# LMI, Control design, Robustness Computational Intelligence, Lecture 9

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#### CONTENT

- Robustness check
- S-procedure
- Robust Control Design
- Quadratic stability

#### ROBUSTNESS, 1

Consider a system  $\dot{\mathbf{x}} = \mathbf{A}\mathbf{x}$ , but when you don't know  $\mathbf{A}$  exactly. In other words, you don't know the model exactly. This is not to say that we know nothing about the model, but there is an uncertainty in our knowledge.

One way to model this model with *uncertainty* is given below:

$$\dot{\mathbf{x}} = (\mathbf{A} + \mathbf{F}\Delta\mathbf{E})\mathbf{x} \tag{1}$$

where **F** and **E** are arbitrary matrices, and  $\Delta$  is a *norm-bounded* matrix:  $||\Delta|| \leq 1$ .

We can think of it this way:  $\mathbf{A} + \mathbf{F}\Delta\mathbf{E}$  is the true but unknown model, and the range of all possible models we can expect is bounded by the possible values of  $\Delta$ .

#### Robustness, 2

Lets write the Lyapunov equation for the system (1):

$$\dot{V} = \mathbf{x}^{\top} (\mathbf{A} + \mathbf{F} \Delta \mathbf{E})^{\top} \mathbf{S} \mathbf{x} + \mathbf{x}^{\top} \mathbf{S} (\mathbf{A} + \mathbf{F} \Delta \mathbf{E}) \mathbf{x} \le 0$$
 (2)

Let us introduce a new variable  $\mathbf{w} = \Delta \mathbf{E} \mathbf{x}$ :

$$\dot{V} = \mathbf{x}^{\top} (\mathbf{A}^{\top} \mathbf{S} + \mathbf{S} \mathbf{A}) \mathbf{x} + \mathbf{w}^{\top} \mathbf{F}^{\top} \mathbf{S} \mathbf{x} + \mathbf{x}^{\top} \mathbf{S} \mathbf{F} \mathbf{w} \le 0$$
 (3)

Let us consider  $\mathbf{w}^{\top}\mathbf{w}$ :

$$\mathbf{w}^{\top}\mathbf{w} = \mathbf{x}^{\top}\mathbf{E}^{\top}\Delta\Delta\mathbf{E}\mathbf{x} \le \mathbf{x}^{\top}\mathbf{E}^{\top}\mathbf{E}\mathbf{x}$$
(4)

which is true because  $||\Delta|| \le 1$ . In fact, the only property of the norm that we need here is that the delta inequality (4) holds.

## Robustness, 3

With  $\mathbf{w}^{\top}\mathbf{w} \leq \mathbf{x}^{\top}\mathbf{E}^{\top}\mathbf{E}\mathbf{x}$  we can write:

$$\mathbf{x}^{\top} \mathbf{E}^{\top} \mathbf{E} \mathbf{x} - \mathbf{w}^{\top} \mathbf{w} \ge 0 \tag{5}$$

Which is the same as:

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{w} \end{bmatrix}^{\top} \begin{bmatrix} \mathbf{E}^{\top} \mathbf{E} & 0 \\ 0 & -\mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{w} \end{bmatrix} \ge 0$$
 (6)

The same way we can rewrite (3):

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{w} \end{bmatrix}^{\top} \begin{bmatrix} \mathbf{A}^{\top} \mathbf{S} + \mathbf{S} \mathbf{A} & \mathbf{S} \mathbf{F} \\ \mathbf{F}^{\top} \mathbf{S} & 0 \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{w} \end{bmatrix} \le 0$$
 (7)

which only need to hold while (6) holds.

#### S-PROCEDURE

There is a way to enforce constraint  $\mathbf{z}^{\top} \mathbf{M} \mathbf{z} \leq 0$  for such  $\mathbf{z}$  that  $\mathbf{z}^{\top} \mathbf{N} \mathbf{z} \geq 0$ . This is called *s-procedure*.

#### Theorem (S-procedure)

If 
$$\gamma > 0$$
 and  $\mathbf{M} + \gamma \mathbf{N} \prec 0$  then  $\mathbf{z}^{\top} \mathbf{N} \mathbf{z} \geq 0 \implies \mathbf{z}^{\top} \mathbf{M} \mathbf{z} \leq 0$ 

#### Robustness, 4

Using s-procedure we enforce (7) when (6) holds:

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{w} \end{bmatrix}^{\top} \begin{bmatrix} \mathbf{A}^{\top} \mathbf{S} + \mathbf{S} \mathbf{A} + \gamma \mathbf{E}^{\top} \mathbf{E} & \mathbf{S} \mathbf{F} \\ \mathbf{F}^{\top} \mathbf{S} & -\gamma \mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{w} \end{bmatrix} \le 0$$
 (8)

In LMI form this is:

$$\begin{bmatrix} \mathbf{A}^{\top} \mathbf{S} + \mathbf{S} \mathbf{A} + \gamma \mathbf{E}^{\top} \mathbf{E} & \mathbf{S} \mathbf{F} \\ \mathbf{F}^{\top} \mathbf{S} & -\gamma \mathbf{I} \end{bmatrix} \prec 0$$
 (9)

This is a condition that the system is stable for all values of  $\Delta$ . The decision variables are **S** and  $\gamma$ .

Consider a system:

$$\dot{\mathbf{x}} = (\mathbf{A} + \mathbf{F}\Delta\mathbf{E})\mathbf{x} + \mathbf{B}\mathbf{u} \tag{10}$$

Let us design control for this system that guarantees stability.

We assume control to be linear  $\mathbf{u} = \mathbf{K}\mathbf{x}$ . The system takes the form:

$$\dot{\mathbf{x}} = (\mathbf{A} + \mathbf{B}\mathbf{K} + \mathbf{F}\Delta\mathbf{E})\mathbf{x} \tag{11}$$

Lyapunov eq. then becomes:

$$\mathbf{x}^{\top} (\mathbf{A} + \mathbf{B}\mathbf{K} + \mathbf{F}\Delta\mathbf{E})^{\top} \mathbf{S}\mathbf{x} + \mathbf{x}^{\top} \mathbf{S} (\mathbf{A} + \mathbf{B}\mathbf{K} + \mathbf{F}\Delta\mathbf{E})\mathbf{x} \le 0$$
 (12)

We introduce new variable  $\mathbf{w} = \Delta \mathbf{E} \mathbf{x}$ :

$$\mathbf{x}^{\top}(\mathbf{A} + \mathbf{B}\mathbf{K})^{\top}\mathbf{S}\mathbf{x} + \mathbf{x}^{\top}\mathbf{S}(\mathbf{A} + \mathbf{B}\mathbf{K})\mathbf{x} + \mathbf{x}^{\top}\mathbf{S}\mathbf{F}\mathbf{w} + \mathbf{w}^{\top}\mathbf{F}^{\top}\mathbf{S}\mathbf{x} \leq 0$$

We re-write it as a matrix:

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{w} \end{bmatrix}^{\top} \begin{bmatrix} (\mathbf{A} + \mathbf{B} \mathbf{K})^{\top} \mathbf{S} + \mathbf{S} (\mathbf{A} + \mathbf{B} \mathbf{K}) & \mathbf{S} \mathbf{F} \\ \mathbf{F}^{\top} \mathbf{S} & 0 \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{w} \end{bmatrix} \le 0$$
 (13)

This equation needs to hold only for  $\mathbf{w} = \Delta \mathbf{E} \mathbf{x}$ .

Notice that  $\mathbf{w}^{\top}\mathbf{w} \leq \mathbf{x}^{\top}\mathbf{E}^{\top}\mathbf{E}\mathbf{x}$ , since  $||\Delta|| \leq 1$ . In matrix form it is:

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{w} \end{bmatrix}^{\top} \begin{bmatrix} \mathbf{E}^{\top} \mathbf{E} & 0 \\ 0 & -\mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{w} \end{bmatrix} \ge 0$$
 (14)

Using S-procedure we get:

$$\begin{bmatrix} (\mathbf{A} + \mathbf{B}\mathbf{K})^{\mathsf{T}} \mathbf{S} + \mathbf{S}(\mathbf{A} + \mathbf{B}\mathbf{K}) + \gamma \mathbf{E}^{\mathsf{T}} \mathbf{E} & \mathbf{S} \mathbf{F} \\ \mathbf{F}^{\mathsf{T}} \mathbf{S} & -\gamma \mathbf{I} \end{bmatrix} \prec 0$$
 (15)

where  $\gamma > 0$ . Dividing the resulting LMI by  $\gamma$  and defining  $\mathbf{S}_{\gamma} = \frac{1}{\gamma} \mathbf{S}$  we get:

$$\begin{bmatrix} (\mathbf{A} + \mathbf{B}\mathbf{K})^{\top} \mathbf{S}_{\gamma} + \mathbf{S}_{\gamma} (\mathbf{A} + \mathbf{B}\mathbf{K}) + \mathbf{E}^{\top} \mathbf{E} & \mathbf{S}_{\gamma} \mathbf{F} \\ \mathbf{F}^{\top} \mathbf{S}_{\gamma} & -\mathbf{I} \end{bmatrix} \prec 0 \qquad (16)$$

The expression:

$$\begin{bmatrix} (\mathbf{A} + \mathbf{B}\mathbf{K})^{\top} \mathbf{S}_{\gamma} + \mathbf{S}_{\gamma} (\mathbf{A} + \mathbf{B}\mathbf{K}) + \mathbf{E}^{\top} \mathbf{E} & \mathbf{S}_{\gamma} \mathbf{F} \\ \mathbf{F}^{\top} \mathbf{S}_{\gamma} & -\mathbf{I} \end{bmatrix} \prec 0$$

can be re-written as:

$$\begin{bmatrix} (\mathbf{A} + \mathbf{B}\mathbf{K})^{\top} \mathbf{S}_{\gamma} + \mathbf{S}_{\gamma} (\mathbf{A} + \mathbf{B}\mathbf{K}) & \mathbf{S}_{\gamma} \mathbf{F} \\ \mathbf{F}^{\top} \mathbf{S}_{\gamma} & -\mathbf{I} \end{bmatrix} - \begin{bmatrix} \mathbf{E}^{\top} \\ \mathbf{0} \end{bmatrix} (-\mathbf{I})^{-1} \begin{bmatrix} \mathbf{E} & \mathbf{0} \end{bmatrix} \prec \mathbf{0}$$

then using Schur complement we get:

$$\begin{bmatrix} (\mathbf{A} + \mathbf{B}\mathbf{K})^{\top} \mathbf{S}_{\gamma} + \mathbf{S}_{\gamma} (\mathbf{A} + \mathbf{B}\mathbf{K}) & \mathbf{S}_{\gamma} \mathbf{F} & \mathbf{E}^{\top} \\ \mathbf{F}^{\top} \mathbf{S}_{\gamma} & -\mathbf{I} & 0 \\ \mathbf{E} & 0 & -\mathbf{I} \end{bmatrix} \prec 0$$
 (17)

We can introduce a new variable  $\mathbf{P} = \mathbf{S}_{\gamma}^{-1}$  and multiply both sides (congruence transformation) of the previously found

inequality by 
$$\begin{bmatrix} \mathbf{P} & 0 & 0 \\ 0 & \mathbf{I} & 0 \\ 0 & 0 & \mathbf{I} \end{bmatrix}, \text{ to get:}$$

$$\begin{bmatrix} \mathbf{P}(\mathbf{A} + \mathbf{B}\mathbf{K})^{\top} + (\mathbf{A} + \mathbf{B}\mathbf{K})\mathbf{P} & \mathbf{F} & \mathbf{P}\mathbf{E}^{\top} \\ \mathbf{F}^{\top} & -\mathbf{I} & 0 \\ \mathbf{E}\mathbf{P} & 0 & -\mathbf{I} \end{bmatrix} \prec 0 \tag{18}$$

Finally, introducing the last new variable  $\mathbf{L} = \mathbf{KP}$  we find an LMI that we can solve:

$$\begin{bmatrix} \mathbf{P}\mathbf{A}^{\top} + \mathbf{A}\mathbf{P} + \mathbf{B}\mathbf{L} + \mathbf{L}^{\top}\mathbf{B}^{\top} & \mathbf{F} & \mathbf{P}\mathbf{E}^{\top} \\ \mathbf{F}^{\top} & -\mathbf{I} & 0 \\ \mathbf{E}\mathbf{P} & 0 & -\mathbf{I} \end{bmatrix} \prec 0 \qquad (19)$$

## Quadratic stability, 1

Let us consider the following system:

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} \tag{20}$$

where  $\mathbf{A} = \sum_{i=1}^{n} \alpha_i \mathbf{A}_i$ ,  $\alpha_i \geq 0$ ,  $\sum_{i=1}^{n} \alpha_i = 1$  with known  $\mathbf{A}_i$  but unknown coefficients  $\alpha_i$ . Is it stable for all possible values of  $\alpha_i$ ? Note that we can't use eigenvalue analysis in this case.

Geometrically, this means **A** is in a polytope with vertices  $\mathbf{A}_i$ .

## QUADRATIC STABILITY, 2

#### Theorem (Quadratic stability)

$$\mathbf{A}_i^{\top} \mathbf{S} + \mathbf{S} \mathbf{A}_i \leq 0 \text{ implies } \dot{\mathbf{x}} = \sum_{i=1}^n \alpha_i \mathbf{A}_i \mathbf{x} \text{ is stable, where } \alpha_i \geq 0,$$
  
$$\sum_{i=1}^n \alpha_i = 1$$

Proof: 
$$\dot{V} = \left(\sum_{i=1}^{n} \alpha_i \mathbf{A}_i\right)^{\top} \mathbf{S} + \mathbf{S} \left(\sum_{i=1}^{n} \alpha_i \mathbf{A}_i\right) \leq 0$$
 can be re-written as:  $\dot{V} = \sum_{i=1}^{n} \left(\alpha_i (\mathbf{A}_i^{\top} \mathbf{S} + \mathbf{S} \mathbf{A}_i)\right)$  and since  $\mathbf{A}_i^{\top} \mathbf{S} + \mathbf{S} \mathbf{A}_i < 0$  and  $\alpha_i \geq 0$ , then  $\dot{V} < 0$ .

## Quadratic stability - Control design, 1

Let us consider the following system:

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{x} \tag{21}$$

where  $\mathbf{A} = \sum_{i=1}^{n} \alpha_i \mathbf{A}_i$ ,  $\alpha_i \geq 0$ ,  $\sum_{i=1}^{n} \alpha_i = 1$  with known  $\mathbf{A}_i$  but unknown coefficients  $\alpha_i$ . How to design control law  $\mathbf{u} = \mathbf{K}\mathbf{x}$  making the system stable for all possible values of  $\alpha_i$ ?

The closed-loop form of the system is:

$$\dot{\mathbf{x}} = (\sum_{i=1}^{n} \alpha_i \mathbf{A}_i + \mathbf{B} \mathbf{K}) \mathbf{x}$$
 (22)

## QUADRATIC STABILITY - CONTROL DESIGN, 2

Let us write Lyapunov eq. for the system:

$$\left(\sum_{i=1}^{n} \alpha_i (\mathbf{A}_i + \mathbf{B}\mathbf{K})\right)^{\top} \mathbf{S} + \mathbf{S} \left(\sum_{i=1}^{n} \alpha_i (\mathbf{A}_i + \mathbf{B}\mathbf{K})\right) \prec 0 \quad (23)$$

We can re-write it as:

$$\sum_{i=1}^{n} \alpha_i \left( (\mathbf{A}_i + \mathbf{B} \mathbf{K})^{\top} \mathbf{S} + \mathbf{S} (\mathbf{A}_i + \mathbf{B} \mathbf{K}) \right) < 0$$
 (24)

Hence if  $(\mathbf{A}_i + \mathbf{B}\mathbf{K})^{\top}\mathbf{S} + \mathbf{S}(\mathbf{A}_i + \mathbf{B}\mathbf{K}) \prec 0$ , the original system is stable.

## Quadratic stability - Control design, 3

From  $(\mathbf{A}_i + \mathbf{B}\mathbf{K})^{\top}\mathbf{S} + \mathbf{S}(\mathbf{A}_i + \mathbf{B}\mathbf{K}) \prec 0$ , we can go on to do control design. Introducing  $\mathbf{P} = \mathbf{S}^{-1}$ , we use congruence transformation multiplying by  $\mathbf{P}$  on both sides:

$$\mathbf{P}(\mathbf{A}_i + \mathbf{B}\mathbf{K})^{\top} + (\mathbf{A}_i + \mathbf{B}\mathbf{K})\mathbf{P} < 0$$
 (25)

Introducing new variable  $\mathbf{L} = \mathbf{KP}$  we get a problem linear in decision variables:

$$\mathbf{P}\mathbf{A}_{i}^{\top} + \mathbf{A}_{i}\mathbf{P} + \mathbf{L}^{\top}\mathbf{B}^{\top} + \mathbf{B}\mathbf{L} \prec 0 \tag{26}$$

where the decision variables are  $\mathbf{P}$  and  $\mathbf{L}$ . The control gain matrix is found as  $\mathbf{K} = \mathbf{L}\mathbf{P}^{-1}$ .

#### APPENDIX A

#### Congruence transformation and definiteness

Consider matrices  $\mathbf{P} \succ 0$ , and  $\mathbf{V} \in \mathbb{R}^{n,n}$  is full rank. We can prove that:

$$\mathbf{P} \succ 0 \implies \mathbf{V}^{\top} \mathbf{P} \mathbf{V} \succ 0 \tag{27}$$

Proof:  $\mathbf{x}^{\top} \mathbf{V}^{\top} \mathbf{P} \mathbf{V} \mathbf{x} = \mathbf{z}^{\top} \mathbf{P} \mathbf{z}$ , where  $\mathbf{z} = \mathbf{V} \mathbf{x}$ . Since  $\mathbf{P} \succ 0$ ,  $\mathbf{z}^{\top} \mathbf{P} \mathbf{z} \ge 0$ , hence  $\mathbf{x}^{\top} \mathbf{V}^{\top} \mathbf{P} \mathbf{V} \mathbf{x} \ge 0$ .

#### Definition

Congruence transformation preserves semi-definiteness:  $det(\mathbf{V}) \neq 0, \ \mathbf{P} \succ 0 \implies \mathbf{V}^{\top} \mathbf{P} \mathbf{V} \succ 0$ 

#### Lecture slides are available via Github, links are on Moodle

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