

2. Convex reformulation in the frequency domain

Yang Zheng

Postdoc, Harvard University

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Outline

1. Recap & LQR as as special case
2. External transfer matrix characterization of internal stability
3. Closed-loop parameterization of stabilizing controllers: SLP and IOP
4. Robust stability and it connections with learning-based control
5. Summary

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Linear time-invariant systems

- ▶ State-space model

$$\begin{aligned}\dot{x} &= Ax + B_1w + B_2u, \\ z &= C_1x + D_{11}w + D_{12}u, \\ y &= C_2x + D_{21}w + D_{22}u,\end{aligned}\tag{1}$$

where $x \in \mathbb{R}^n$, $u \in \mathbb{R}^m$, $w \in \mathbb{R}^d$, $y \in \mathbb{R}^p$, $z \in \mathbb{R}^q$ are the state vector, control action, external disturbance, measurement, and regulated output, respectively.

- ▶ Dynamic controller

$$\begin{aligned}\dot{\xi} &= A_k\xi + B_ky, \\ u &= C_k\xi + D_ky,\end{aligned}\tag{2}$$

where $\xi \in \mathbb{R}^{n_k}$ is the internal state of the controller.

Frequency domain

- ▶ Plant model

$$\mathbf{P} = \left[\begin{array}{c|cc} A & B_1 & B_2 \\ \hline C_1 & D_{11} & D_{12} \\ C_2 & D_{21} & D_{22} \end{array} \right] = \begin{bmatrix} \mathbf{P}_{11} & \mathbf{P}_{12} \\ \mathbf{P}_{21} & \mathbf{P}_{22} \end{bmatrix},$$

where $\mathbf{P}_{ij} = C_i(sI - A)^{-1}B_j + D_{ij}$. We refer to \mathbf{P} as the open-loop plant model.

- ▶ Controller $\mathbf{u} = \mathbf{K}\mathbf{y}$, where $\mathbf{K} = C_k(sI - A)^{-1}B_k + D_k$.

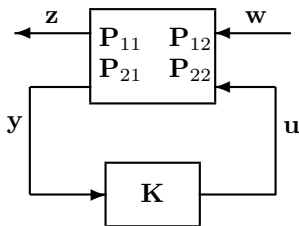


Figure: Interconnection of the plant \mathbf{P} and controller \mathbf{K}

Optimal control

- ▶ General optimal control formulation

$$\begin{aligned} & \min_{\mathbf{K}} f(\mathbf{P}, \mathbf{K}) \\ & \text{subject to } \mathbf{K} \text{ internally stabilizes } \mathbf{P}. \end{aligned} \quad (3)$$

where $f(\mathbf{P}, \mathbf{K})$ defines a certain performance index.

- ▶ Specifically

Frequency-domain formulation

$$\begin{aligned} & \min_{\mathbf{K}} \|\mathbf{T}_{zw}\| \\ & \text{subject to } \mathbf{K} \in \mathcal{C}_{\text{stab}}, \end{aligned}$$

where

$$\mathbf{T}_{zw} = \mathbf{P}_{11} + \mathbf{P}_{12}\mathbf{K}(I - \mathbf{P}_{22}\mathbf{K})^{-1}\mathbf{P}_{21}.$$

State-space formulation

$$\begin{aligned} & \min \left\| \left[\begin{array}{cc|c} A + B_2 D_k C_2 & B_2 C_k & B_1 + B_2 D_k D_{21} \\ \hline B_k C_2 & A_k & B_k D_{21} \\ \hline C_1 + D_{12} D_k C_2 & D_{12} C_k & D_{11} + D_{12} D_k D_{21} \end{array} \right] \right\| \\ & \text{s.t. } \left[\begin{array}{cc} A + B_2 D_k C_2 & B_2 C_k \\ \hline B_k C_2 & A_k \end{array} \right] \text{ is stable.} \end{aligned}$$

LQR as a special case of \mathcal{H}_2 optimal control

Deterministic case

$$\begin{aligned} \min \quad & \int_0^\infty x^\top Q x + u^\top R u \, dt \\ \text{s.t.} \quad & \dot{x} = Ax + Bu \\ & x(0) = x_0, \end{aligned}$$

where $Q \succ 0, R \succ 0$ are weight matrices and $x_0 \in \mathbb{R}^n$.

Both of them are equivalent to

$$\begin{aligned} \min_K \quad & \\ \text{subject to} \quad & \end{aligned}$$

where $B_1 = I, B_2 = B$.

Recap & LQR as as special case

Stochastic case

$$\begin{aligned} \min \quad & \mathbb{E} \left[\lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T x^\top Q x + u^\top R u \, dt \right] \\ \text{s.t.} \quad & \dot{x} = Ax + Bu + w \end{aligned}$$

where $Q \succ 0, R \succ 0$ are weight matrices and $w \sim N(0, I)$.

$$\|\mathbf{T}_{zw}\|_{\mathcal{H}_2}^2$$

$$\dot{x} = Ax + B_1 w + B_2 u$$

$$z = \begin{bmatrix} Q^{\frac{1}{2}} \\ 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ R^{\frac{1}{2}} \end{bmatrix} u$$

$$u = Kx,$$

\mathcal{H}_2 norm

Consider a stable transfer matrix $\mathbf{T} = C(sI - A)^{-1}B$

$$\begin{aligned}\|\mathbf{T}\|_{\mathcal{H}_2}^2 &:= \frac{1}{2\pi} \int_{-\infty}^{\infty} \text{Trace}(T^*(j\omega)T(j\omega)) d\omega \\ &= \int_0^{\infty} \text{Trace}\left((Ce^{At}B)^T(Ce^{At}B)\right) dt\end{aligned}$$

- **Deterministic interpretation:** Let e_k be the standard unit vector and denote the output

$$\dot{x} = Ax, \quad z = Cx, \quad x(0) = Be_k,$$

by $z_k(t)$. Squared \mathcal{H}_2 norm is energy sum of output transients:

$$\sum_{k=1}^m \int_0^{\infty} z_k(t)^T z_k(t) dt = \int_0^{\infty} \text{Trace}\left((Ce^{At}B)^T(Ce^{At}B)\right) dt = \|\mathbf{T}\|_{\mathcal{H}_2}^2.$$

- **Stochastic interpretation:** If w is white noise and $\dot{x} = Ax + Bw, z = Cx$ then

$$\lim_{t \rightarrow \infty} \mathbb{E}\left(z(t)^T z(t)\right) = \|\mathbf{T}\|_{\mathcal{H}_2}^2$$

The squared \mathcal{H}_2 -norm equals the asymptotic variance of output.

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Static state feedback

- ▶ Set of internally stabilizing controllers

$$\mathcal{C}_{\text{stab}} = \left\{ \mathbf{K} \mid \hat{A} := \begin{bmatrix} A + B_2 D_k C_2 & B_2 C_k \\ B_k C_2 & A_k \end{bmatrix} \text{ is stable} \right\},$$

where $\mathbf{K} = C_k(zI - A_k)^{-1}B_k + D_k$.

- ▶ Consider a static state feedback case

$$\begin{array}{l} \dot{x} = Ax + B_2 u \\ u = D_k x. \end{array} \quad \left| \quad \mathcal{C}_{\text{ss}} = \{D_k \in \mathbb{R}^{m \times n} \mid A + B_2 D_k \text{ is stable}\}.$$

- ▶ Lyapunov inequality

$$\begin{aligned} A + B_2 K \text{ is stable} &\iff \exists P \succ 0, (A + B_2 K)^T P + P(A + B_2 K) \prec 0 \\ &\iff \exists X \succ 0, X(A + B_2 K)^T + (A + B_2 K)X \prec 0 \\ &\iff \exists X \succ 0, Y \in \mathbb{R}^{m \times n}, XA^T + YB_2^T + AX + B_2 Y \prec 0 \end{aligned}$$

Therefore, we have

$$\mathcal{C}_{\text{ss}} = \{K = YX^{-1} \mid X \succ 0, Y \in \mathbb{R}^{m \times n}, XA^T + YB_2^T + AX + B_2 Y \prec 0\},$$

External transfer function characterization

Lemma

Consider a transfer matrix $\mathbf{T}(s) = C(sI - A)^{-1}B + D$. If (A, B, C) is detectable and stabilizable, then

$$\mathbf{T}(s) \in \mathcal{RH}_\infty \iff A \text{ is stable.}$$

Two useful facts:

- ▶ The set of stable matrices $\{A \in \mathbb{R}^{n \times n} \mid A \text{ is stable}\}$ is non-convex, but finite-dimensional;
- ▶ The set of stable transfer matrices $\{\mathbf{T}(s) \mid \mathbf{T}(s) \in \mathcal{RH}_\infty\}$ is convex, but infinite-dimensional;

External transfer function characterization

Set of internally stabilizing controllers

$$\mathcal{C}_{\text{stab}} = \left\{ \mathbf{K} \mid \hat{A} := \begin{bmatrix} A + B_2 D_k C_2 & B_2 C_k \\ B_k C_2 & A_k \end{bmatrix} \text{ is stable} \right\},$$

where $\mathbf{K} = C_k(zI - A_k)^{-1}B_k + D_k$.

- ▶ Consider the plant $\mathbf{P}_{22} = C_2(sI - A)^{-1}B_2$,

$$\dot{x} = Ax + B_2u + \delta_x,$$

$$y = C_2x + \delta_y$$

- ▶ A dynamic controller

$$\dot{\xi} = A_k\xi + B_k y$$

$$u = C_k\xi + D_k y + \delta_u.$$

Closed-loop responses from (δ_y, δ_u) to (\mathbf{y}, \mathbf{u}) as

$$\begin{bmatrix} \mathbf{y} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{Y} & \mathbf{W} \\ \mathbf{U} & \mathbf{Z} \end{bmatrix} \begin{bmatrix} \delta_y \\ \delta_u \end{bmatrix},$$

where $\mathbf{Y} = (I - \mathbf{P}_{22}\mathbf{K})^{-1}$, $\mathbf{W} = (I - \mathbf{P}_{22}\mathbf{K})^{-1}\mathbf{P}_{22}$, and

$$\mathbf{U} = \mathbf{K}(I - \mathbf{P}_{22}\mathbf{K})^{-1}, \mathbf{Z} = (I - \mathbf{K}\mathbf{P}_{22})^{-1}.$$

External transfer function characterization

Closed-loop responses from (δ_y, δ_u) to (\mathbf{y}, \mathbf{u}) as

$$\begin{bmatrix} \mathbf{y} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{Y} & \mathbf{W} \\ \mathbf{U} & \mathbf{Z} \end{bmatrix} \begin{bmatrix} \delta_y \\ \delta_u \end{bmatrix},$$

Lemma

The feedback system is internally stable if and only if the transfer matrix from (δ_y, δ_u) to (\mathbf{y}, \mathbf{u}) is stable.

State-space realization of the transfer matrix from (δ_y, δ_u) to (\mathbf{y}, \mathbf{u}) as

$$\left(\begin{bmatrix} \delta_y \\ \delta_u \end{bmatrix} \rightarrow \begin{bmatrix} \mathbf{y} \\ \mathbf{u} \end{bmatrix} \right) = \hat{C}_2(zI - \hat{A})^{-1}\hat{B}_2 + \begin{bmatrix} I & 0 \\ D_k & I \end{bmatrix},$$

where

$$\hat{A} = \begin{bmatrix} A + B_2 D_k C_2 & B_2 C_k \\ B_k C_2 & A_k \end{bmatrix}, \quad \hat{B}_2 = \begin{bmatrix} B_2 D_k & B_2 \\ B_k & 0 \end{bmatrix}, \quad \hat{C}_2 = \begin{bmatrix} C_2 & 0 \\ D_k C_2 & C_k \end{bmatrix}.$$

It remains to prove that (\hat{A}, \hat{B}_2) is stabilizable and (\hat{A}, \hat{C}_2) is detectable.

External transfer function characterization

The stabilizability of (\hat{A}, \hat{B}_2) can be seen from the following fact

$$\begin{bmatrix} A + B_2 D_k C_2 & B_2 C_k \\ B_k C_2 & A_k \end{bmatrix} + \begin{bmatrix} B_2 D_k & B_2 \\ B_k & 0 \end{bmatrix} \begin{bmatrix} -C_2 & F_k \\ F & 0 \end{bmatrix} = \begin{bmatrix} A + B_2 F & B_2 C_k + B_2 D_k F_k \\ 0 & A_k + B_k F_k \end{bmatrix}$$

which will be stable if $A + B_2 F$ and $A_k + B_k F_k$ are stable. The detectability of (\hat{A}, \hat{C}_2) can be shown in a similar way.

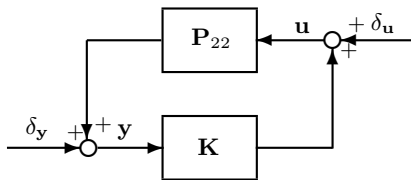


Figure: Input-output stability

External transfer matrix characterization

Look at closed-loop response from (δ_x, δ_y) to \mathbf{x}, \mathbf{u} . It is not difficult to derive that

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{R} & \mathbf{N} \\ \mathbf{M} & \mathbf{L} \end{bmatrix} \begin{bmatrix} \delta_x \\ \delta_y \end{bmatrix},$$

where $\mathbf{R} = (zI - A - B_2\mathbf{K}C_2)^{-1}$, $\mathbf{M} = \mathbf{K}C_2\mathbf{R}$, and

$$\mathbf{U} = \mathbf{R}B_2\mathbf{K}, \quad \mathbf{L} = \mathbf{K}C_2\mathbf{R}B_2\mathbf{K} + \mathbf{K}.$$

Lemma

The feedback system is internally stable if and only if the transfer matrix from (δ_x, δ_y) to (\mathbf{x}, \mathbf{u}) is stable.

A state-space realization of the transfer matrix from (δ_x, δ_y) to (\mathbf{x}, \mathbf{u}) as

$$\left(\begin{bmatrix} \delta_x \\ \delta_y \end{bmatrix} \rightarrow \begin{bmatrix} \mathbf{x} \\ \mathbf{u} \end{bmatrix} \right) = \hat{C}_1(zI - \hat{A})^{-1}\hat{B}_1 + \begin{bmatrix} 0 & 0 \\ 0 & D_k \end{bmatrix},$$

where

$$\hat{B}_1 = \begin{bmatrix} I & B_2D_k \\ 0 & B_k \end{bmatrix}, \quad \hat{C}_1 = \begin{bmatrix} I & 0 \\ D_kC_2 & C_k \end{bmatrix}$$

A summary

Four equivalent statements

- ▶ \mathbf{K} internally stabilizes the plant \mathbf{P} ;

- ▶ $\hat{A} := \begin{bmatrix} A + B_2 D_k C_2 & B_2 C_k \\ B_k C_2 & A_k \end{bmatrix}$ is stable

- ▶ The following closed-loop responses are stable (input-output parameterization)

$$\begin{bmatrix} \mathbf{y} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{Y} & \mathbf{W} \\ \mathbf{U} & \mathbf{Z} \end{bmatrix} \begin{bmatrix} \delta_y \\ \delta_u \end{bmatrix},$$

- ▶ The following closed-loop responses are stable (system-level parameterization)

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{R} & \mathbf{N} \\ \mathbf{M} & \mathbf{L} \end{bmatrix} \begin{bmatrix} \delta_x \\ \delta_y \end{bmatrix},$$

Two special cases

► Open-loop stable plants:

Corollary

If the LTI system is open-loop stable (i.e., A is stable), then $\mathbf{K} \in \mathcal{C}_{stab}$ if and only if $(\delta_y \rightarrow \mathbf{u}) := \mathbf{U} \in \mathcal{RH}_\infty$.

The state-space representation is

$$\mathbf{U} = [D_k C \quad C_k] (zI - \hat{A})^{-1} \begin{bmatrix} BD_k \\ B_k \end{bmatrix} + D_k.$$

Considering the fact that the following matrix

$$\hat{A} + \begin{bmatrix} BD_k \\ B_k \end{bmatrix} [-C \quad F_k] = \begin{bmatrix} A & BC_k + BD_k F_k \\ 0 & A_k + B_k F_k \end{bmatrix},$$

is stable when A and $A_k + B_k F_k$ are stable.

► State feedback

Corollary

If $C = I$, then $\mathbf{K} \in \mathcal{C}_{stab}$ if and only if $(\delta_x \rightarrow \begin{bmatrix} \mathbf{x} \\ \mathbf{u} \end{bmatrix}) := \begin{bmatrix} \mathbf{R} \\ \mathbf{M} \end{bmatrix} \in \mathcal{RH}_\infty$.

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Closed-loop parameterization: IOP

Input-output parameterization

$$\begin{bmatrix} \mathbf{y} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{Y} & \mathbf{W} \\ \mathbf{U} & \mathbf{Z} \end{bmatrix} \begin{bmatrix} \delta_y \\ \delta_u \end{bmatrix},$$

Corollary

If the LTI system is open-loop stable, then we have

$$\mathcal{C}_{stab} = \left\{ \mathbf{K} = \mathbf{U}\mathbf{Y}^{-1} \mid \begin{bmatrix} I & -\mathbf{P}_{22} \end{bmatrix} \begin{bmatrix} \mathbf{Y} \\ \mathbf{U} \end{bmatrix} = I, \mathbf{U} \in \mathcal{RH}_\infty \right\}.$$

\Rightarrow With $\mathbf{K} \in \mathcal{C}_{stab}$, it is not difficult to derive

$$\begin{bmatrix} \mathbf{y} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} (I - \mathbf{P}_{22}\mathbf{K})^{-1} \\ \mathbf{K}(I - \mathbf{P}_{22}\mathbf{K})^{-1} \end{bmatrix} \delta_y.$$

Let us define $\mathbf{Y} = (I - \mathbf{P}_{22}\mathbf{K})^{-1}$ and $\mathbf{U} = \mathbf{K}(I - \mathbf{P}_{22}\mathbf{K})^{-1}$.

Since $\mathbf{K} \in \mathcal{C}_{stab}$, we know that $\mathbf{U} \in \mathcal{RH}_\infty$. Also, by definition, $\mathbf{K} = \mathbf{U}\mathbf{Y}^{-1}$.

Finally, it is very easy to verify that

$$\mathbf{Y} - \mathbf{P}_{22}\mathbf{U} = (I - \mathbf{P}_{22}\mathbf{K})^{-1} - \mathbf{P}_{22}\mathbf{K}(I - \mathbf{P}_{22}\mathbf{K})^{-1} = I.$$

Closed-loop parameterization: IOP

Input-output parameterization

$$\begin{bmatrix} \mathbf{y} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{Y} & \mathbf{W} \\ \mathbf{U} & \mathbf{Z} \end{bmatrix} \begin{bmatrix} \delta_y \\ \delta_u \end{bmatrix},$$

Corollary

If the LTI system is open-loop stable, then we have

$$\mathcal{C}_{stab} = \left\{ \mathbf{K} = \mathbf{U}\mathbf{Y}^{-1} \left| \begin{bmatrix} I & -\mathbf{P}_{22} \end{bmatrix} \begin{bmatrix} \mathbf{Y} \\ \mathbf{U} \end{bmatrix} = I, \mathbf{U} \in \mathcal{RH}_\infty \right. \right\}.$$

\Leftarrow . Given \mathbf{Y} and \mathbf{U} satisfying the condition, we show that $\mathbf{K} = \mathbf{U}\mathbf{Y}^{-1} \in \mathcal{C}_{stab}$. We only need to show the response from δ_y to \mathbf{u} is Stable.

In particular, we have

$$\begin{aligned} \mathbf{u} &= \mathbf{K}(I - \mathbf{P}_{22}\mathbf{K})^{-1}\delta_y \\ &= \mathbf{U}\mathbf{Y}^{-1}(I - \mathbf{P}_{22}\mathbf{U}\mathbf{Y}^{-1})^{-1}\delta_y \\ &= \mathbf{U}\delta_y, \end{aligned}$$

where the last equality used the affine relationship $\mathbf{Y} - \mathbf{P}_{22}\mathbf{U} = I$.

Closed-loop parameterization of stabilizing controllers: SLP and IOP

Closed-loop parameterization: SLP

System-level parameterization

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{R} & \mathbf{N} \\ \mathbf{M} & \mathbf{L} \end{bmatrix} \begin{bmatrix} \delta_x \\ \delta_y \end{bmatrix},$$

Corollary

If $C_2 = I$, then we have

$$\mathcal{C}_{stab} = \left\{ \mathbf{K} = \mathbf{M}\mathbf{R}^{-1} \mid \begin{bmatrix} zI - A & -B_2 \end{bmatrix} \begin{bmatrix} \mathbf{R} \\ \mathbf{M} \end{bmatrix} = I, \mathbf{M}, \mathbf{R} \in \mathcal{RH}_\infty \right\}.$$

\Leftarrow Consider \mathbf{M}, \mathbf{R} satisfying the condition. We define $\mathbf{K} = \mathbf{M}\mathbf{R}^{-1}$, and show this controller $\mathbf{K} \in \mathcal{C}_{stab}$.

It is sufficient to show

$$\mathbf{x} = (sI - A - B_2\mathbf{K})^{-1}\delta_x = (sI - A - B_2\mathbf{M}\mathbf{R}^{-1})^{-1}\delta_x = \mathbf{R}\delta_x$$

$$\mathbf{u} = \mathbf{K}(sI - A - B_2\mathbf{K})^{-1}\delta_x = \mathbf{M}\delta_x$$

General case: Input-output parameterization

$$\begin{bmatrix} I & -\mathbf{P}_{22} \end{bmatrix} \begin{bmatrix} \mathbf{Y} & \mathbf{W} \\ \mathbf{U} & \mathbf{Z} \end{bmatrix} = \begin{bmatrix} I & 0 \end{bmatrix}, \quad (4a)$$

$$\begin{bmatrix} \mathbf{Y} & \mathbf{W} \\ \mathbf{U} & \mathbf{Z} \end{bmatrix} \begin{bmatrix} -\mathbf{P}_{22} \\ I \end{bmatrix} = \begin{bmatrix} 0 \\ I \end{bmatrix}, \quad (4b)$$

$$\mathbf{Y}, \mathbf{U}, \mathbf{W}, \mathbf{Z} \in \mathcal{RH}_\infty. \quad (4c)$$

Theorem (Input-output parameterization)

The set of all internally stabilizing controllers can be represented as

$$\mathcal{C}_{stab} = \{\mathbf{K} = \mathbf{U}\mathbf{Y}^{-1} \mid \mathbf{Y}, \mathbf{U}, \mathbf{W}, \mathbf{Z} \text{ are in the affine subspace (4a)-(4c)}\}.$$

$$\begin{array}{l} \min_{\mathbf{K}} \|\mathbf{P}_{11} + \mathbf{P}_{12}\mathbf{K}(I - \mathbf{P}_{22}\mathbf{K})^{-1}\mathbf{P}_{21}\| \\ \text{subject to } \mathbf{K} \in \mathcal{C}_{stab}, \end{array} \left| \begin{array}{l} \min_{\mathbf{Y}, \mathbf{U}, \mathbf{W}, \mathbf{Z}} \|\mathbf{P}_{11} + \mathbf{P}_{12}\mathbf{U}\mathbf{P}_{21}\| \\ \text{subject to (4a) - (4c).} \end{array} \right.$$

General case: System-level synthesis

$$[sI - A \quad -B_2] \begin{bmatrix} \mathbf{R} & \mathbf{N} \\ \mathbf{M} & \mathbf{L} \end{bmatrix} = [I \quad 0], \quad (5a)$$

$$\begin{bmatrix} \mathbf{R} & \mathbf{N} \\ \mathbf{M} & \mathbf{L} \end{bmatrix} \begin{bmatrix} sI - A \\ -C_2 \end{bmatrix} = \begin{bmatrix} I \\ 0 \end{bmatrix}, \quad (5b)$$

$$\mathbf{R}, \mathbf{M}, \mathbf{N} \in \mathcal{RH}_\infty, \quad \mathbf{L} \in \mathcal{RH}_\infty. \quad (5c)$$

Theorem (System-level parameterization)

For strictly proper plants, the set of all internally stabilizing controllers can be represented as

$$\mathcal{C}_{stab} = \{ \mathbf{K} = \mathbf{L} - \mathbf{M}\mathbf{R}^{-1}\mathbf{N} \mid \mathbf{R}, \mathbf{M}, \mathbf{N}, \mathbf{L} \text{ are in the affine subspace (5a)-(5c)} \}.$$

System-level synthesis

$$\min_{\mathbf{R}, \mathbf{M}, \mathbf{N}, \mathbf{L}} \left\| \begin{bmatrix} C_1 & D_{12} \end{bmatrix} \begin{bmatrix} \mathbf{R} & \mathbf{N} \\ \mathbf{M} & \mathbf{L} \end{bmatrix} \begin{bmatrix} B_1 \\ D_{21} \end{bmatrix} + D_{11} \right\|$$

subject to (5a) – (5c).

Summary

- ▶ Optimal controller synthesis problem

$$\min_{\mathbf{K}} \quad \|\mathbf{P}_{11} + \mathbf{P}_{12}\mathbf{K}(I - \mathbf{P}_{22}\mathbf{K})^{-1}\mathbf{P}_{21}\|$$

subject to $\mathbf{K} \in \mathcal{C}_{\text{stab}}$.

Four equivalent statements

- ▶ \mathbf{K} internally stabilizes the plant \mathbf{P} ;

- ▶ $\hat{A} := \begin{bmatrix} A + B_2 D_k C_2 & B_2 C_k \\ B_k C_2 & A_k \end{bmatrix}$ is stable

- ▶ The following closed-loop responses are stable (IOP)

$$\begin{bmatrix} \mathbf{y} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{Y} & \mathbf{W} \\ \mathbf{U} & \mathbf{Z} \end{bmatrix} \begin{bmatrix} \delta_y \\ \delta_u \end{bmatrix},$$

- ▶ The following closed-loop responses are stable (SLP)

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{R} & \mathbf{N} \\ \mathbf{M} & \mathbf{L} \end{bmatrix} \begin{bmatrix} \delta_x \\ \delta_y \end{bmatrix},$$

Other issues

- ▶ Numerical computation: Finite impulse responses (only work for discrete-time systems)

$$\mathbf{H}(z) = \sum_{k=1}^T H_k \frac{1}{z^k}$$

- ▶ See a Github repository here:
https://github.com/zhengy09/h2_clp.
- ▶ Distributed control $\mathbf{K} \in \mathcal{S}$:

$$\mathbf{K} = \mathbf{U}\mathbf{Y}^{-1} \in \mathcal{S}$$

$$\mathbf{K} = \mathbf{L} - \mathbf{M}\mathbf{R}^{-1}\mathbf{N} \in \mathcal{S}$$

- ▶ State-space realization of these controllers
- ▶ Numerical robustness: the affine constraints can never be exactly satisfied in numerical computation ...

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Robust stability: SLP in the state feedback case

- ▶ Suppose we only have estimation \hat{A} and \hat{B}_2 , where $\|A - \hat{A}\| \leq \epsilon_A$ and $\|B - \hat{B}_2\| \leq \epsilon_B$.
- ▶ How can we design a stabilizing controller for the true system (A, B_2) based on the information (\hat{A}, \hat{B}_2) and ϵ_A, ϵ_B ?
- ▶ We find $\hat{\mathbf{M}}, \hat{\mathbf{R}} \in \mathcal{RH}_\infty$ that satisfies

$$[sI - \hat{A} \quad -\hat{B}_2] \begin{bmatrix} \hat{\mathbf{R}} \\ \hat{\mathbf{M}} \end{bmatrix} = I. \quad (6)$$

- ▶ Then, the controller $\mathbf{K} = \hat{\mathbf{M}}\hat{\mathbf{R}}^{-1}$ stabilizes (\hat{A}, \hat{B}_2) . What happens if we apply $\mathbf{K} = \hat{\mathbf{M}}\hat{\mathbf{R}}^{-1}$ to the true system (A, B_2) ?
- ▶ we have

$$[sI - A \quad -B_2] \begin{bmatrix} \hat{\mathbf{R}} \\ \hat{\mathbf{M}} \end{bmatrix} = I + \mathbf{\Delta},$$

where $\mathbf{\Delta} = \Delta_A \hat{\mathbf{R}} + \Delta_B \hat{\mathbf{M}}$. Then it is not difficult to show that if $\|\Delta\|_\infty < 1$, the controller $\mathbf{K} = \hat{\mathbf{M}}\hat{\mathbf{R}}^{-1}$ stabilizes the true system (A, B_2) as well.

Robust stability: IOP for open-loop stable plants

- ▶ Suppose we have the transfer matrix estimation $\hat{\mathbf{P}}_{22}$, with $\|\mathbf{P}_{22} - \hat{\mathbf{P}}_{22}\|_{\infty} \leq \epsilon$.
- ▶ We find $\hat{\mathbf{Y}}, \hat{\mathbf{U}} \in \mathcal{RH}_{\infty}$ that satisfies

$$\begin{bmatrix} I & -\hat{\mathbf{P}}_{22} \end{bmatrix} \begin{bmatrix} \hat{\mathbf{Y}} \\ \hat{\mathbf{U}} \end{bmatrix} = I.$$

- ▶ Then, the controller $\mathbf{K} = \hat{\mathbf{U}}\hat{\mathbf{Y}}^{-1}$ stabilizes the plant $\hat{\mathbf{P}}_{22}$.
- ▶ For the true plant \mathbf{P}_{22} , we have

$$\begin{bmatrix} I & -\mathbf{P}_{22} \end{bmatrix} \begin{bmatrix} \hat{\mathbf{Y}} \\ \hat{\mathbf{U}} \end{bmatrix} = I + \mathbf{\Delta},$$

where $\mathbf{\Delta} = \Delta\mathbf{P}_{22}$.

- ▶ Then it is not difficult to show that if $\|\Delta\|_{\infty} < 1$, the controller $\mathbf{K} = \hat{\mathbf{U}}\hat{\mathbf{Y}}^{-1}$ stabilizes the true system \mathbf{P}_{22} as well.

Outline

1. Recap & LQR as as special case
2. External transfer matrix characterization of internal stability
3. Closed-loop parameterization of stabilizing controllers: SLP and IOP
4. Robust stability and it connections with learning-based control
5. Summary

A summary

Four equivalent statements

- ▶ \mathbf{K} internally stabilizes the plant \mathbf{P} ;
- ▶ $\hat{A} := \begin{bmatrix} A + B_2 D_k C_2 & B_2 C_k \\ B_k C_2 & A_k \end{bmatrix}$ is stable
- ▶ The following closed-loop responses are stable (input-output parameterization)

$$\begin{bmatrix} \mathbf{y} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{Y} & \mathbf{W} \\ \mathbf{U} & \mathbf{Z} \end{bmatrix} \begin{bmatrix} \delta_y \\ \delta_u \end{bmatrix},$$

- ▶ The following closed-loop responses are stable (system-level parameterization)

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{u} \end{bmatrix} = \begin{bmatrix} \mathbf{R} & \mathbf{N} \\ \mathbf{M} & \mathbf{L} \end{bmatrix} \begin{bmatrix} \delta_x \\ \delta_y \end{bmatrix},$$

Summary

- ▶ Optimal controller synthesis

$$\begin{aligned} & \min_{\mathbf{K}} \quad \|\mathbf{P}_{11} + \mathbf{P}_{12}\mathbf{K}(I - \mathbf{P}_{22}\mathbf{K})^{-1}\mathbf{P}_{21}\| \\ & \text{subject to} \quad \mathbf{K} \in \mathcal{C}_{\text{stab}}, \end{aligned}$$

- ▶ Input-output parameterization

$$\begin{aligned} & \min_{\mathbf{Y}, \mathbf{U}, \mathbf{W}, \mathbf{Z}} \quad \|\mathbf{P}_{11} + \mathbf{P}_{12}\mathbf{U}\mathbf{P}_{21}\| \\ & \text{subject to} \quad (4a) - (4c). \end{aligned}$$

- ▶ System-level parameterization

$$\begin{aligned} & \min_{\mathbf{R}, \mathbf{M}, \mathbf{N}, \mathbf{L}} \quad \left\| \begin{bmatrix} C_1 & D_{12} \end{bmatrix} \begin{bmatrix} \mathbf{R} & \mathbf{N} \\ \mathbf{M} & \mathbf{L} \end{bmatrix} \begin{bmatrix} B_1 \\ D_{21} \end{bmatrix} + D_{11} \right\| \\ & \text{subject to} \quad (5a) - (5c). \end{aligned}$$

Other topics

- ▶ State-space formulation

$$\begin{aligned} \min \quad & \left\| \left[\begin{array}{cc|c} A + B_2 D_k C_2 & B_2 C_k & B_1 + B_2 D_k D_{21} \\ B_k C_2 & A_k & B_k D_{21} \\ \hline C_1 + D_{12} D_k C_2 & D_{12} C_k & D_{11} + D_{12} D_k D_{21} \end{array} \right] \right\| \\ \text{s.t.} \quad & \left[\begin{array}{cc} A + B_2 D_k C_2 & B_2 C_k \\ B_k C_2 & A_k \end{array} \right] \text{ is stable.} \end{aligned}$$

- ▶ How to derive convex reformulation in the state space? This needs to specify the norm of the cost function. We can have LMI formulations (the topic in the next week).
- ▶ How to deal with structural constraint $\mathcal{K} \in \mathcal{S}$?