

# **ECE 68000: MODERN AUTOMATIC CONTROL**

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LMI solvers---mincx and gevp

# Minimizer of a Linear Objective Subject to LMI Constraints

- Invoked using the function `mincx`



$$\begin{array}{ll}\text{minimize} & \mathbf{c}^\top \mathbf{x} \\ \text{subject to} & \mathbf{A}(\mathbf{x}) \prec \mathbf{B}(\mathbf{x}).\end{array}$$

- $\mathbf{A}(\mathbf{x}) \prec \mathbf{B}(\mathbf{x})$  is a shorthand notation for general structured LMI systems

# Example

$$\begin{array}{ll}\text{minimize} & \mathbf{c}^\top \mathbf{x} \\ \text{subject to} & \mathbf{Ax} \leq \mathbf{b},\end{array}$$

where

$$\mathbf{c}^\top = \begin{bmatrix} 4 & 5 \end{bmatrix},$$
$$\mathbf{A} = \begin{bmatrix} 1 & 1 \\ 1 & 3 \\ 2 & 1 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 8 \\ 18 \\ 14 \end{bmatrix}.$$

- We first solve the feasibility problem, that is, we find an  $\mathbf{x}$  such that  $\mathbf{Ax} \leq \mathbf{b}$ , using the `fasp` solver
- Then, we solve the above minimization problem using the `mincx` solver

## Example— feasibility problem code

```
% Enter problem data
A =[1 1;1 3;2 1];b=[8 18 14]';c=[-4 -5]';
setlmis([]);
X=lmivar(2,[2 1]);
lmitem([1 1 1 X],A(1,:),1);
lmitem([1 1 1 0],-b(1));
lmitem([1 2 2 X],A(2,:),1);
lmitem([1 2 2 0],-b(2));
lmitem([1 3 3 X],A(3,:),1);
lmitem([1 3 3 0],-b(3));
lmis=getlmis;
disp('---feasp result ---')
[tmin,xfeas]=feasp(lmis);
x_feasp=dec2mat(lmis,xfeas,X)
```

## Example—minimization problem code

```
% Enter problem data
A =[1 1;1 3;2 1];b=[8 18 14]';c=[-4 -5]';
setlmis([]);
X=lmivar(2,[2 1]);
lmitem([1 1 1 X],A(1,:),1);
lmitem([1 1 1 0],-b(1));
lmitem([1 2 2 X],A(2,:),1);
lmitem([1 2 2 0],-b(2));
lmitem([1 3 3 X],A(3,:),1);
lmitem([1 3 3 0],-b(3));
lmis=getlmis;
disp('---mincx result---')
[obj,x_min]=mincx(lmis,c,[0.0001 1000 0 0 1])
```

# Example—output

- The feasp produces

$$\mathbf{x}_{feasp} = \begin{bmatrix} -64.3996 \\ -25.1712 \end{bmatrix}.$$

- The mincx produces

$$\mathbf{x}_{mincx} = \begin{bmatrix} 3.0000 \\ 5.0000 \end{bmatrix}.$$

# Function `defcx`

- The function `defcx` is used to construct the vector  $\mathbf{c}$  used by the LMI solver `mincx`
- Suppose that we wish to solve the optimization problem

$$\begin{array}{ll}\text{minimize} & \text{trace}(\mathbf{P}) \\ \text{subject to} & \mathbf{A}^\top \mathbf{P} + \mathbf{P} \mathbf{A} \prec 0,\end{array}$$

where  $\mathbf{P} = \mathbf{P}^\top \succ 0$ .

- Can use the function `mincx` to solve the above problem
- However, to use the function `mincx`, we need a vector  $\mathbf{c}$  such that

$$\mathbf{c}^\top \mathbf{x} = \text{trace}(\mathbf{P}).$$

## Example: using function defcx

- After specifying the LMIs and obtaining their internal representation using, for example, the command `lmisys=getlmis`, we can obtain the desired  $\mathbf{c}$  with MATLAB's code,

```
q=decnbr(lmisys);  
c=zeros(q,1);  
for j=1:q  
    Pj=defcx(lmisys,j,P);  
    c(j)=trace(Pj);  
end
```

- Having obtained the vector  $\mathbf{c}$ , we can use the function `mincx` to solve the optimization problem



# Generalized Eigenvalue Minimization Problem

- The problem:

$$\begin{array}{ll}\text{minimize} & \lambda \\ \text{subject to} & \\ & \mathbf{C}(\mathbf{x}) \prec \mathbf{D}(\mathbf{x}) \\ & \mathbf{0} \prec \mathbf{B}(\mathbf{x}) \\ & \mathbf{A}(\mathbf{x}) \prec \lambda \mathbf{B}(\mathbf{x}).\end{array}$$

- Need to distinguish between standard LMI constraints of the form  $\mathbf{C}(\mathbf{x}) \prec \mathbf{D}(\mathbf{x})$  and the linear-fractional LMIs of the form  $\mathbf{A}(\mathbf{x}) \prec \lambda \mathbf{B}(\mathbf{x})$  that are concerned with the generalized eigenvalue  $\lambda$ , that is, the LMIs involving  $\lambda$
- The number of linear-fractional constraints is specified with `nflc`
- The generalized eigenvalue minimization problem under LMI constraints is solved by calling the solver `gevp`

# Basic structure of the gevp solver

- The number of linear-fractional constraints is specified with `nflc`
- The generalized eigenvalue minimization problem under LMI constraints is solved by calling the solver `gevp`
- The basic structure of the `gevp` solver:  
`[lopt,xopt]=gevp{lmysys,nflc}`
- It returns `lopt`, which is the global minimum of the generalized eigenvalue, and `xopt`, which is the optimal decision vector variable
- The argument `lmysys` is the system of LMIs,  $\mathbf{C}(\mathbf{x}) \prec \mathbf{D}(\mathbf{x})$ ,  $0 \prec \mathbf{B}(\mathbf{x})$ , and  $\mathbf{A}(\mathbf{x}) \prec \lambda \mathbf{B}(\mathbf{x})$  for  $\lambda = 1$ .
- The corresponding optimal values of the matrix variables are obtained using `dec2mat`

# Application of the gevp solver

- Consider a system model

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x}, \quad \mathbf{x}(0) = \mathbf{x}_0,$$

where the matrix  $\mathbf{A} \in \mathbb{R}^{n \times n}$  is asymptotically stable

- We wish to estimate the decay rate of the system's trajectory  $\mathbf{x}(t)$ , that is, we wish to find constants  $\eta > 0$  and  $M > 0$  such that

$$\|\mathbf{x}(t)\| \leq e^{-\eta t} M(\mathbf{x}_0)$$

- Because  $\mathbf{A}$  is asymptotically stable, by Lyapunov's theorem, for any  $\mathbf{Q} = \mathbf{Q}^\top \succ 0$  there exists  $\mathbf{P} = \mathbf{P}^\top \succ 0$  such that

$$\mathbf{A}^\top \mathbf{P} + \mathbf{P} \mathbf{A} = -\mathbf{Q},$$

that is,

$$\mathbf{x}^\top (\mathbf{A}^\top \mathbf{P} + \mathbf{P} \mathbf{A}) \mathbf{x} = -\mathbf{x}^\top \mathbf{Q} \mathbf{x}$$

# Application of the gevp solver—contd.

- Let

$$V = \mathbf{x}^\top \mathbf{P} \mathbf{x}$$

- Then

$$\dot{V} = \mathbf{x}^\top (\mathbf{A}^\top \mathbf{P} + \mathbf{P} \mathbf{A}) \mathbf{x},$$

which is the Lyapunov derivative of  $V$  evaluated on trajectories of  $\dot{\mathbf{x}} = \mathbf{A} \mathbf{x}$

- Let

$$\alpha = \min \left( -\frac{\dot{V}}{V} \right).$$

# Decay rate

- Let

$$\alpha = \min \left( -\frac{\dot{V}}{V} \right).$$

- We have

$$\dot{V} \leq -\alpha V$$

- Therefore,

$$V(t) \leq e^{-\alpha t} V(0).$$

- We refer to  $\alpha$  as the decay rate of  $V$
- We have

$$\mathbf{x}(t)^\top \mathbf{P} \mathbf{x}(t) \leq e^{-\alpha t} \mathbf{x}_0^\top \mathbf{P} \mathbf{x}_0$$

- Hence

$$\lambda_{\min}(\mathbf{P}) \|\mathbf{x}\|_2^2 \leq \mathbf{x}^\top \mathbf{P} \mathbf{x}$$

## Decay rate—contd.

- Because  $\mathbf{P} = \mathbf{P}^\top \succ 0$ , we have that  $\lambda_{\min}(\mathbf{P}) > 0$
- Combining and dividing both sides by  $\lambda_{\min}(\mathbf{P}) > 0$  gives

$$\|\mathbf{x}(t)\|^2 \leq e^{-\alpha t} \frac{\mathbf{x}_0^\top \mathbf{P} \mathbf{x}_0}{\lambda_{\min}(\mathbf{P})}$$

- Hence,

$$\|\mathbf{x}(t)\| \leq e^{-\frac{\alpha}{2}t} \sqrt{\frac{\mathbf{x}_0^\top \mathbf{P} \mathbf{x}_0}{\lambda_{\min}(\mathbf{P})}}.$$

- Represent  $\mathbf{P} = \mathbf{P}^\top$  as

$$\mathbf{P} = \mathbf{P}^{1/2} \mathbf{P}^{1/2}.$$

# Computing decay rate

- Hence,

$$\begin{aligned}\mathbf{x}_0^\top \mathbf{P} \mathbf{x}_0 &= \mathbf{x}_0^\top \mathbf{P}^{1/2} \mathbf{P}^{1/2} \mathbf{x}_0 \\ &= \|\mathbf{P}^{1/2} \mathbf{x}_0\|^2\end{aligned}$$

- Hence

$$\|\mathbf{x}(t)\| \leq e^{-\frac{\alpha}{2}t} \frac{\|\mathbf{P}^{1/2} \mathbf{x}_0\|}{\sqrt{\lambda_{\min}(\mathbf{P})}}$$

and we obtain

$$\eta = \alpha/2 \quad \text{and} \quad M(\mathbf{x}_0) = \frac{\|\mathbf{P}^{1/2} \mathbf{x}_0\|}{\sqrt{\lambda_{\min}(\mathbf{P})}}.$$

- Finding  $\alpha$  is equivalent to minimizing  $\alpha$  subject to

$$\begin{aligned}\mathbf{P} &\succ 0 \\ \mathbf{A}^\top \mathbf{P} + \mathbf{P} \mathbf{A} &\prec -\alpha \mathbf{P}.\end{aligned}$$

## Example: Computing decay rate

```
A =[-1.1853    0.9134    0.2785
     0.9058   -1.3676    0.5469
     0.1270    0.0975   -3.0000];

setlmis([]);
P = lmivar(1,[3 1])
lmitem([ -1 1 1 P],1,1)           % P
lmitem([ 1 1 1 0],.01)           % P > 0.01*I
lmitem([ 2 1 1 P],1,A,'s')
% linear fractional constraint: left-hand side
lmitem([ -2 1 1 P],1,1)
% linear fractional constraint: right-hand side
lmis = getlmis;
[gamma,P_opt]=gevp(lmis,1);
P=dec2mat(lmis,P_opt,P)
alpha=-gamma
```



## Example: decay rate

- Matrix  $A$

$$A = \begin{bmatrix} -1.1853 & 0.9134 & 0.2785 \\ 0.9058 & -1.3676 & 0.5469 \\ 0.1270 & 0.0975 & -3.0000 \end{bmatrix};$$

- Solution

$$\alpha = 0.6561 \quad \text{and} \quad \mathbf{P} = \begin{bmatrix} 0.6996 & -0.7466 & -0.0296 \\ -0.7466 & 0.8537 & -0.2488 \\ -0.0296 & -0.2488 & 3.2307 \end{bmatrix}$$