Robust Control 2018

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Plan of attack:

Today's topic: Evaluate \mathcal{H}_{∞} based robust stability and performance claims.

- \mathcal{H}_{∞} -norm performance specifications.
- The small gain theorem.
- Robust stability specifications.
- Proof of the small gain theorem:
 - Argument principle.
 - Loop transforms.
 - Instability theorems.

Suppose $L(j\omega)$ is given by the following:



Is
$$\|\frac{10(s/20+1)}{s+1}S(s)\|_{\infty} \le 1$$
?

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To check bounded gain, can use the singular value plot or the bounded real lemma.

Bounded real Lemma: Given any $G \in \mathcal{R}^{n \times m}$, the following are equivalent:

(i) $||G||_{\infty} \le 1$.

(ii) For any minimal realisation of *G* there exists a $P \succ 0$ such that

$$\begin{bmatrix} A^T P + PA + C^T C & PB + C^T D \\ B^T P + DC^T & D^T D - I \end{bmatrix} \preceq 0.$$

The Small Gain Theorem



Given $G \in \mathcal{R}^{n \times m}$, the following are equivalent:

- (i) The feedback interconnection of G and Δ is stable for all $\Delta \in \mathcal{R}^{m \times n}$ such that $\|\Delta\|_{\infty} < 1$.
- (ii) $||G||_{\infty} \le 1$.

The Small Gain Theorem

Can be generalized significantly. For example:

- Real rationality of G and Δ can be removed.
- Linearity of Δ can be removed.

\mathcal{H}_{∞} -norm performance specifications are robust stability specifications.

By the small gain theorem, $\|\mathcal{F}_l(P(s), K(s))\|_{\infty} \leq 1$ is equivalent to stability of the following for $\|\Delta\|_{\infty} < 1$:











 $||W(s)S(s)||_{\infty} \leq 1 \implies$ stability for all

 $P_{\Delta}(s) \in \{Q(s) : Q(s) = (I + \Delta(s))^{-1} P(s), \|\Delta(s)\|_{\infty} < 1\}.$

Questions:

- What are the robust stability equivalents of an H_∞-norm specification on each element of the gang of four?
- What about the gang of four stability margin

$$b_{P,C} = \left\| \begin{bmatrix} P \\ I \end{bmatrix} (I - CP)^{-1} \begin{bmatrix} -C & I \end{bmatrix} \right\|_{\infty}^{-1}$$
?

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Exercise

Consider

$$g(s) = \frac{s-4}{(s-2)(s-3)}.$$

Let us evaluate g(s) on a square contour.






























The Argument Principle



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The Argument Principle

Given $g(s) \in \mathcal{R}$ and a closed contour C,

w.n.o.
$$g(s) = Z - P$$

where Z,P are the number of zeros and poles of $g(\boldsymbol{s})$ contained in C.

The Argument Principle

Given $G(s) \in \mathcal{R}^{n \times n}$ and a closed contour C,

w.n.o.
$$\det(G(s)) = Z - P$$

where Z, P are the number of zeros and poles of G(s) contained in C.

We will use the argument principle to show that (ii) implies that

 $\det(I+G(s)\Delta(s))\neq 0$

for all *s* in the closed right half plane.



The Nyquist contour.

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Consider

$$g_{\lambda}(s) = \det \left(I + \lambda G(s) \Delta(s) \right).$$

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Now vary λ from 0 to 1.























Recall that for $C \in \mathbb{C}^{n \times n}$

$$|\det C| = \prod_i \sigma_i(C).$$

Therefore if $\underline{\sigma}(C) > 0$, det $C \neq 0$.

Now recall that $\underline{\sigma}(A+B) \ge \underline{\sigma}(A) - \overline{\sigma}(B)$. Therefore since $\|G\|_{\infty} \le 1$ and $\|\Delta\|_{\infty} < 1$,

$$|g_{\lambda}(j\omega)| \ge 1 - \lambda \overline{\sigma}(G(j\omega)\Delta(j\omega)) > 0.$$

Therefore for every point on the contour, $g_{\lambda}(s) \neq 0$,

$$\implies$$
 w.n.o. $g_0(s) =$ w.n.o. $g_1(s)$.

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Since P = 0, \implies the interconnection is stable.

Exercise

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Can extend the small gain theorem using loop transforms:



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Stable if $\|G(s)(I+G(s))^{-1}\|_{\infty} \leq 1$ and $\|I+\Delta(s)\|_{\infty} < 1$.

- Are we sure $(\tilde{Y}(s), \tilde{U}(s))$ bounded implies that (Y(s), U(s)) is bounded?
- e How can we interpret such loop transforms?

Recall from last time that we used the projective line to understand $G(s)(I + G(s))^{-1}$. Use the chain description:

$$\begin{bmatrix} I & 0 \\ I & I \end{bmatrix} \begin{bmatrix} G(s) \\ I \end{bmatrix}$$

Using the Projective Line

Signal interpretation:

$$\begin{bmatrix} \tilde{Y} \\ \tilde{U} \end{bmatrix} = \begin{bmatrix} I & 0 \\ I & I \end{bmatrix} \begin{bmatrix} Y \\ U \end{bmatrix}, \begin{bmatrix} Y \\ U \end{bmatrix} = \begin{bmatrix} G \\ I \end{bmatrix} U.$$

- The first equation describes the loop transform.
- In the second, U generates the 'behaviour' (use coprime factorisation for unstable G)

Using the Projective Line

Let $M \in \mathcal{R}^{n \times n}$ define a loop transform

$$\begin{bmatrix} \tilde{Y}(s) \\ \tilde{U}(s) \end{bmatrix} = M(s) \begin{bmatrix} Y(s) \\ U(s) \end{bmatrix}$$

Boundedness of $(\tilde{Y}(s), \tilde{U}(s))$ is equivalent to boundedness of (Y(s), U(s)) if and only if $M, M^{-1} \in \mathcal{RH}_{\infty}$.

Small gain of Y(s) = G(s)U(s) is equivalent to

$$\begin{bmatrix} Y(s) \\ U(s) \end{bmatrix}^* \begin{bmatrix} -I & 0 \\ 0 & I \end{bmatrix} \begin{bmatrix} Y(s) \\ U(s) \end{bmatrix} \ge 0$$

Suppose

$$\begin{bmatrix} Y(s) \\ U(s) \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} I & I \\ -I & I \end{bmatrix} \begin{bmatrix} \tilde{Y}(s) \\ \tilde{U}(s) \end{bmatrix}?$$

Small gain is equivalent to:

$$\left(\frac{1}{\sqrt{2}}\begin{bmatrix}I&I\\-I&I\end{bmatrix}\begin{bmatrix}\tilde{Y}(s)\\\tilde{U}(s)\end{bmatrix}\right)^*\begin{bmatrix}-I&0\\0&I\end{bmatrix}\frac{1}{\sqrt{2}}\begin{bmatrix}I&I\\-I&I\end{bmatrix}\begin{bmatrix}\tilde{Y}(s)\\\tilde{U}(s)\end{bmatrix}\geq 0.$$

Small gain is equivalent to:

$$\begin{bmatrix} \tilde{Y}(s) \\ \tilde{U}(s) \end{bmatrix}^* \begin{bmatrix} 0 & I \\ I & 0 \end{bmatrix} \begin{bmatrix} \tilde{Y}(s) \\ \tilde{U}(s) \end{bmatrix} \ge 0.$$

Small gain is equivalent to:

$$\begin{bmatrix} \tilde{Y}(s) \\ \tilde{U}(s) \end{bmatrix}^* \begin{bmatrix} 0 & I \\ I & 0 \end{bmatrix} \begin{bmatrix} \tilde{Y}(s) \\ \tilde{U}(s) \end{bmatrix} \ge 0.$$
$$\implies \int_{-\infty}^{\infty} \tilde{y}^T(t) \tilde{u}(t) dt \ge 0.$$
The Passivity Theorem

What about the transfer function given by the chain description:

$$\frac{1}{\sqrt{2}} \begin{bmatrix} I & I \\ -I & I \end{bmatrix}^{-1} \begin{bmatrix} G(s) \\ I \end{bmatrix}?$$

The Passivity Theorem

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Exercise

The Passivity Theorem

- Small gain of the pair (y, u) is equivalent to passivity of (ỹ, ũ).
- (y, u) are called scattering variables in this context.
- Connections to IQCs, the KYP lemma, chain scattering, ...

Observe that in our proof of the small gain theorem we only used our \mathcal{H}_{∞} -norm bound to bound the largest singular value of *G* on the Nyquist contour.

\mathcal{L}_∞ is the space of essentially bounded measurable functions on the imaginary axis, with norm

$$||f(j\omega)||_{\mathcal{L}_{\infty}} = \operatorname{ess\,sup}_{\omega \in \mathbb{R}} |f(j\omega)|.$$

For real rational functions:

$$||f(j\omega)||_{\mathcal{L}_{\infty}} = \sup_{\omega \in \mathbb{R}} |f(j\omega)|.$$

For matrices of real rational functions:

$$\|F(j\omega)\|_{\mathcal{L}_{\infty}} = \sup_{\omega \in \mathbb{R}} \overline{\sigma}(F(j\omega)).$$

By our previous argument, $\|G(j\omega)\|_{\mathcal{L}_{\infty}} \leq 1$ and $\|\Delta(j\omega)\|_{\mathcal{L}_{\infty}} < 1$ is sufficient to conclude that

w.n.o.
$$\det(I + G\Delta) = 0 = Z - P.$$

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$$\det(I + G\Delta) = 0 = Z - P.$$

 \Longrightarrow that the number of closed loop poles equals the number of open loop poles.

The Small Gain Theorem (2)



Given $G \in \mathcal{R}^{n \times m}$, the following are equivalent:

- (i) The feedback interconnection of *G* and Δ has *P* unstable poles for all $\Delta \in \mathcal{R}^{m \times n}$ such that $\|\Delta\|_{\infty} < 1$.
- (ii) $||G||_{\mathcal{L}_{\infty}} \leq 1$ and G has P unstable poles.

The Small Gain Theorem (2)

- Can broaden applicability with the same projective line arguments.
- Prease and linear requirements can be relaxed.
- Solution Can check $||G||_{\mathcal{L}_{\infty}} \leq 1$ using LMIs.