# **Lecture 24: Time-Varying Systems Theory and Control**

Textbook Section 6.1 & 6.2

Dr. Jordan D. Larson

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  - Requires additional dynamical systems theory
- Introduces concepts used in nonlinear systems as well

• Time-varying dynamical system:

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- Initial value problem (IVP): system at time  $t_0 \ge 0$  as some  $\vec{x}(t_0) = \vec{x}_0 \in \mathbb{R}^n$ 
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  - May have many solutions, one unique solution, or no existing solution
- Existence and uniqueness of solutions to IVPs for non-LTI dynamical systems not always guaranteed

#### **Existence and Uniqueness of Solutions**

 Cauchy-Peano Theorem: sufficient conditions for IVP to admit solution, may not be unique:

If, for some T > 0, some  $\epsilon > 0$ , and  $f(t, \vec{x}) : \mathbb{R} \times \mathbb{R}^n \to \mathbb{R}^n$  continuous in closed region, i.e.

$$B = \{(t, \vec{x}) : |t - t_0| \le T, ||\vec{x} - \vec{x}_0|| \le \epsilon\} \subseteq \mathbb{R} \times \mathbb{R}^n$$
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Then, exists  $t_0 < t_1 < \le T$  s.t. IVP has at least one continuously differentiable solution  $\vec{x}(t)$  on interval  $[t_0, T]$ 

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- Does not guarantee uniqueness of solution
- Key constraint that yields uniqueness: **Lipschitz condition** whereby  $f(t, \vec{x})$  satisfies inequality

$$||f(t, \vec{x}) - f(t, \vec{y})|| \le L||\vec{x} - \vec{y}||$$
 (4)

for all  $(t, \vec{x})$  and  $(t, \vec{y})$  in some "neighborhood" of  $(t_0, \vec{x}_0)$  with finite constant L > 0

### **Existence and Uniqueness of Solutions (continued)**

• Sufficient conditions for IVP to admit *local* existence and uniqueness of solution: If, for some  $\epsilon > 0$ ,  $f(t, \vec{x}) : \mathbb{R} \times \mathbb{R}^n \to \mathbb{R}^n$  piece-wise continuous in t and satisfies Lipschitz condition, i.e.

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then, exists some  $\delta > 0$  such that IVP for state equation  $\dot{\vec{x}} = f(t, \vec{x})$  with  $\vec{x}(t_0) = \vec{x}_0$  has unique solution over  $[t_0, t_0 + \delta]$ .

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• Sufficient conditions for IVP to admit *global* existence and uniqueness of solution: If, for some finite L > 0 and  $f(t, \vec{x}) : \mathbb{R} \times \mathbb{R}^n \to \mathbb{R}^n$  piece-wise continuous in t and globally Lipschitz in  $\vec{x}$ , i.e.

$$||f(t, \vec{x}) - f(t, \vec{y})|| \le L||\vec{x} - \vec{y}||, \ \forall \ \vec{x}, \vec{y} \in \mathbb{R}^n, \ \forall \ t \in [t_0, t_1]$$
 (6)

then, IVP has unique solution over  $[t_0, t_1]$  where final time  $t_1$  may be arbitrarily large

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- Origin  $\mathbb{R}^n$ : **equilibrium point** for unforced, time-varying system at  $t_0 = 0$  if

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• Define: new time as  $\tau = t - t_0$  and new state as  $\vec{z}(\tau) = \vec{x}(\tau + t_0) - \vec{x}$ 

$$\frac{d\vec{z}(\tau)}{d\tau} = \frac{\vec{x}(\tau + t_0)}{dt} = f(\tau + t_0, \vec{z}(\tau) + \bar{\vec{x}}) = g(\tau, \vec{z}(\tau))$$
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- With  $q(0, \vec{0}) = f(t_0, \vec{x}) = 0$
- Can shift equilibrium point to origin and initial time to zero

### Zero and Origin as Arbitrary Equilibrium

• Suppose one has state trajectory  $\vec{x}(t)$  that starts at  $t = t_0$ , i.e.

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= f(\tau + t_0, \vec{z}(\tau) + \vec{x}(\tau + t_0)) - f(\tau + t_0, z(\tau) + \vec{x}(\tau + t_0)) = g(\tau, \vec{z}(\tau))$$
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- With  $g(0, \vec{0}) = \vec{0}$
- Analyzing these new dynamics around origin as equilibrium point while starting at  $t_0$ : determine original system behavior around original nonzero equilibrium  $\vec{x}$ 
  - I.e. can assess system relative dynamics w.r.t. any time-dependent trajectory  $\vec{\vec{x}}(t)$ , starting at arbitrary initial time  $t_0 \geq 0$
  - Key for stability analysis

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### **Lyapunov Stability**

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- System stability can be interpreted as continuity of system trajectories, w.r.t. initial conditions, over *infinite* time interval
- Infinite time interval highlights primary notion of stability as continuity property of Lipschitz-continuous differential equations holding infinitely in time
  - Let  $\vec{x}(t, \vec{x}_0)$  define unique solution of  $\dot{\vec{x}} = f(t, \vec{x})$  with initial condition  $\vec{x}(t_0) = \vec{x}_0$  which exists on finite, possibly open-ended interval  $[t_0, T)$
- Continuity property of  $\vec{x}(t, \vec{x}_0)$  due to changes in  $\vec{x}_0$ :
  - Given any constant  $\epsilon>0$ , there must exist sufficiently small constant  $\delta>0$  such that for all perturbed initial conditions  $\vec{x}_0+\Delta\vec{x}_0$  with  $\|\Delta\vec{x}_0\|\leq\delta$  Corresponding perturbed solution  $\vec{x}(t,\vec{x}_0+\Delta\vec{x}_0)$  deviates from original  $\vec{x}_0$  by no more than  $\epsilon$ 
    - I.e.  $\|\vec{x}(t, \vec{x}_0 + \Delta \vec{x}_0) \vec{x}(t, \vec{x}_0)\| \le \epsilon$ , for all  $t_0 \le t < T$

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- Lyapunov stability of equilibrium point,  $\vec{x} = \vec{0}$ , i.e. origin, for time-varying unforced dynamics:
  - Stable: if for any  $\epsilon > 0$  and  $t \geq 0$  there exists some  $\delta(\epsilon, t_0) > 0$  such that for all initial conditions  $\|\vec{x}_0\| < \delta$  and for all  $t \geq t_0 \geq 0$ , the corresponding system trajectories bounded, i.e.  $\|\vec{x}(t)\| < \epsilon$
  - Otherwise, unstable
  - Given outer "hyper-sphere"  $B_{\epsilon} = \{ \overrightarrow{x} \in \mathbb{R}^n : \| \overrightarrow{x} \| \le \epsilon \}$ , can find inner "hyper-sphere"  $B_{\delta} = \{ \overrightarrow{x} \in \mathbb{R}^n : \| \overrightarrow{x} \| \le \delta \}$ , such that any trajectory that starts inside  $B_{\delta}$  will evolve inside  $B_{\epsilon}$  for *all* future times

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- ullet System trajectories of time-varying dynamical systems depend on initial time  $t_0$ 
  - Stability of equilibrium point for time-varying systems may depend on  $t_0$
  - Equilibrium point has **uniform stability** if stable and  $\delta$  does not depend on  $t_0$

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- Equilibrium point,  $\vec{x} = \vec{0}$ , i.e. origin, has **uniform asymptotic stability**: If uniformly stable and there exists constant c > 0 independent of  $t_0$  such that  $\vec{x} \to 0$  as  $t \to \infty$  for all  $\|\vec{x}\| \le c$  uniformly in  $t_0$

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- Uniform asymptotic stability typically highly desirably property of control system design: able to maintain their closed-loop performance in presence of state perturbations and disturbances

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  - Provide verifiable sufficient conditions for stability of nominal trajectory without explicit knowledge of system solutions

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- **Lyapunov's indirect method**: can determine stability of equilibrium point, i.e. origin, for nonlinear, time-invariant *n*-dimensional systems by linearizing system dynamics about equilibrium point
  - Utilized in previous systems theory discussions
- Lyapunov's direct method requires concepts:
  - Positive and negative definite functions
  - Time derivative of scalar function along state trajectories of differential equation, i.e. possible solutions

# **Positive and Negative Definite Functions**

- $V(\overrightarrow{x}): \mathbb{R}^n \to \mathbb{R}$  of vector argument  $\overrightarrow{x} \in \mathbb{R}^n$ : **locally positive definite** If  $V(\overrightarrow{0}) = 0$  and there exists constant  $\epsilon$  such that V > 0 for all  $\overrightarrow{x} \in \mathbb{R}^n$  in neighborhood of origin, i.e.  $B_{\epsilon} = \{\overrightarrow{x} \in \mathbb{R}^n : ||\overrightarrow{x}|| \le \epsilon\}$ 
  - If  $\epsilon = \infty$ , then  $V(\vec{x})$  globally positive definite

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- $V(\overrightarrow{x}): \mathbb{R}^n \to \mathbb{R}$  of vector argument  $\overrightarrow{x} \in \mathbb{R}^n$ : **locally negative definite** If  $V(\overrightarrow{0}) = 0$  and there exists constant  $\epsilon$  such that V < 0 for all  $\overrightarrow{x} \in \mathbb{R}^n$  in neighborhood of origin, i.e.  $B_{\epsilon} = \{\overrightarrow{x} \in \mathbb{R}^n : ||\overrightarrow{x}|| \le \epsilon\}$ 
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- $\nabla V(\vec{x}) = [\frac{\partial V}{\partial x_1}, ..., \frac{\partial V}{\partial x_n}]$ : row vector gradient of  $V(\vec{x})$  w.r.t.  $\vec{x}$
- Time derivative of  $V(\vec{x})$  depends not only on function  $V(\vec{x})$  but also on system dynamics under consideration
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- Changing latter while keeping same  $V(\vec{x})$  may result in different  $\dot{V}(\vec{x})$
- Lyapunov direct method: let  $\vec{\vec{x}} = \vec{0} \in \mathbb{R}^n$  as equilibrium point for time-varying dynamics, initial conditions drawn from domain  $D \subset \mathbb{R}^n$  with  $\vec{\vec{x}} \in D$  and  $t_0 = 0$

## **Lyapunov Direct Method**

• If on domain D, there exists continuously differentiable locally positive definite function  $V(\vec{x}): D \to \mathbb{R}$ , whose time derivative along system trajectories locally negative semi-definite, i.e.

$$\dot{V}(\vec{x}) = \nabla V(\vec{x}) f(t, \vec{x}) \le 0 \tag{13}$$

for all  $t \ge 0$  and for all  $\vec{x} \in D$ , then  $\vec{x} = \vec{0}$  locally uniformly stable

• If  $\dot{V}(\vec{x}) < 0$  for all  $t \ge 0$ , i.e. time derivative along the system trajectories locally negative definite, then  $\bar{\vec{x}} = 0$  locally uniformly asymptotically stable

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- If  $\dot{V}(\vec{x}) < 0$  for all  $t \ge 0$ , i.e. time derivative along the system trajectories locally negative definite, then  $\vec{x} = \vec{0}$  locally uniformly asymptotically stable
- Any locally positive definite  $V(\vec{x})$ : Lyapunov function candidate, if satisfies time derivative condition: Lyapunov function
  - Existence of Lyapunov function sufficient to claim uniform stability for equilibrium point, if one cannot be found, nothing can be stated about stability of equilibrium point
  - Lyapunov functions not unique

- Lyapunov function viewable as "energy-like" function for testing system stability
  - If values of *V* do not increase along system trajectories, then origin uniformly stable
  - If *V* strictly decreases, then, in addition, system trajectories will approach origin asymptotically
  - Uniform asymptotic stability requires subset of D known as region of attraction, i.e. starting there, system solutions will converge to origin
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- Define  $\Omega_c = \{ \vec{x} \in \mathbb{R}^n : V(\vec{x}) \leq c \}$  as union of interior set of  $V_c$  and  $V_c$  itself

• Consider converging sequence  $\lim_{k\to\infty} \vec{x}_k = \vec{a}$  with all  $\vec{x}$  from  $\Omega_c$ , then limit point  $\vec{a}$  must also be in  $\Omega_c$ 

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- $\Omega_c$  closed, bounded, and belongs to  $\mathbb{R}^n$ : **compact set**
- Krasovskii-LaSalle theorem:

If  $\vec{x} = \vec{0}$ : equilibrium point of  $\dot{\vec{x}} = f(t, \vec{x})$  and  $V(\vec{x}) : \mathbb{R}^n \to \mathbb{R}$ : radially unbounded Lyapunov function

Then  $\vec{x}$  globally uniformly asymptotically stable equilibrium point

• Simple example of radially unbounded Lyapunov function candidates include quadratic form  $V(\vec{x}) = \vec{x}^T P \vec{x}$  where P symmetric positive definite matrix, i.e.  $P = P^T > 0$ 

#### **Linear, Time-Varying Dynamical System**

• Linear, time-varying (LTV) systems:

$$\dot{\vec{x}}(t) = A(t)\vec{x}(t) + B(t)\vec{u}(t) 
\vec{y}(t) = C(t)\vec{x}(t) + D(t)\vec{u}(t)$$
(14)

Similar to LTI, general solution of LTV/LPV system

$$\vec{X}(t) = \Phi(t, t_0) \vec{X}(t_0) + \int_{t_0}^t \Phi(t, \tau) B(\tau) \vec{U}(\tau) d\tau$$
 (15)

•  $\Phi(t, t_0)$ : state transition matrix of LTV/LPV system

$$\Phi(t,t_0) = I + \int_{\tau}^{t} A(\sigma_1) d\sigma_1 + \int_{\tau}^{t} A(\sigma_1) \int_{\tau}^{\sigma_1} A(\sigma_2) d\sigma_2 d\sigma_1 + \cdots$$
 (16)

•  $\Phi(t, t_0)$  converges uniformly and absolutely to solution that exists and unique

• Lyapunov candidate function with P > 0

$$V(\vec{x}) = \vec{x}^T P \vec{x} \tag{17}$$

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Stable LTV systems: globally uniformly asymptotically stable

(17)

(18)

(19)

(20)

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• Linear, time-varying controllability Gramian using state-transition matrix:

$$W_C(t_0, t_1) = \int_{t_0}^{t_1} \Phi(t_0, t) B(t) B(t)^T \Phi(t_0, t)^T dt$$
 (21)

•  $W_C(t_0, t_1)$ : symmetric, positive semi-definite, satisfies

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## Linear, Parameter-Varying Systems

Special class of LTV systems: linear, parameter-varying (LPV) systems:

$$\vec{x}(t) = A(\vec{\beta}(t))\vec{x}(t) + B(\vec{\beta}(t))\vec{u}(t) 
\vec{y}(t) = C(\vec{\beta}(t))\vec{x}(t) + D(\vec{\beta}(t))\vec{u}(t)$$
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- $\vec{\beta}(t)$ : parameter vector must be continuously differentiable function of time
- State-space matrices continuous functions of  $\vec{\beta}$
- Typically arise when linearizing nonlinear system about equilibrium specified by  $\vec{\beta}$
- $\vec{\beta}(t)$  typically also has restricted admissible trajectories for any particular realization
  - Typically at least restricted by some upper and lower bounds, e.g.  $\vec{\beta}_{II}$  and  $\vec{\beta}_{L}$
  - May also be rate restricted by some upper and lower bounds. e.g.  $\vec{\beta}_{II}$  and  $\vec{\beta}_{II}$
  - $\mathcal{B}$ : set of admissible trajectories for  $\vec{\beta}$

$$\mathcal{B} = \left\{ \overrightarrow{\beta}(t) : \mathbb{R}_+ \to \mathbb{R}^{n_\beta} \text{ such that: } \overrightarrow{\beta}(t) \text{ continuously differentiable,} \right.$$

(26)

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## **LPV Control Design Approaches**

- Common LPV control design approach: gain scheduling
  - Compute array of optimal LTI controllers at variety of *different* equilibrium points across ranges of admissible  $\vec{\beta}$ , e.g. obtain *different* optimal LTI feedback gain matrices
  - Blend various LTI controllers to obtain single, integrated control design, e.g. interpolate between different optimal LTI feedback gain matrices if operating away from equilibrium points
  - Works well for locally "smooth" systems i.e. linearized dynamics reasonably accurate near given equilibrium condition
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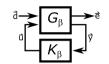
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- More advanced approaches for LPV systems proposed based on specific structure of state-space matrices, e.g. linear or rational dependence on  $\vec{\beta}$
- For arbitrary dependence, specify performance of LPV system,  $G_{\beta}$  in terms of maximum possible induced  $\mathcal{L}_{2\leftarrow 2}$  gain from input  $\vec{u}$  to output  $\vec{v}$ , i.e.

$$\|G_{\beta}\|_{2\leftarrow 2} = \max_{\vec{\beta}\in\mathcal{B}, 0\neq \|\vec{u}\|_{2}\leq \infty, \vec{x}(0)=0} \frac{\|\vec{y}\|_{2}}{\|\vec{u}\|_{2}}$$
(27)

#### State Feedback Linear, Parameter-Varying OCP



• LPV plant:

$$\dot{\vec{x}}(t) = A(\vec{\beta})\vec{x}(t) + \begin{bmatrix} B_{1}(\vec{\beta}) & B_{2}(\vec{\beta}) \end{bmatrix} \begin{bmatrix} d(t) \\ \vec{u}(t) \end{bmatrix} 
\begin{bmatrix} \vec{e}(t) \\ \vec{y}(t) \end{bmatrix} = \begin{bmatrix} C_{1}(\vec{\beta}) \\ I \end{bmatrix} \vec{x}(t) + \begin{bmatrix} 0 & D_{12}(\vec{\beta}) \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \vec{d}(t) \\ \vec{u}(t) \end{bmatrix}$$
(28)

State feedback LPV controller:

$$\vec{u} = D_K(\vec{\beta})\vec{x} \tag{29}$$

Could also use output feedback or observer feedback

#### **Generalized Bounded Real Lemma**

- As  $G_{\beta}$  time-varying, *cannot* interpret as  $\mathcal{H}_{\infty}$ -norm in frequency domain
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  - Generalized bounded real lemma: derive sufficient conditions for upper bound on  $\|G_{\beta}\|_{2\leftarrow 2}$
- Rate-unbounded LPV system globally uniformly asymptotically stable and  $\|G_{\beta}\|_{2\leftarrow 2} \leq \gamma$ : If there exists P>0 such that  $\forall \vec{\beta} \in [\vec{\beta}_I, \vec{\beta}_{IJ}]$ :

$$\begin{bmatrix} A^{T}(\overrightarrow{\beta})P + PA(\overrightarrow{\beta}) & PB(\overrightarrow{\beta}) \\ B^{T}(\overrightarrow{\beta})P & -I \end{bmatrix} + \frac{1}{\gamma^{2}} \begin{bmatrix} C^{T}(\overrightarrow{\beta}) \\ D^{T}(\overrightarrow{\beta}) \end{bmatrix} \begin{bmatrix} C(\overrightarrow{\beta}) & D(\overrightarrow{\beta}) \end{bmatrix} < 0$$
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If there exists P > 0 such that  $\forall \vec{\beta} \in [\vec{\beta}_L, \vec{\beta}_U]$ :

$$\begin{bmatrix} A^{T}(\overrightarrow{\beta})P + PA(\overrightarrow{\beta}) & PB(\overrightarrow{\beta}) \\ B^{T}(\overrightarrow{\beta})P & -I \end{bmatrix} + \frac{1}{\gamma^{2}} \begin{bmatrix} C^{T}(\overrightarrow{\beta}) \\ D^{T}(\overrightarrow{\beta}) \end{bmatrix} \begin{bmatrix} C(\overrightarrow{\beta}) & D(\overrightarrow{\beta}) \end{bmatrix} < 0$$
(30)

• To show  $||G_{\beta}||_{2 \leftarrow 2} \leq \gamma$ : left and right multiply by  $[\vec{x}^T \vec{u}^T]^T$ :

$$\dot{\vec{x}}^T P \vec{x} + \vec{x}^T P \dot{\vec{x}} + \frac{1}{2} \vec{y}^T \vec{y} - \vec{u}^T \vec{u} \le 0$$
(31)

$$\dot{V} + \frac{1}{2^2} \vec{y}^T \vec{y} - \vec{u}^T \vec{u} \le 0 \tag{32}$$

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#### **Generalized Bounded Real Lemma (continued)**

• Integrate from t = 0 to  $t = t_f$ :

$$V(\vec{x}(t_f)) - V(\vec{x}(0)) + \frac{1}{\gamma^2} \int_0^{t_f} \vec{y}^T \vec{y} dt - \int_0^{t_f} \vec{u}^T \vec{u} dt \le 0$$
 (33)

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•  $V(\vec{x}(t_f)) > 0$  and  $V(\vec{x}(0)) = 0$ :

$$\int_0^{t_f} \vec{y}^T \vec{y} dt \le \gamma^2 \int_0^{t_f} \vec{u}^T \vec{u} dt$$
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 (34)

• As  $t_f \to \infty$ :

$$\|\vec{\mathbf{V}}\|_2^2 < \gamma^2 \|\vec{\mathbf{U}}\|_2^2$$

$$\|G_{\beta}\|_{2 \leftarrow 2} < \gamma \tag{36}$$

(35)

(33)

- Matrix inequality as parameterized LMI condition: one LMI for each value of  $\vec{\beta} \in \mathcal{B}$ 
  - In practice: infinite LMI conditions approximated by enforcing only on finite grid of points
  - Finite dimensional LMI conditions directly obtained to bound  $\|G_{\beta}\|_{2\leftarrow 2}$  without approximation if state matrices have rational dependence on  $\vec{\beta}$

- Matrix inequality as parameterized LMI condition: one LMI for each value of  $\vec{\beta} \in \mathcal{B}$ 
  - In practice: infinite LMI conditions approximated by enforcing only on finite grid of points
  - Finite dimensional LMI conditions directly obtained to bound  $\|G_{\beta}\|_{2\leftarrow 2}$  without approximation if state matrices have rational dependence on  $\vec{\beta}$
- Rate-bounded LPV system: alter matrix inequality with variation on Lyapunov theory
  - Check  $2^{n_{\beta}}$  LMI conditions evaluated at endpoints defined by hypercube of elements of  $\vec{\beta}_U$  and  $\vec{\beta}_L$
  - Also need to search over infinitely dimensional space of functions,  $P(\beta)$ , must be restricted to finite dimensional subspace
  - In practice: specify collection of scalar basis functions,  $g_i(\vec{\beta})$ , and use linear combination

$$P(\vec{\beta}) = \sum_{i=1}^{N} g_i(\vec{\beta}) P_i \tag{37}$$

• *P<sub>i</sub>*: symmetric matrices that form finite collection of decision variables, often polynomials chosen as basis functions

# **State Feedback LPV Optimal Control**

Closed-loop LPV system:

$$\vec{x}(t) = \left( A(\vec{\beta}) + B_2(\vec{\beta}) D_K(\vec{\beta}) \right) \vec{x}(t) + B_1(\vec{\beta}) \vec{d}(t) 
\vec{e}(t) = \left( C_1(\vec{\beta}) + D_{12}(\vec{\beta}) D_K(\vec{\beta}) \right) \vec{x}(t)$$
(38)

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• By generalized bounded real lemma, closed-loop LPV system internally stable and  $\|G_{\beta}\|_{2\leftarrow 2} \leq \gamma$  if there exists P > 0 such that  $\forall \vec{\beta} \in [\vec{\beta}_L, \vec{\beta}_U]$ :

$$\begin{bmatrix} A_{L}^{T}(\overrightarrow{\beta})P + PA_{L}(\overrightarrow{\beta}) & PB_{L}(\overrightarrow{\beta}) \\ B_{L}^{T}(\overrightarrow{\beta})P & -I \end{bmatrix} + \frac{1}{\gamma^{2}} \begin{bmatrix} C_{L}^{T}(\overrightarrow{\beta}) \\ D_{L}^{T}(\overrightarrow{\beta}) \end{bmatrix} \begin{bmatrix} C_{L}(\overrightarrow{\beta}) & D_{L}(\overrightarrow{\beta}) \end{bmatrix} < 0$$
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(39)

Use change of variables:

$$Q = P^{-1} \tag{40}$$

$$R(\vec{\beta}) = D_K(\vec{\beta})Q \tag{41}$$

### **State Feedback LPV Optimal Control (continued)**

By Schur Complement Lemma:

$$\begin{bmatrix} QA^{T}(\vec{\beta}) + A(\vec{\beta})Q + R^{T}(\vec{\beta})B_{2}^{T}(\vec{\beta}) + B_{2}(\vec{\beta})R(\vec{\beta}) & B_{1}(\vec{\beta}) & (C_{1}(\vec{\beta})Q + D_{12}(\vec{\beta})R(\vec{\beta}))^{T} \\ B_{1}^{T}(\vec{\beta}) & \gamma^{-2}I & 0 \\ C_{1}(\vec{\beta})Q + D_{12}(\vec{\beta})R(\vec{\beta}) & 0 & -I \end{bmatrix} < 0$$
(42)

# **State Feedback LPV Optimal Control (continued)**

By Schur Complement Lemma:

$$\begin{bmatrix} \overrightarrow{QA^T}(\vec{\beta}) + \overrightarrow{A}(\vec{\beta})Q + \overrightarrow{R^T}(\vec{\beta})B_2^T(\vec{\beta}) + B_2(\vec{\beta})R(\vec{\beta}) & B_1(\vec{\beta}) & (C_1(\vec{\beta})Q + D_{12}(\vec{\beta})R(\vec{\beta}))^T \\ B_1^T(\vec{\beta}) & \gamma^{-2}I & 0 \\ C_1(\vec{\beta})Q + D_{12}(\vec{\beta})R(\vec{\beta}) & 0 & -I \end{bmatrix} < 0$$
(42)

State feedback LPV OCP as SDP in (Q, R, γ):

$$(Q, R(\vec{\beta}), \gamma)^{opt} = \underset{Q \in \mathbb{S}^{n_X}, R \in \mathbb{R}^{n_U \times n_X}, \gamma > 0}{\operatorname{argmin}} \gamma$$

$$\underbrace{\text{subject to:}} \begin{bmatrix} QA^T(\vec{\beta}) + A(\vec{\beta})Q + R^T(\vec{\beta})B_2^T(\vec{\beta}) + B_2(\vec{\beta})R(\vec{\beta}) & B_1(\vec{\beta}) & (C_1(\vec{\beta})Q + D_{12}(\vec{\beta})R(\vec{\beta}))^T \\ B_1^T(\vec{\beta}) & -\gamma^{-2}I & 0 \\ C_1(\vec{\beta})Q + D_{12}(\vec{\beta})R(\vec{\beta}) & 0 & -I \end{bmatrix} < 0$$

$$Q > 0$$

$$(43)$$

# **State Feedback LPV Optimal Control (continued)**

By Schur Complement Lemma:

$$\begin{bmatrix} QA^{T}(\vec{\beta}) + A(\vec{\beta})Q + R^{T}(\vec{\beta})B_{2}^{T}(\vec{\beta}) + B_{2}(\vec{\beta})R(\vec{\beta}) & B_{1}(\vec{\beta}) & (C_{1}(\vec{\beta})Q + D_{12}(\vec{\beta})R(\vec{\beta}))^{T} \\ B_{1}^{T}(\vec{\beta}) & \gamma^{-2}I & 0 \\ C_{1}(\vec{\beta})Q + D_{12}(\vec{\beta})R(\vec{\beta}) & 0 & -I \end{bmatrix} < 0$$
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$$Q > 0$$

$$(43)$$

• State feedback LPV optimal controller as  $\vec{u}(t) = D_K(\vec{\beta})\vec{x}(t)$  reconstructed with

$$D_K(\vec{\beta}) = R(\vec{\beta})Q^{-1} \tag{44}$$

• Solved approximately by gridding over  $[\vec{\beta}_L, \vec{\beta}_U]$ 

# Gain-Scheduling

#### **Summary**

- Non-LTI systems requires additional dynamical systems theory
  - Need to show existence and uniqueness: Lipschitz sufficient
  - Need Lyapunov direct stability method for various definitions of stability for equilibrium points

End

### **Summary**

- Non-LTI systems requires additional dynamical systems theory
  - Need to show existence and uniqueness: Lipschitz sufficient
  - Need Lyapunov direct stability method for various definitions of stability for equilibrium points

- Linear, Time-Varying (LTV) systems exhibit excellent stability properties using Lyapunov theory
  - General solution, controllability, and observability analysis similar to LTI systems using state-transition matrix

End