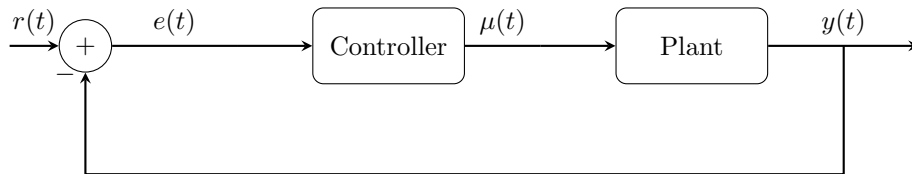


# ECE311 Course Notes

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0.1



Where

- $\mu(t)$  is the control input (decision variable)
- $y(t)$  is the output variable (measured with sensors and also the target of our control)
- $r(t)$  is the reference signal. We want  $y(t) \rightarrow r(t)$  as  $t \rightarrow \infty$
- $e(t)$  is the tracking error. We want  $e(t) \rightarrow 0$  as  $t \rightarrow \infty$

## 1 Signals

### 1.1 Time Constant

$$e^{-A \cdot t} \leftrightarrow e^{-t/\tau} \rightarrow \tau = \frac{1}{A}$$

## 2 Control System Models

### 2.1 Non-Linear Time Invariant State Space (NN)

$$x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \quad y = \begin{bmatrix} y_1 \\ \vdots \\ y_p \end{bmatrix} \quad u = \begin{bmatrix} u_1 \\ \vdots \\ u_m \end{bmatrix}$$

$$\dot{x} = f(x, u) = f(x_1, \dots, x_n, u_1, \dots, u_m)$$
$$y = h(x, u) = h(x_1, \dots, x_n, u_1, \dots, u_m)$$

## 2.2 LTI State Space Models

$$\begin{aligned} \dot{x}_i &= a_{i1}x_1 + \dots + a_{in}x_n + b_{i1}u_1 + \dots + b_{im}u_m \\ \dot{y}_j &= c_{j1}x_1 + \dots + c_{jn}x_n + d_{j1}u_1 + \dots + d_{jm}u_m \\ \dot{x} &= Ax + Bu \quad y = Cx + Du \quad x = [x_1 \dots x_n]^T \end{aligned}$$

## 2.3 LTI Input/Output Models (I/O) Models

$$\begin{aligned} \frac{d^n y}{dt^n} + a_{n-1} \frac{d^{n-1} y}{dt^{n-1}} + \dots + a_1 \frac{dy}{dt} + a_0 y &= \\ b_m \frac{d^m u}{dt^m} + b_{m-1} \frac{d^{m-1} u}{dt^{m-1}} + \dots + b_1 \frac{du}{dt} + b_0 u & \end{aligned}$$

for  $m \leq n$

## 3 Equilibrium

For a NN system, a state  $\bar{x} \in \mathbb{R}$  is an **equilibrium** if

$$f(\bar{x}, \bar{u}) = [0, \dots, 0]^T$$

## 4 Linearization

$$\begin{aligned} \bar{x} &= [\bar{x}_1, \dots, \bar{x}_n]^T, \quad \bar{u} \\ \tilde{x} := x - \bar{x} \quad \tilde{u} := u - \bar{u} \quad \tilde{y} := y - h(\bar{x}, \bar{u}) \end{aligned}$$

Then, the **linearization** of  $\bar{x}$  is given by

$$\begin{aligned} \dot{\tilde{x}} &= A\tilde{x} + B\tilde{u} & \tilde{y} &= C\tilde{x} + D\tilde{u} \\ A &= \left[ \frac{\partial f}{\partial x} \Big|_{(\bar{x}, \bar{u})} \right] & B &= \left[ \frac{\partial f}{\partial u} \Big|_{(\bar{x}, \bar{u})} \right] \\ C &= \left[ \frac{\partial h}{\partial x} \Big|_{(\bar{x}, \bar{u})} \right] & D &= \left[ \frac{\partial h}{\partial u} \Big|_{(\bar{x}, \bar{u})} \right] \end{aligned}$$

## 5 Matrix Inverses

$$\begin{aligned} A^{-1} &= \frac{\text{adj}(A)}{\det A} = \frac{(C)^T}{\det A}, \quad C_{ij} = (-1)^{i+j} M_{ij} \\ \text{adj}(A) &= \begin{bmatrix} a_{22} & -a_{12} \\ -a_{21} & a_{11} \end{bmatrix} \end{aligned}$$

$$\text{adj}(A) = \begin{pmatrix} + \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} & - \begin{vmatrix} a_{12} & a_{13} \\ a_{32} & a_{33} \end{vmatrix} & + \begin{vmatrix} a_{12} & a_{13} \\ a_{22} & a_{23} \end{vmatrix} \\ - \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} & + \begin{vmatrix} a_{11} & a_{13} \\ a_{31} & a_{33} \end{vmatrix} & - \begin{vmatrix} a_{11} & a_{13} \\ a_{21} & a_{23} \end{vmatrix} \\ + \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix} & - \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} & + \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} \end{pmatrix} \quad (1)$$

## 6 Final Value Theorem

If  $\lim_{t \rightarrow \infty} f(t)$  exists, then

$$\lim_{t \rightarrow \infty} f(t) = \lim_{s \rightarrow 0} sF(s)$$

### 6.1 FVT Existence Condition

A signal  $f(t)$  is bounded iff  $F(s)$  has poles with real part  $\leq 0$  and non-repeated poles with real part  $= 0$

## 7 Initial Value Theorem

$$\lim_{t \rightarrow 0} f(t) = \lim_{s \rightarrow \infty} sF(s)$$

## 8 Laplace Transform (LT)

Let  $f(t)$  be a function  $f : \mathbb{R} \rightarrow \mathbb{R}$ . Then

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

Where  $F : \mathbb{C} \rightarrow \mathbb{C}$ . The LT exists if

1.  $f(t)$  is Piecewise Continuous (PWC)
2.  $\exists M \geq 0, a \in \mathbb{R}$  s.t.  $|f(t)| \leq Me^{at} \forall t \geq 0$

### 8.1 Basic Laplace Table

$$\begin{array}{l|l} \mathcal{L}\{1(t)\} = \frac{1}{s} & \mathcal{L}\{t^n\} = \frac{n!}{s^{n+1}} \\ \mathcal{L}\left\{\frac{t^k}{k!}e^{at}\right\} = \frac{1}{(s-a)^{k+1}} & \mathcal{L}\{e^{at}\} = \frac{1}{s-a} \\ \mathcal{L}\{\sin(kt)\} = \frac{k}{s^2+k^2} & \mathcal{L}\{\cos(kt)\} = \frac{s}{s^2+k^2} \\ \mathcal{L}\{\sinh(kt)\} = \frac{k}{s^2-k^2} & \mathcal{L}\{\cosh(kt)\} = \frac{s}{s^2-k^2} \end{array}$$

## 8.2 First Translation Theorem

$$\mathcal{L}\{e^{at}f(t)\}(s) = \mathcal{L}\{f(t)\}(s-a) = F(s-a)$$

## 8.3 Second Translation Theorem

$$\mathcal{L}\{f(t-a)u(t-a)\} = e^{-as}F(s)$$

where  $u$  is the unit step function and  $a > 0$ .

## 8.4 Transforms of Derivatives

If  $f, f', \dots, f^{(n-1)}$  are cts on  $[0, \infty)$  and are of expon. order, and if  $f^{(n)}(t)$  is piecewise cts on  $[0, \infty)$ , then

$$\mathcal{L}\{f^{(n)}(t)\} = s^n F(s) - s^{n-1}f(0) - s^{n-2}f'(0) - \dots - f^{(n-1)}(0)$$

$$\mathcal{L}\{f''(t)\} = s^2F(s) - sf(0) - f'(0)$$

## 8.5 Derivatives of Transforms

If  $\mathcal{L}\{f(t)\} = F(s)$  and  $n = 1, 2, 3, \dots$ , then

$$\mathcal{L}\{t^n f(t)\} = (-1)^n \frac{d^n}{ds^n} F(s)$$

## 8.6 Transform of Integrals

$$\mathcal{L}\left\{\int_0^t f(\tau)d\tau\right\} = \frac{F(s)}{s} \quad \int_0^t f(\tau)d\tau = \mathcal{L}^{-1}\left\{\frac{F(s)}{s}\right\}$$

# 9 Inverse Laplace Transform

## 9.1 Basic Inverse Laplace Transforms

$$\begin{array}{l|l} \mathcal{L}^{-1}\left\{\frac{1}{s}\right\} = 1 & \mathcal{L}^{-1}\{1\} = \delta(t) \\ \mathcal{L}^{-1}\left\{\frac{1}{s-a}\right\} = e^{at} & \mathcal{L}^{-1}\left\{\frac{n!}{s^{n+1}}\right\} = t^n \\ \mathcal{L}^{-1}\left\{\frac{k}{s^2+k^2}\right\} = \sin(kt) & \mathcal{L}^{-1}\left\{\frac{s}{s^2+k^2}\right\} = \cos(kt) \\ \mathcal{L}^{-1}\left\{\frac{k}{s^2-k^2}\right\} = \sinh(kt) & \mathcal{L}^{-1}\left\{\frac{s}{s^2-k^2}\right\} = \cosh(kt) \end{array}$$

## 9.2 Inverse Laplace Transform Formula

$$f(t) = \frac{1}{2\pi i} \int_{\sigma-i\infty}^{\sigma+i\infty} F(z)e^{zt} dz = \sum_{k=1}^n \text{Res}(e^{st}F(s), s_k)$$

### 9.3 Residue Calculation

In general, the residue of a function  $F(s)$  at a pole can be calculated with

$$\text{Res}(F(s), s_k) = \frac{1}{(n-1)!} \lim_{s \rightarrow s_k} \frac{d^{n-1}}{ds^{n-1}} (s - s_0)^n F(s)$$

Where  $n \geq 1$  is the order of the function  $F(s)$ .

## 10 Model Conversions

### 10.1 Input/Output to Transfer Function

For an IO model of the form

$$\frac{d^n y}{dt^n} + a_{n-1} \frac{d^{n-1} y}{dt^{n-1}} + \cdots + a_0 y = b_m \frac{d^m u}{dt^m} + \cdots + b_0 u$$

with  $y(0) = \dot{y}(0) = \ddot{y}(0) = \cdots = 0$ , the equivalent Transfer Function model is given by

$$G(s) = \frac{b_m s^m + \cdots + b_0}{s^n + a_{n-1} s^{n-1} + \cdots + a_0}$$

### 10.2 Transfer Function to Input/Output

For a Transfer Function model of the form,

$$Y(s) = G(s)U(s)$$

the equivalent Input/Output model is given by

$$y(t) = \mathcal{L}^{-1}\{Y(s)\} = \mathcal{L}^{-1}\{G(s)U(s)\}$$
$$y(t) = g(t) * u(t) = \int_0^t g(t - \tau)u(\tau)d\tau$$

### 10.3 State Space to Transfer Function

For a State Space model of the form,

$$\dot{x} = Ax + Bu$$
$$y = Cx + Du$$

the equivalent Transfer Function model is given by

$$Y(s) = [C(sI - A)^{-1}B + D]U(s)$$
$$G(S) = C(sI - A)^{-1}B + D$$

#### 10.3.1 Notes

The values of  $S \in \mathbb{C}$  for which  $sI - A$  is not invertible are poles of  $G(s)$

## 10.4 Transfer Function to State Space

$$V(s) = \frac{1}{s^n + \dots + a_0} U(s), Y(s) = (b_m s^m + \dots + b_0) V(s)$$

$$f(x, u) = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ & & \dots & & \\ 0 & 0 & 0 & \dots & 1 \\ -a_0 & -a_1 & -a_2 & \dots & -a_{n-1} \end{bmatrix} x + \begin{bmatrix} 0 \\ 0 \\ \dots \\ 0 \\ 1 \end{bmatrix} u$$

$$y = [b_0 \ b_1 \ \dots \ b_m \ 0 \ \dots \ 0] x$$

## 11 Poles

$$\text{1st Order: } (s + p_1) \quad \text{2nd Order: } [(s + \sigma)^2 + \omega_d^2]$$

### 11.1 Pole Poly Representations

$$\frac{as + b}{[(s + \sigma)^2 + \omega_d^2]} \leftrightarrow \frac{as + b}{s^2 + 2\zeta\omega_n s + \omega_n^2}$$

$$\sigma = \zeta\omega_n \quad \omega_d = \sqrt{\omega_n^2 - \zeta^2\omega_n^2} = \omega_n \sqrt{1 - \zeta^2}$$

$$\zeta = \frac{\sigma}{\sqrt{\sigma^2 + \omega_d^2}} \quad \omega_n = \sqrt{\sigma^2 + \omega_d^2}$$

## 12 Transient Performance

### 12.1 2nd Order Systems

#### 12.1.1 Settling Time

$$T_s \approx -\frac{\ln(2 \cdot 10^{-2} \sqrt{1 - \zeta^2})}{\zeta\omega_n} \approx \frac{4}{\zeta\omega_n} = \frac{4}{\sigma}$$

#### 12.1.2 Percent Overshoot and Peak Time

$$T_p = \frac{\pi}{\omega_d} = \frac{\pi}{\omega_n \sqrt{1 - \zeta^2}}, \zeta = \frac{-\ln(\%OS)}{\sqrt{\pi^2 + \ln^2(\%OS)}}$$

$$\%OS = y(T_p) - 1 = e^{\frac{-\zeta\pi}{\sqrt{1 - \zeta^2}}}$$

#### 12.1.3 Rise Time

$$T_r\omega_n \propto \zeta \quad \longrightarrow \quad T_r\omega_n \approx \frac{1.8}{\omega_n}$$

### 12.1.4 Effect of Additional Pole/Zeroes

Additional Pole/Zeroes in LHP have little effect, as long as  $Re\{P\} \ll \sigma \rightarrow Re\{P\} \leq 10 \cdot \sigma$ . Zeroes in RHP (**Nonminimum Phase**) and change sign of  $y(\infty)$ .

## 12.2 Higher Order (Dominant Pole)

### 12.2.1 Phase Margin

$$PM = \tan^{-1} \left( \frac{2\zeta}{\sqrt{-2\zeta^2 + \sqrt{1 + 4\zeta^4}}} \right)$$

$$PM \approx 100\zeta \quad \text{for } 0 \leq \zeta \leq 0.6$$

### 12.2.2 Bandwidth

$$\omega_{BW} = \omega_n \sqrt{1 - 2\zeta^2 + \sqrt{2 - 4\zeta^2(1 - \zeta)^2}}$$

$$T_s \approx \frac{4}{\zeta\omega_n} = \frac{4}{\zeta\omega_{BW}} \omega_n \sqrt{1 - 2\zeta^2 + \sqrt{2 - 4\zeta^2(1 - \zeta)^2}}$$

### 12.2.3 Crossover Frequency

$$\omega_c = \omega_{BW} \frac{\sqrt{\sqrt{1 + 4\zeta^4} - 2\zeta^2}}{\sqrt{1 - 2\zeta^2 + \sqrt{2 - 4\zeta^2(1 - \zeta)^2}}} \approx 0.635\omega_{BW}$$

$$\omega_c \approx 0.5 \cdot \omega_{BW} \quad \omega_c \leq \omega_{BW} \leq 2\omega_c$$

## 13 Stability

### 13.1 Internal Stability

A system is **Internally Stable** if  $\forall x_0 \in \mathbb{R}$  the solution of  $\dot{x} = Ax$  with I.C.  $x(0) = x_0$  is **bounded**.

### 13.2 Asymptotic Stability (AS)

A system is **asymptotically stable** if  $\forall x_0 \in \mathbb{R}^n$  with I.C.  $x(0) = x_0$ ,  $x(t) \rightarrow 0$  as  $t \rightarrow \infty$ .

$$\dot{x} = Ax \quad X(s) = \frac{\text{Adj}(sI - A)x_0}{\det(sI - A)}$$

AS if all poles (all eigenvalues) of  $X(s)$  in OLHP.

### 13.3 Input/Output Stability (BIBO)

A system is **BIBO Stable** if for any bounded input  $x(t)$ , the output  $y(t)$  is also bounded.

$$Y(s) = G(s)U(s) \quad G(s) = \frac{C \text{Adj}(sI - A)B + D}{\det(sI - A)}$$

BIBO Stable if all poles of  $G(s)$  in OLHP.

### 13.4 Routh Array

# of sign variations = # of roots with real part  $< 0$

$$b_1 = \frac{-1}{a_{n-1}} \det \begin{bmatrix} 1 & a_{n-2} \\ a_{n-1} & a_{n-3} \end{bmatrix} \quad b_2 = \frac{-1}{a_{n-1}} \det \begin{bmatrix} 1 & a_{n-4} \\ a_{n-1} & a_{n-5} \end{bmatrix}$$

$$c_1 = \frac{-1}{b_1} \det \begin{bmatrix} a_{n-1} & a_{n-3} \\ b_1 & b_2 \end{bmatrix} \quad c_2 = \frac{-1}{b_1} \det \begin{bmatrix} a_{n-1} & a_{n-5} \\ b_1 & b_3 \end{bmatrix}$$

## 14 Basic (Standard) Control Problem

$$E(s) = \frac{1}{1 + CG} R(s) + \frac{-G}{1 + CG} D(s) = E_R \cdot R + E_D \cdot D$$

$$U(s) = \frac{C}{1 + CG} R(s) + \frac{1}{1 + CG} D(s)$$

Closed Loop System BIBO Stable if  $G4$  BIBO Stable.

### 14.1 $G4$ Stability

1. No illegal pole/zero cancellations in  $CG$
2. Zeroes of  $1 + CG$  in OLHP

### 14.2 Type

A TF has type  $l$  if it has exactly  $l$  poles at 0. Suppose  $R(s)$  has type  $K$ . If  $CG$  has type  $K - 1$ , then  $e(\infty)$  is nonzero, finite. If  $CG$  has type  $K - 2$ , then  $e(\infty)$  is unbounded.

## 15 Internal Model Principle (IMP)

$R(s), D(s)$  rational, strictly proper. Then  $e(t) \rightarrow 0$  iff 1.  $G4$  BIBO Stable 2. Poles of  $R$  are also poles of  $CG$  ( $CG$  type  $K_R$ ) 3. Poles of  $D$  are also poles of  $C$  ( $C$  type  $K_D$ )



## 16 Controllers

### 16.1 PD Controllers

Not physically implementable (unless  $\dot{y}(t)$  sensor).

$$C(s) = K(T_d \cdot s + 1) \quad \leftrightarrow \quad u(t) = Ke(t) + KT_d\dot{e}(t)$$

Use to increase the PM (by a max of  $\pi/2$  by placing  $\frac{1}{T_d}$  before the  $\omega_c$ )

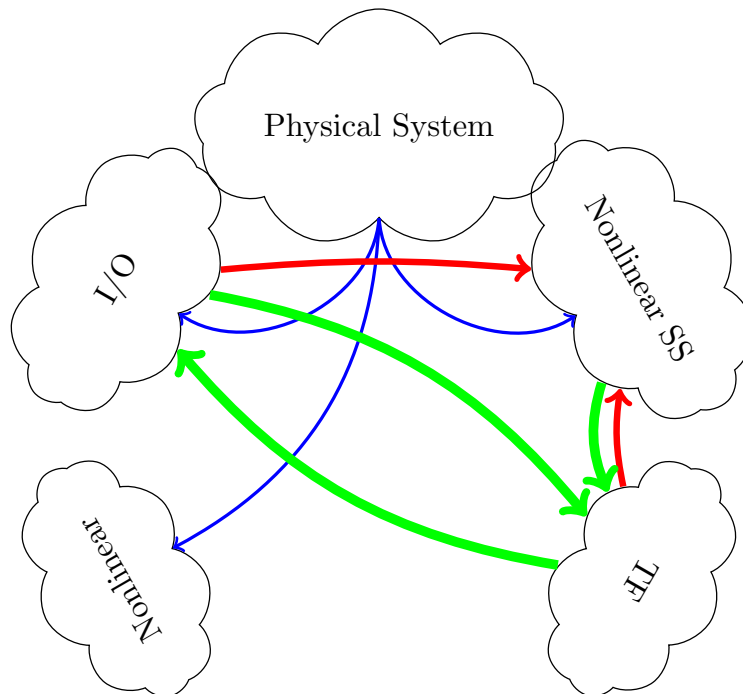


Figure 1: Blue arrows represent conversions from a physical model to a mathematical model. Green arrows represent unique conversions between mathematical models of systems. Red arrows represent non-unique conversions between mathematical models.