

Lecture 10: Pole Placement and Robust Tracking

10.1 Control Configuration

- (a) Open Loop, as shown in Figure 9.1(a).
- (b) Unity Feedback Configuration, as shown in Figure 9.1(b).

$$\hat{u}(s) = C(s)[p\hat{r}(s) - \hat{y}(s)] \quad (1)$$

- (c) Controller-Estimator or Plant-Input-Output-Feedback Configuration, as shown in Figure 9.1(c).

$$\hat{u}(s) = \frac{1}{1 + C_1(s)}\hat{r}(s) - \frac{C_2(s)}{1 + C_1(s)}\hat{y}(s) \quad (2)$$

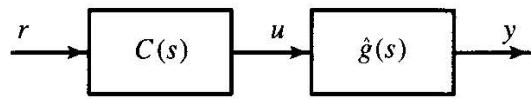
- (d) Two-Parameter Configuration, as shown in Figure 9.1(d).

$$\hat{u}(s) = C_1(s)\hat{r}(s) - C_2(s)\hat{y}(s) \quad (3)$$

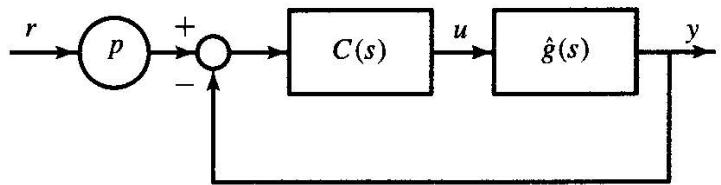
Terminology:

Minimum-phase zeros: Zeros with negative real parts

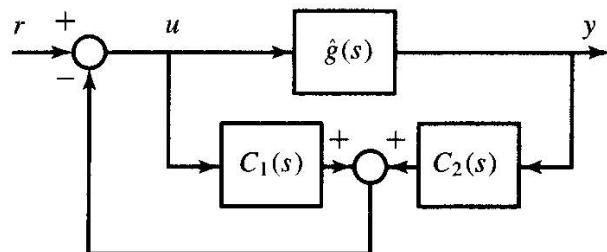
Nonminimum-phase zeros: Zeros with zero or positive real parts



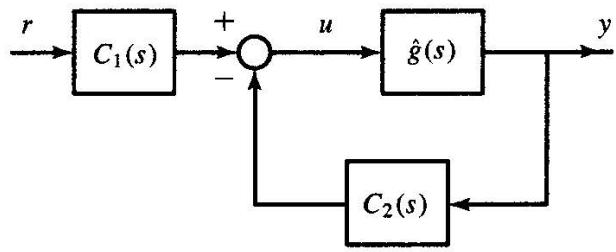
(a)



(b)



(c)



(d)

Figure 9.1 Control configurations.

Compensator Equation

$$A(s)D(s) + B(s)N(s) = F(s) \quad (4)$$

where $D(s)$, $N(s)$ and $F(s)$ are given polynomials and $A(s)$ and $B(s)$ are unknown polynomials.

This is known as *Diophantine Equation*.

Theorem 10.1 Given polynomials $D(s)$ and $N(s)$, polynomial solutions $A(s)$ and $B(s)$ exist in (4) for any polynomial $F(s)$ iff $D(s)$ and $N(s)$ are coprime.

Proof. Necessity. Suppose $D(s)$ and $N(s)$ are not coprime and contain the same factor $(s + a)$. Then $(s + a)$ will appear in $F(s)$. Thus, if $F(s)$ does not contain this factor, then no solution in (4).

Sufficiency. If $D(s)$ and $N(s)$ are coprime, then $\exists \bar{A}(s)$ and $\bar{B}(s)$ such that

$$\bar{A}(s)D(s) + \bar{B}(s)N(s) = 1 \quad (5)$$

For any $F(s)$, this implies

$$[F(s)\bar{A}(s)] D(s) + [F(s)\bar{B}(s)] N(s) = F(s) \quad (6)$$

This completes the proof. □

10.2 Unity-Feedback Configuration

Consider Fig 9.1(b), unity feedback configuration. Let $\hat{g}(s) = N(s)/D(s)$ and $C(s) = B(s)/A(s)$.

Then the transfer function from r to y is

$$\hat{g}_c(s) = \frac{pC(s)\hat{g}(s)}{1 + C(s)\hat{g}(s)} = \frac{p\frac{B(s)}{A(s)}\frac{N(s)}{D(s)}}{1 + \frac{B(s)}{A(s)}\frac{N(s)}{D(s)}} = \frac{pB(s)N(s)}{A(s)D(s) + B(s)N(s)} \quad (11)$$

Let

$$A(s)D(s) + B(s)N(s) = F(s) \quad (12)$$

Then, the closed-loop system has poles as in $F(s) = 0$ and zeros as in $B(s)N(s) = 0$. (Note: extra zeros as in $B(s) = 0$.)

If $D(s)$ and $N(s)$ are coprime, then, according to Theorem 10.1, we can solve (12) for $A(s)$ and $B(s)$ for any given $F(s)$

Of course, no guarantee that $C(s) = B(s)/A(s)$ be proper!

How to solve (12)? Let

$$\begin{aligned}D(s) &= D_0 + D_1s + D_2s^2 + \cdots + D_ns^n, & D_n \neq 0 \\N(s) &= N_0 + N_1s + N_2s^2 + \cdots + N_ns^n \\A(s) &= A_0 + A_1s + A_2s^2 + \cdots + A_ms^m \\B(s) &= B_0 + B_1s + B_2s^2 + \cdots + B_ms^m \\F(s) &= F_0 + F_1s + F_2s^2 + \cdots + F_{n+m}s^{n+m}\end{aligned}$$

Substituting the above into (12), we have

$$\begin{aligned}A_0D_0 + B_0N_0 &= F_0 \\A_0D_1 + B_0N_1 + A_1D_0 + B_1N_0 &= F_1 \\A_0D_2 + B_0N_2 + A_1D_1 + B_1N_1 + A_2D_0 + B_2N_0 &= F_2 \\\vdots \\A_{m-1}D_n + B_{m-1}N_n + A_mD_{n-1} + B_mN_{n-1} &= F_{n+m-1} \\A_mD_n + B_mN_n &= F_{n+m}\end{aligned}$$

i.e.

$$\begin{bmatrix} A_0 & B_0 & A_1 & B_1 & \cdots & A_m & B_m \end{bmatrix} \mathbf{S}_m = \begin{bmatrix} F_0 & F_1 & \cdots & F_{n+m} \end{bmatrix} \quad (13)$$

with $[2(m+1)] \times (n+m+1)$ matrix

$$\mathbf{S}_m = \left[\begin{array}{cccccc|ccc} D_0 & D_1 & \cdots & D_m & \cdots & D_n & 0 & \cdots & 0 \\ N_0 & N_1 & \cdots & N_m & \cdots & N_n & 0 & \cdots & 0 \\ \hline 0 & D_0 & \cdots & D_{m-1} & \cdots & D_{n-1} & D_n & \cdots & 0 \\ 0 & N_0 & \cdots & N_{m-1} & \cdots & N_{n-1} & N_n & \cdots & 0 \\ \vdots & \vdots & & \vdots & & \vdots & \vdots & & \vdots \\ 0 & 0 & \cdots & D_0 & \cdots & D_{n-m} & D_{n-m+1} & \cdots & D_n \\ 0 & 0 & \cdots & N_0 & \cdots & N_{n-m} & N_{n-m+1} & \cdots & N_n \end{array} \right] \quad (14)$$

Note that (13) has a solution for *any* $F(s)$ iff \mathbf{S}_m is of full column rank (but the solution may not be unique).

A necessary condition for \mathbf{S}_m to have full column rank is that \mathbf{S}_m is square or has more rows than columns:

$$2(m + 1) \geq n + m + 1, \quad \text{i.e.} \quad m \geq n - 1$$

If $m < n - 1$, (13) may have solution for some $F(s)$, but not for all $F(s)$. Hence, arbitrary pole assignment is not possible if the compensator $C(s)$ is of degree less than $n - 1$.

If $m = n - 1$, then \mathbf{S}_{n-1} is $(2n) \times (2n)$, and is just the transpose of the Sylvester resultant in Chapter 6. Recall that \mathbf{S}_{n-1} is nonsingular iff $D(s)$ and $N(s)$ are coprime.

Note: if $D(s)$ and $N(s)$ are coprime, then \mathbf{S}_n is $2(n + 1) \times (2n + 1)$ and is also of full column rank $2n + 1$, because $D_n \neq 0$.

In fact, if $D(s)$ and $N(s)$ are coprime, then \mathbf{S}_m is of full column rank $n+m+1$ for all $m \geq n - 1$.

Theorem 10.2 Consider the unity-feedback system in Fig 9.1(b). The plant is described by

$$\hat{g}(s) = \frac{N(s)}{D(s)}, \quad \text{with } \begin{cases} N(s) \text{ and } D(s) \text{ coprime,} \\ \deg N(s) < \deg D(s) = n \end{cases}$$

Let $m \geq n - 1$. Then for any polynomial $F(s)$ of degree $(n + m)$, there exists a proper compensator

$$C(s) = \frac{B(s)}{A(s)}, \quad \text{with } \deg B(s) \leq \deg A(s) = m$$

such that the overall transfer function is

$$\hat{g}_c(s) = \frac{pB(s)N(s)}{A(s)D(s) + B(s)N(s)} = \frac{pB(s)N(s)}{F(s)}$$

Furthermore, $C(s)$ can be obtained by solving (13), and the solution $A(s)$ and $B(s)$ is unique if $m = n - 1$, and not unique if $m > n - 1$.

Proof. Existence of solution follows from the fact that \mathbf{S}_m is of full column rank $n + m + 1$ if $m \geq n - 1$.

Only need to show $C(s)$ is proper, i.e., $A_m \neq 0$.

Note that $\deg N(s) < \deg D(s) = n$. So $N_n = 0$.

Then the last eqn of (13) becomes

$$A_m D_n = F_{n+m}$$

As $\deg F(s) = m$, i.e., $F_{n+m} \neq 0$, hence, $A_m \neq 0$.

When $m = n - 1$, \mathbf{S}_{n-1} is square and nonsingular, so the solution exists and is unique.

When $m > n - 1$, the solution is not unique. □

Regulation and Tracking

Regulation is achieved if all poles of $\hat{g}_c(s)$ (i.e., roots of $F(s) = 0$) have negative real parts. Then, for step reference $\hat{r}(t) = a/s$, we have

$$\hat{y}(s) = \hat{g}_c(s)\hat{r}(s) = \hat{g}_c(s)\frac{a}{s}$$

and

$$\lim_{t \rightarrow \infty} y(t) = \lim_{s \rightarrow 0} s\hat{y}(s) = \hat{g}(0)a$$

As $\hat{g}(0) = p \frac{B_0 N_0}{F_0}$, so tracking of a step input can be achieved by letting

$$p = \frac{F_0}{B_0 N_0} \tag{15}$$

Robust Tracking and Disturbance Rejection

The tracking solution in (15) is affected by variation of plant and controller parameters N_0 , B_0 and p .

Robust tracking and disturbance rejection problem: to make output y track asymptotically a class of reference signal r even with the presence of disturbance and plant parameter variations.

Consider the system shown in Fig. 9.2.

First, let $r(\infty) \neq 0$ and $w(\infty) \neq 0$. Otherwise, closed-loop stability implies asymptotic tracking — a trivial case.

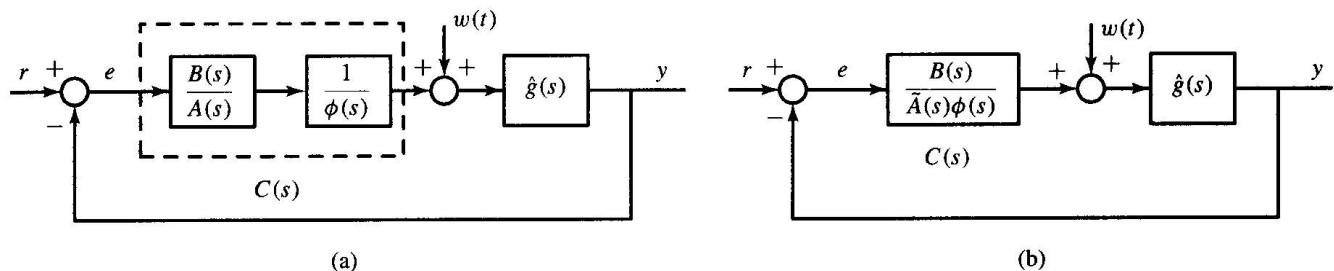


Figure 9.2 Robust tracking and disturbance rejection.

We must have some knowledge on $r(t)$ and $w(t)$. Assume that

$$\hat{r}(s) = \mathcal{L}[r(t)] = \frac{N_r(s)}{D_r(s)}, \quad \hat{w}(s) = \mathcal{L}[w(t)] = \frac{N_w(s)}{D_w(s)} \quad (20)$$

where $D_r(s)$ and $D_w(s)$ are known, but $N_r(s)$ and $N_w(s)$ unknown.

For example, $r(t) = a$ ($t \geq 0$) (step input) and $w(t) = b + c \sin(\omega_0 t + d)$ with a, b, c, d unknown but ω_0 known. Note that $\hat{r}(s) = a/s$ and $\hat{w}_0(s) = \frac{N_w(s)}{s^2 + \omega_0^2}$.

Theorem 10.3 Consider the unity-feedback system in Fig. 9.2(a) with a strictly proper plant transfer function $\hat{g}(s) = N(s)/D(s)$, with $D(s)$ and $N(s)$ coprime. The reference signal $r(t)$ and disturbance $w(t)$ are modeled as $\hat{r}(s) = N_r(s)/D_r(s)$ and $\hat{w}(s) = N_w(s)/D_w(s)$. Let $\phi(s)$ be the least common denominator of the unstable poles of $\hat{r}(s)$ and $\hat{w}(s)$. If no root of $\phi(s)$ is a zero of $\hat{g}(s)$, then there exists a proper compensator such that the overall system will track $r(t)$ and reject $w(t)$, both asymptotically and robustly.

Proof. If no root of $\phi(s)$ is a zero of $\hat{g}(s) = N(s)/D(s)$, then $D(s)\phi(s)$ and $N(s)$ are coprime. Thus, $\exists C(s) = B(s)/A(s)$ such that

$$A(s)D(s)\phi(s) + B(s)N(s) = F(s)$$

with $F(s)$ having any desired roots (in particular, stable roots).

Claim: The following compensator, as in Fig 9.2(a), achieves the design.

$$C(s) = \frac{B(s)}{A(s)\phi(s)}$$

First,

$$\hat{g}_{yw}(s) = \frac{\frac{N(s)}{D(s)}}{1 + \frac{B(s)}{A(s)\phi(s)} \frac{N(s)}{D(s)}} = \frac{N(s)A(s)\phi(s)}{F(s)}$$

Thus, output due to $w(t)$ is

$$\hat{y}_w(s) = \hat{g}_{yw}(s)\hat{w}(s) = \frac{N(s)A(s)\phi(s)}{F(s)} \frac{N_w(s)}{D_w(s)} \quad (21)$$

Note: All unstable roots of $D_w(s)$ are canceled by $\phi(s)$, all poles of $\hat{y}_w(s)$ are stable. Hence,

$$\lim_{t \rightarrow \infty} y_w(t) = 0, \quad \text{i.e., disturbance rejection.}$$

Now look at $y(t)$ due to $r(t)$:

$$\begin{aligned}
 \hat{y}_r(s) &= \hat{g}_{yr}(s)\hat{r}(s) = \frac{B(s)N(s)}{A(s)D(s)\phi(s) + B(s)N(s)}\hat{r}(s) \\
 \implies \hat{e}(s) &\stackrel{\triangle}{=} \hat{r}(s) - \hat{y}_r(s) = [1 - \hat{g}_{yr}(s)]\hat{r}(s) \\
 &= \frac{A(s)D(s)\phi(s)}{F(s)} \frac{N_r(s)}{D_r(s)}
 \end{aligned} \tag{22}$$

Again all unstable roots of $D_r(s)$ are canceled by $\phi(s)$, we have,

$$\lim_{t \rightarrow \infty} e(t) = \lim_{s \rightarrow 0} s\hat{e}(s) = 0, \text{ i.e., } \lim_{t \rightarrow \infty} [y_r(t) - r(t)] = 0$$

Hence

$$y(t) = y_w(t) + y_r(t), \quad \text{and} \quad \lim_{t \rightarrow \infty} [y(t) - r(t)] = 0$$

i.e., tracking and disturbance rejection. This is achieved so long as the system is BIBO, i.e., $A(s)D(s)\phi(s) + B(s)N(s)$ is stable. \square

No pole-zero cancelation in system transfer function! Hence, can be implemented in practice.

Note: $1/\phi(s)$ captures all the unstable poles of $\hat{r}(s)$ and $\hat{w}(s)$. The insertion of $1/\phi(s)$ in the loop is called the *internal model principle*.

Since $B(s)/A(s)$ is proper, the compensator

$$C(s) = \frac{B(s)}{A(s)\phi(s)}$$

is always strictly proper.

When $\deg A(s) = m = n - 1$, the solution $A(s)$ and $B(s)$ is unique.

However, when $\deg A(s) = m \geq n$, the solution is not unique. We can use the extra parameter to embed the internal model $\phi(s)$ into $C(s) = B(s)/A(s)$ directly. See Section 9.2.3 of CT Chen for details.

Example 10.1 Consider the plant $\hat{g}(s) = \frac{s-2}{s^2-1}$. Design a unity-feedback system with a set of desired poles to track robustly any step reference input.

Introduce the internal model $1/\phi(s) = 1/s$. Then,

$$A(s)D(s)\phi(s) + B(s)N(s) = F(s)$$

Note that $\tilde{D}(s) = D(s)\phi(s) = (s^2 - 1)s = 0 - s + 0s^2 + s^3$ and $N(s) = -2 + s + 0s^2 + 0s^3$.

Since $\tilde{D}(s)$ is of degree 3, select $A(s)$ and $B(s)$ to have degree 2. Hence, $F(s)$ has degree 5. Select the five desired poles at -2 , $-2 \pm j1$, and $-1 \pm j2$. Then,

$$F(s) = (s+2)(s^2+4s+5)(s^2+2s+5) = S^5 + 8s^4 + 30s^3 + 66s^2 + 85s + 50$$

Form

$$[A_0 \ B_0 \ A_1 \ B_1 \ A_2 \ B_2] \begin{bmatrix} 0 & -1 & 0 & 1 & 0 & 0 \\ -2 & 1 & 0 & 0 & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & -1 & 0 & 1 & 0 \\ 0 & -2 & 1 & 0 & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & -1 & 0 & 1 \\ 0 & 0 & -2 & 1 & 0 & 0 \end{bmatrix} = [50 \ 85 \ 66 \ 30 \ 8 \ 1]$$

which gives the solution

$$[A_0 \ B_0 \ A_1 \ B_1 \ A_2 \ B_2] = [127.3 \ -25 \ 0 \ -118.7 \ 1 \ -96.3]$$

Thus, the compensator is

$$C(s) = \frac{B(s)}{A(s)\phi(s)} = \frac{-96.3s^2 - 118.7s - 25}{(s^2 + 127.3)s}$$

Embedding Internal Models

Consider

$$A(s)D(s) + B(s)N(s) = F(s)$$

If $\deg[D(s)] = n$ and $\deg[A(s)] = n - 1$, then the solution $A(s)$ and $B(s)$ is unique. If we increase $\deg[A(s)]$ to n , then there is one free parameter we can select to include an internal model in the compensator.

Example 10.2 Consider the plant in Example 10.1. Since $\deg[D(s)] = 2$, select $\deg[A(s)] = 2$. Then, $F(s)$ has degree 4 and can be selected as

$$F(s) = (s^2 + 4s + 5)(s^2 + 2s + 5) = s^4 + 6s^3 + 18s^2 + 30s + 25$$

Form

$$[A_0 \ B_0 \ A_1 \ B_1 \ A_2 \ B_2] \begin{bmatrix} -1 & 0 & 1 & 0 & 0 \\ -2 & 1 & 0 & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & -1 & 0 & 1 & 0 \\ 0 & -2 & 1 & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & -1 & 0 & 1 \\ 0 & 0 & -2 & 1 & 0 \end{bmatrix} = [25 \ 30 \ 18 \ 6 \ 1]$$

For the proper compensator

$$C(s) = \frac{B_0 + B_1 s + B_2 s^2}{A_0 + A_1 s + A_2 s^2}$$

to contain $1/s$ as a factor, require $A_0 = 0$. Solving the above equation, we get

$$[A_0 \ B_0 \ A_1 \ B_1 \ A_2 \ B_2] = [0 \ -12.5 \ 34.8 \ -38.7 \ 1 \ -28.8]$$

and the compensator

$$C(s) = \frac{B(s)}{A(s)} = \frac{-28.8s^2 - 38.7s - 12.5}{s^2 + 34.8s}$$

which can achieve robust tracking and has one degree less than the one obtained in Example 10.1.

Example 10.3 Consider the unity-feedback system in Fig. 9.2(b) with $\hat{g}(s) = 1/s$. Design a proper compensator $C(s) = B(s)/A(s)$ so that the system will track asymptotically any step reference input and reject disturbance $w(t) = a\sin(2t + \theta)$ with unknown a and θ .

To reject the disturbance, $A(s)$ must contain the disturbance model $(s^2 + 4)$. For reference input, since the plant already contains s , we do not need to include it in $A(s)$. Consider

$$A(s)D(s) + B(s)N(s) = F(s)$$

Let

$$A(s) = \tilde{A}_0(s^2 + 4), \quad B(s) = B_0 + B_1s + B_2s^2$$

Define

$$\tilde{D}(s) = D(s)(s^2 + 4) = s(s^2 + 4) = \tilde{D}_0 + \tilde{D}_1s + \tilde{D}_2s^2 + \tilde{D}_3s^3$$

Then, $\tilde{A}_0\tilde{D}(s) + B(s)N(s) = F(s)$ leads to

$$\begin{bmatrix} \tilde{A}_0 & B_0 & B_1 & B_2 \end{bmatrix} \begin{bmatrix} \tilde{D}_0 & \tilde{D}_1 & \tilde{D}_2 & \tilde{D}_3 \\ N_0 & N_1 & 0 & 0 \\ 0 & N_0 & N_1 & 0 \\ 0 & 0 & N_0 & N_1 \end{bmatrix} = [F_0 \ F_1 \ F_2 \ F_3]$$

If we set

$$F(s) = (s+2)(s^2 + 2s + 2) = s^3 + 4s^2 + 6s + 4$$

Then,

$$\begin{bmatrix} \tilde{A}_0 & B_0 & B_1 & B_2 \end{bmatrix} \begin{bmatrix} 0 & 4 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} = [4 \ 6 \ 4 \ 1]$$

leading to the solution [1 4 2 4]. The compensator is

$$C(s) = \frac{4s^2 + 2s + 4}{s^2 + 4}$$

A Servomechanism Problem: A State Space Analysis

Consider

$$\begin{aligned}\dot{x}(t) &= \mathbf{A}x(t) + \mathbf{B}u(t) + \mathbf{E}w(t), & x(t_0) = x_0 \\ y(t) &= \mathbf{C}x(t) + \mathbf{F}w(t)\end{aligned}\tag{23}$$

where $u(t) \in \mathbf{R}^m$ is the control input, $w(t) \in \mathbf{R}^{q \times 1}$ is the disturbance signal and $y(t) \in \mathbf{R}^p$ is the output.

Assume $p = m$ (i.e. same # of input and output).

Objective: Use output feedback to make output track a reference input with zero steady-state error in face of unknown constant disturbance, and the coefficients of the characteristic polynomial are arbitrarily assignable. This is called a *servomechanism problem*.

Strategy: Use observer to get an output feedback control law.

The disturbance signal $w(t)$ is described by

$$\dot{w}(t) = \mathbf{A}_w w(t)$$

where \mathbf{A}_w has only eigenvalues with non-negative real parts.

Augmented system

$$\begin{bmatrix} \dot{x}(t) \\ \dot{w}(t) \end{bmatrix} = \begin{bmatrix} \mathbf{A} & \mathbf{E} \\ \mathbf{0} & \mathbf{A}_w \end{bmatrix} \begin{bmatrix} x(t) \\ w(t) \end{bmatrix} + \begin{bmatrix} \mathbf{B} \\ \mathbf{0} \end{bmatrix} u(t)$$

$$y(t) = [\mathbf{C} \quad \mathbf{F}] \begin{bmatrix} x(t) \\ w(t) \end{bmatrix} \quad (24)$$

Construct a full-order observer for (24):

$$\begin{bmatrix} \dot{\hat{x}}(t) \\ \dot{\hat{w}}(t) \end{bmatrix} = \begin{bmatrix} \mathbf{A} & \mathbf{E} \\ \mathbf{0} & \mathbf{A}_w \end{bmatrix} \begin{bmatrix} \hat{x}(t) \\ \hat{w}(t) \end{bmatrix} + \begin{bmatrix} \mathbf{B} \\ \mathbf{0} \end{bmatrix} u(t) + \begin{bmatrix} \mathbf{H}_1 \\ \mathbf{H}_2 \end{bmatrix} [y(t) - \hat{y}(t)]$$

$$\hat{y}(t) = [\mathbf{C} \quad \mathbf{F}] \begin{bmatrix} \hat{x}(t) \\ \hat{w}(t) \end{bmatrix} \quad (25)$$

Let the error signals be $e_x(t) = x(t) - \hat{x}(t)$ and $e_w(t) = w(t) - \hat{w}$, and the error dynamics is given by

$$\begin{bmatrix} \dot{e}_x(t) \\ \dot{e}_w(t) \end{bmatrix} = \begin{bmatrix} \mathbf{A} - \mathbf{H}_1\mathbf{C} & \mathbf{E} - \mathbf{H}_1\mathbf{F} \\ -\mathbf{H}_2\mathbf{C} & \mathbf{A}_w - \mathbf{H}_2\mathbf{F} \end{bmatrix} \begin{bmatrix} e_x(t) \\ e_w(t) \end{bmatrix} \quad (26)$$

Consider the feedback using the estimated state

$$u(t) = \mathbf{K}_1 \hat{x}(t) + \mathbf{K}_2 \hat{w}(t) + \mathbf{N} r(t) \quad (27)$$

The closed-loop system is given by

$$\begin{aligned} \begin{bmatrix} \dot{x}(t) \\ \dot{\hat{x}}(t) \\ \dot{\hat{w}}(t) \end{bmatrix} &= \begin{bmatrix} \mathbf{A} & \mathbf{B}\mathbf{K}_1 & \mathbf{B}\mathbf{K}_2 \\ \mathbf{H}_1\mathbf{C} & \mathbf{A} + \mathbf{B}\mathbf{K}_1 - \mathbf{H}_1\mathbf{C} & \mathbf{E} + \mathbf{B}\mathbf{K}_2 - \mathbf{H}_1\mathbf{F} \\ \mathbf{H}_2\mathbf{C} & -\mathbf{H}_2\mathbf{C} & \mathbf{A}_w - \mathbf{H}_2\mathbf{F} \end{bmatrix} \begin{bmatrix} x(t) \\ \hat{x}(t) \\ \hat{w}(t) \end{bmatrix} \\ &\quad + \begin{bmatrix} \mathbf{BN} \\ \mathbf{BN} \\ \mathbf{0}_{q \times m} \end{bmatrix} r(t) + \begin{bmatrix} \mathbf{E} \\ \mathbf{H}_1\mathbf{F} \\ \mathbf{H}_2\mathbf{F} \end{bmatrix} w(t) \quad (28) \\ y(t) &= [\mathbf{C} \ \mathbf{0} \ \mathbf{0}] \begin{bmatrix} x(t) \\ \hat{x}(t) \\ \hat{w}(t) \end{bmatrix} + \mathbf{F} w(t) \end{aligned}$$

Apply the following variable change

$$\begin{bmatrix} x(t) \\ e_x(t) \\ -\hat{w}(t) \end{bmatrix} = \begin{bmatrix} \mathbf{I}_n & \mathbf{0}_{n \times n} & \mathbf{0}_{n \times q} \\ \mathbf{I}_n & -\mathbf{I}_n & \mathbf{0}_{n \times q} \\ \mathbf{0}_{q \times n} & \mathbf{0}_{q \times n} & -\mathbf{I}_q \end{bmatrix} \begin{bmatrix} x(t) \\ \hat{x}(t) \\ \hat{w}(t) \end{bmatrix}$$

and the closed-loop state equation becomes

$$\begin{aligned} \begin{bmatrix} \dot{x}(t) \\ \dot{e}_x(t) \\ -\dot{\hat{w}}(t) \end{bmatrix} &= \begin{bmatrix} \mathbf{A} + \mathbf{B}\mathbf{K}_1 & -\mathbf{B}\mathbf{K}_1 & -\mathbf{B}\mathbf{K}_2 \\ 0 & \mathbf{A} - \mathbf{H}_1\mathbf{C} & \mathbf{E} - \mathbf{H}_1\mathbf{F} \\ 0 & -\mathbf{H}_2\mathbf{C} & \mathbf{A}_w - \mathbf{H}_2\mathbf{F} \end{bmatrix} \begin{bmatrix} x(t) \\ e_x(t) \\ -\hat{w}(t) \end{bmatrix} \\ &\quad + \begin{bmatrix} \mathbf{BN} \\ \mathbf{0}_{n \times m} \\ \mathbf{0}_{q \times m} \end{bmatrix} r(t) + \begin{bmatrix} \mathbf{E} \\ \mathbf{E} - \mathbf{H}_1\mathbf{F} \\ -\mathbf{H}_2\mathbf{F} \end{bmatrix} w(t) \quad (29) \\ y(t) &= [\mathbf{C} \ \mathbf{0} \ \mathbf{0}] \begin{bmatrix} x(t) \\ \hat{x}(t) \\ \hat{w}(t) \end{bmatrix} + \mathbf{F}w(t) \end{aligned}$$

Because of the block-diagonal structure in (29), the closed-loop characteristic polynomial is given by

$$\det \left[s\mathbf{I} - (\mathbf{A} + \mathbf{B}\mathbf{K}_1) \right] \det \left[\begin{array}{cc} s\mathbf{I} - (\mathbf{A} - \mathbf{H}_1\mathbf{C}) & \mathbf{E} - \mathbf{H}_1\mathbf{F} \\ -\mathbf{H}_2\mathbf{C} & s\mathbf{I} - \mathbf{A}_w + \mathbf{H}_2\mathbf{F} \end{array} \right]$$

which depends only on \mathbf{K}_1 , \mathbf{H}_1 and \mathbf{H}_2 . Notice the controller poles and the observer poles are separated!

Next, we use the additional freedom K_2 and N in the controller to achieve input-output objectives of asymptotic tracking and disturbance rejection. Straightforward calculations show that

$$\begin{aligned}
 Y(s) = & C \left[sI - (A + BK_1) \right]^{-1} BNR(s) \\
 & + C \left[sI - (A + BK_1) \right]^{-1} EW(s) \\
 & - \left[C \left[sI - (A + BK_1) \right]^{-1} BK_1 , \right. \\
 & \quad \left. - C \left[sI - (A + BK_1) \right]^{-1} BK_2 \right] \\
 & \times \begin{bmatrix} sI - (A - H_1C) & -E + H_1F \\ H_2C & sI + H_2F \end{bmatrix}^{-1} \\
 & \times \begin{bmatrix} E - H_1F \\ -H_2F \end{bmatrix} W(s) + FW(s)
 \end{aligned} \tag{30}$$

For step reference and disturbance inputs

$$R(s) = \mathbf{r}_0 \frac{1}{s}, \quad W(s) = \mathbf{w}_0 \frac{1}{s}$$

applying the final value theorem of Laplace Transform and noting that

$$\begin{bmatrix} -(A - H_1 C) & -E + H_1 F \\ H_2 C & H_2 F \end{bmatrix}^{-1} \begin{bmatrix} E - H_1 F \\ -H_2 F \end{bmatrix} = \begin{bmatrix} 0 \\ -I_q \end{bmatrix}$$

we have the following steady-state output

$$\begin{aligned} \lim_{t \rightarrow \infty} y(t) = & -C(A + BK_1)^{-1}BN\mathbf{r}_0 \\ & + [F - C(A + BK_1)^{-1}E \\ & \quad -C(A + BK_1)^{-1}BK_2] \mathbf{w}_0 \end{aligned} \tag{31}$$

Then, if $C(A + BK_1)^{-1}B$ is invertible, then we can choose

$$\begin{aligned} N &= -[C(A + BK_1)^{-1}B]^{-1} \\ K_2 &= NC(A + BK_1)^{-1}E - NF \end{aligned} \tag{33}$$

simplifies (31) to

$$\lim_{t \rightarrow \infty} y(t) = \mathbf{r}_0 \tag{34}$$

Finally, we establish the equivalence between the invertibility of the following matrices (assuming $A + BK_1$ nonsingular):

$$C(A + BK_1)^{-1}B \text{ is invertible} \quad (*)$$

$$\begin{bmatrix} A & B \\ C & 0 \end{bmatrix} \text{ is invertible} \quad (**)$$

$$\begin{bmatrix} A + BK_1 & BN \\ C & 0 \end{bmatrix} \text{ is invertible} \quad (***)$$

The equivalence between $(**)$ and $(***)$ is easily established by noting that N is assumed nonsingular and that

$$\begin{bmatrix} A & B \\ C & 0 \end{bmatrix} \begin{bmatrix} I & 0 \\ K_1 & N \end{bmatrix} = \begin{bmatrix} A + BK_1 & BN \\ C & 0 \end{bmatrix}$$

and the equivalence between $(*)$ and $(***)$ can be established by noting that

$$\begin{bmatrix} I & 0 \\ -C(A + BK_1)^{-1} & I \end{bmatrix} \begin{bmatrix} A + BK_1 & BN \\ C & 0 \end{bmatrix} = \begin{bmatrix} A + BK_1 & BN \\ 0 & -C(A + BK_1)^{-1}BN \end{bmatrix}$$

and the assumption that $A + BK_1$ is invertible, which can be deduced from its stability.

The above development is summarized in the following Theorem 15.9.

Theorem 10.4: Suppose the plant (23) is controllable for $E = 0$, the augmented plant (24) is observable, and the following $(n+m) \times (n+m)$ matrix is invertible

$$\begin{bmatrix} A & B \\ C & 0 \end{bmatrix}$$

Then a linear dynamic output feedback of the form (27) and (25) has the following properties. The gains K_1 , H_1 and H_2 can be chosen such that the closed-loop state equation (28) is exponentially stable with any desired characteristic polynomial coefficients.

Furthermore, the gains in (33) are such that for any constant reference input $r(t) = \mathbf{r}_0$ and constant disturbance $w(t) = \mathbf{w}_0$, the response of the closed-loop state equation satisfies (34). \square