^+ = \begin{bmatrix} A & 0 \\ -C & 0 \end{bmatrix} \begin{bmatrix} x \\ \epsilon \end{bmatrix} + \begin{bmatrix} B \\ 0 \end{bmatrix} u + \begin{bmatrix} 0 \\ I \end{bmatrix} r$$

4 State Estimation

4.1 Luenberger Observer



$$\hat{x}^+(t) = (A - LC)\hat{x}(t) + Bu(t) + Ly(t)$$

$$\hat{y}(t) = C\hat{x}(t)$$

\Rightarrow Exactly the same as finding a control gain K

$$L = p_{cl}^*(A) \mathcal{O}^{-1} [0, \dots, 0, 1]^T$$

4.2 LQE

$$Q = \mathbb{E}[w(t)w(t)^T], \quad R = \mathbb{E}[n(t)n(t)^T], \quad \forall t \geq 0$$

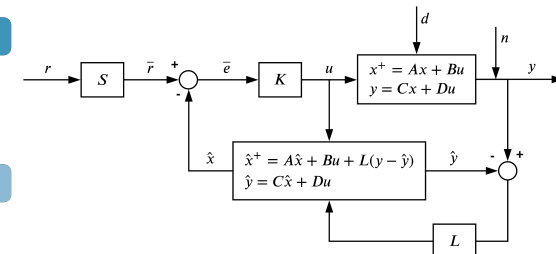
Problem: Find L, such that the steady-state covariance of the state error is minimized.

Solution:

$$0 = AY + YA^T - YC^T R^{-1} CY + Q$$

$$L = -YC^T R^{-1}$$

5 Dynamic Output Feedback



5.1 LQG

$$\begin{bmatrix} x \\ \eta \end{bmatrix}^+ = \begin{bmatrix} A - BK & BK \\ 0 & A - LC \end{bmatrix} \begin{bmatrix} x \\ \eta \end{bmatrix} + \begin{bmatrix} BKS \\ 0 \end{bmatrix} r$$

$$y = [C \quad 0] \begin{bmatrix} x \\ \eta \end{bmatrix}$$

5.1.1 LQG Servo

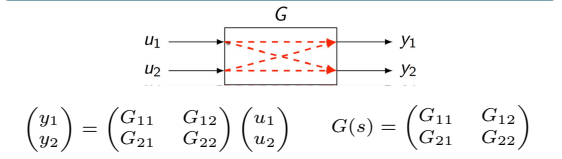
$$\begin{bmatrix} x \\ x_I \\ \eta \end{bmatrix}^+ = \begin{bmatrix} A - BK & -BK_I & BK \\ -C & 0 & 0 \\ 0 & 0 & A - LC \end{bmatrix} \begin{bmatrix} x \\ x_I \\ \eta \end{bmatrix} + \begin{bmatrix} BKS \\ I \\ 0 \end{bmatrix} r$$
$$y = \begin{bmatrix} C & 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ x_I \\ \eta \end{bmatrix}$$

5.1.2 LQG Stability

LQG guarantees closed loop stability, but the margins can be arbitrarily small.

6 MIMO

6.1 Transfer Function



Push through identity

$$G_1(I + G_2G_1)^{-1} = (I + G_1G_2)^{-1}G_1$$

MIMO state space to tf

$$G(s) = C(sI - A)^{-1}B + D$$

6.2 State Space

$$x^+ = \underbrace{\underbrace{A}_{n \times n}}_{n \times n} \underbrace{x}_{n \times 1} + \underbrace{\underbrace{B}_{n \times m}}_{n \times m} \underbrace{u}_{m \times 1}$$
$$y = \underbrace{\underbrace{C}_{l \times n}}_{l \times n} \underbrace{x}_{n \times 1} + \underbrace{\underbrace{D}_{l \times m}}_{l \times m} \underbrace{u}_{m \times 1}$$

- $x \in \mathbb{R}^n$, where n is the order of the system
- $u \in \mathbb{R}^m$, where m is the number of inputs
- $y \in \mathbb{R}^l$, where l is the number of outputs