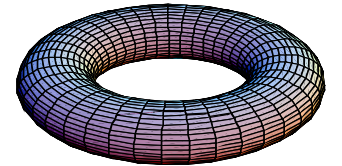


# Robust Control Design via Feedback Linearization

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# Outline

- Perturbations of feedback linearizable systems
- Triangular forms
- Lyapunov Redesign

# Setup

$$\dot{x} = f(x) + g(x)u$$

$$y = h(x)$$

$$x \in R^n, u \in R, y \in R$$

$f(x), g(x), h(x)$  smooth

$$\dot{z} = Az + bv$$

$$y = cz$$

$$A = \begin{bmatrix} 0 & 1 & 0 & & \\ 0 & 0 & 1 & 0 & \\ \vdots & & \ddots & \ddots & \\ & & & \ddots & 1 \\ 0 & & & 0 & 0 \end{bmatrix}, b = \begin{bmatrix} 0 \\ \vdots \\ \vdots \\ 0 \\ 1 \end{bmatrix}, c = [1 \quad 0 \quad \dots \quad 0]$$

# Perturbations of SISO Feedback Linearizable Systems

$$\dot{x} = f(x) + \varphi(x) + [g(x) + \gamma(x)]u$$

$$y = h(x)$$

Assumption: the nominal system is feedback linearizable.

What happens when the nominal system transformation is applied to the actual system?

# Matched Uncertainty, etc.

$$\mathcal{G}_i = \text{span} \{ g, \dots, ad_f^i g \}, 0 \leq i \leq n-1$$

**Definition:** Suppose the system is of relative degree  $r$ .  
We say that the perturbation satisfies:

- The *triangularity condition* if  $ad_\varphi \mathcal{G}_i \in \mathcal{G}_{i+1}, 0 \leq i \leq r-3$
- The *strict triangularity condition* if  $ad_\varphi \mathcal{G}_i \in \mathcal{G}_{i+1}, 0 \leq i \leq r-2$
- The *extended matching condition* if  $\varphi \in \mathcal{G}_1$
- The *matching condition* if  $\varphi \in \mathcal{G}_0$

# Triangular Forms ( ) $r = n$

• *triangularity*  $\Rightarrow$

$$\dot{z}_i = z_{i+1} + \phi_i(z_1, \dots, z_{i+1}), \quad 1 \leq i \leq n-1$$

$$\dot{z}_n = \alpha(x(z)) + \phi_n(z_1, \dots, z_n) + \rho(x(z))u$$

*least restrictive*

• *strict triangularity*  $\Rightarrow$

$$\dot{z}_i = z_{i+1} + \phi_i(z_1, \dots, z_i), \quad 1 \leq i \leq n-1$$

$$\dot{z}_n = \alpha(x(z)) + \phi_n(z_1, \dots, z_n) + \rho(x(z))u$$

• *extended matching*  $\Rightarrow$

$$\dot{z}_i = z_{i+1}, \quad 1 \leq i \leq n-2$$

$$\dot{z}_{n-1} = \phi_{n-1}(z_1, \dots, z_n)$$

$$\dot{z}_n = \alpha(x(z)) + \phi_n(z_1, \dots, z_n) + \rho(x(z))u$$

• *matching*  $\Rightarrow$

$$\dot{z}_i = z_{i+1}, \quad 1 \leq i \leq n-1$$

$$\dot{z}_n = \alpha(x(z)) + \phi_n(z_1, \dots, z_n) + \rho(x(z))u$$

*most restrictive*



# Example

$$\dot{x} = \begin{bmatrix} x_2 \\ x_3 - x_1 - x_2 \\ x_4 \\ x_1 - x_3 - x_4 / 2 \end{bmatrix} + \begin{bmatrix} 0 \\ -\Delta_1(x_1, x_2) \\ 0 \\ -\Delta_2(x_3, x_4) \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} u$$

$f$ 
 $\phi$ 
 $g$

$$\mathcal{G}_0 = \left\{ \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} \right\}, \quad \mathcal{G}_1 = \left\{ \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} \right\}, \quad \mathcal{G}_2 = \left\{ \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} \right\}$$

# Example Cont'd

$$ad_{\phi} \mathcal{G}_0 = \left\{ \begin{bmatrix} 0 \\ 0 \\ 0 \\ \partial\Delta_2/\partial x_4 \end{bmatrix} \right\}, ad_{\phi} \mathcal{G}_1 = \left\{ \begin{bmatrix} 0 \\ 0 \\ 0 \\ \partial\Delta_2/\partial x_3 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ \partial\Delta_2/\partial x_4 \end{bmatrix} \right\},$$

$$ad_{\phi} \mathcal{G}_2 = \left\{ \begin{bmatrix} 0 \\ \partial\Delta_1/\partial x_2 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ \partial\Delta_2/\partial x_3 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ \partial\Delta_2/\partial x_4 \end{bmatrix} \right\}$$

⇓

*strict triangularity*



# Example Cont'd

$$z_1 = x_1$$

$$z_2 = x_2$$

$$z_3 = -x_1 - x_2 + x_3$$

$$z_4 = x_1 - x_3 + x_4$$

$$x_1 = z_1$$

$$x_2 = z_2$$

$$x_3 = z_1 + z_2 + z_3$$

$$x_4 = z_2 + z_3 + z_4$$

$$\Leftrightarrow$$

$$\dot{z} = \begin{bmatrix} z_2 \\ z_3 - \Delta_1(z_1, z_2) \\ z_4 + \Delta_1(z_1, z_2) \\ \frac{1}{2}(-3z_2 - 5z_3 - 3z_4 - 2\Delta_2(z_1 + z_2 + z_3, z_2 + z_3 + z_4)) \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} u$$

# Lyapunov Redesign Process

- Begin with an exactly feedback linearizable nominal system with matched uncertainty
- Apply the nominal control  $u^*(z)$  to actual system and test stability using a Lyapunov function
- Add a new component to nominal control,  $u(z) = u^*(z) + \rho(z)^{-1}\mu(z)$  to enhance stability
- Choose  $\mu(z)$  via Lyapunov design

# Problem Setup

- A linearizable system with matched uncertainty is transformable to:

$$(*) \quad \dot{z} = Az + E[\alpha(z) + \Delta(z, u, t) + \rho(z)u]$$

- Choose a control based on nominal system:

$$u^*(t) = \rho^{-1}(z)(-\alpha(z) + Kz), \quad \text{with } (A + EK) \text{ stable}$$

# Nominal Closed Loop System

$$(**) \quad \dot{z} = (A + EK)z$$

$(A + EK)$  stable  $\Rightarrow$  Lyapunov function

$$V(z) = z^T Pz, \quad \dot{V}(z) = -z^T Qz - \|z\|^2$$

$$P(A + EK) + (A + EK)^T P = -Q - I, \quad Q = Q^T > 0$$

along trajectories of (\*\*)

Lyapunov  
Equation

# Actual Closed Loop System

- Apply  $u^*$  to actual system (\*). Try  $V(z)$

$$\dot{V}(z) = -z^T Q z - \|z\|^2 + 2z^T P E \Delta$$

- Assume

$$\|\Delta\| < \gamma \|z\|, \quad \gamma \geq 0$$

$$\dot{V} \leq -(\lambda_{\min}(Q) + 1)\|z\|^2 + 2\gamma \|PE\| \|z\|^2$$

⇓

*stability* if  $\gamma < (\lambda_{\min}(Q) + 1) / 2\|PE\|$

This implies  
some inherent  
robustness

# Redesign (we can do more) ~ 1

$$u = u^* + \rho^{-1} \mu$$

- Assume the uncertainty satisfies:

$$\Delta(0, 0, t) = 0, \forall t$$

$$\|\Delta(z, u^* + \rho^{-1} \mu, t)\| \leq \sigma(z) \|z\| + k \|\mu\|, 0 \leq k < 1, \sigma \text{ smooth}$$

- The actual closed loop dynamics are:

$$\dot{z} = (A + EK)z + E(\mu + \Delta(z, u^* + \rho^{-1} \mu, t))$$

- The time derivative of  $V$  along trajectories is

$$\dot{V} = -z^T Qz - \|z\|^2 + 2z^T PE(\mu + \Delta)$$

# Redesign ~ 2

- Set  $w^T := z^T P E$  and try to achieve  $-\|z\|^2 + w^T (\mu + \Delta) \leq 0$

- Notice that

$$w^T \mu + w^T \Delta \leq w^T \mu + \|w\| \|\Delta\| \leq w^T \mu + \|w\| [\sigma(z) \|z\| + k \|\mu\|]$$

- For a smooth control, set  $\mu := -w\kappa, \kappa(z) > 0$

$$-\|z\|^2 + w^T \mu + w^T \Delta \leq -\|w\|^2 \kappa (1-k) + \|w\| \sigma(z) \|z\| - \|z\|^2$$

$$\kappa = \frac{1}{4(1-k)} \sigma^2 \Rightarrow$$

$$-\|z\|^2 + w^T \mu + w^T \Delta \leq -\frac{1}{4} \|w\|^2 \sigma^2 + \sigma \|w\| \|z\| - \|z\|^2 = -\left(\frac{1}{2} \|w\| \sigma - \|z\|\right)^2$$

# Redesign ~ 3

- So any  $\kappa > \frac{1}{4(1-k)}\sigma^2$  will do
- In particular

$$\mu = -\left(\sigma_0 + \frac{\sigma^2(z)}{4(1-k)}\right)E^T Pz, \quad \sigma_0 > 0$$



# Redesign ~ 4

- There are other possibilities, recall

$$\dot{V} = -z^T Q z - \|z\|^2 + w^T (\mu + \Delta)$$

- And

$$w^T \mu + w^T \Delta \leq w^T \mu + \|w\| [\sigma(z) \|z\| + k \|\mu\|]$$

- So choose

$$\mu = -\frac{\eta(z)}{1-k} \frac{w}{\|w\|}, \quad \eta(z) > \sigma(z) \|z\|$$

# Example

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} x_2 \\ -0.2x_2 + x_1^3 / 2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} \left[ u + \kappa x_1^3 + au \right]$$

$$\kappa \in [0,1], \quad a \in [-0.1,0.1]$$

$$\alpha = -0.1x_2 + x_1 + x_1^3 / 2, \quad \rho = 1$$

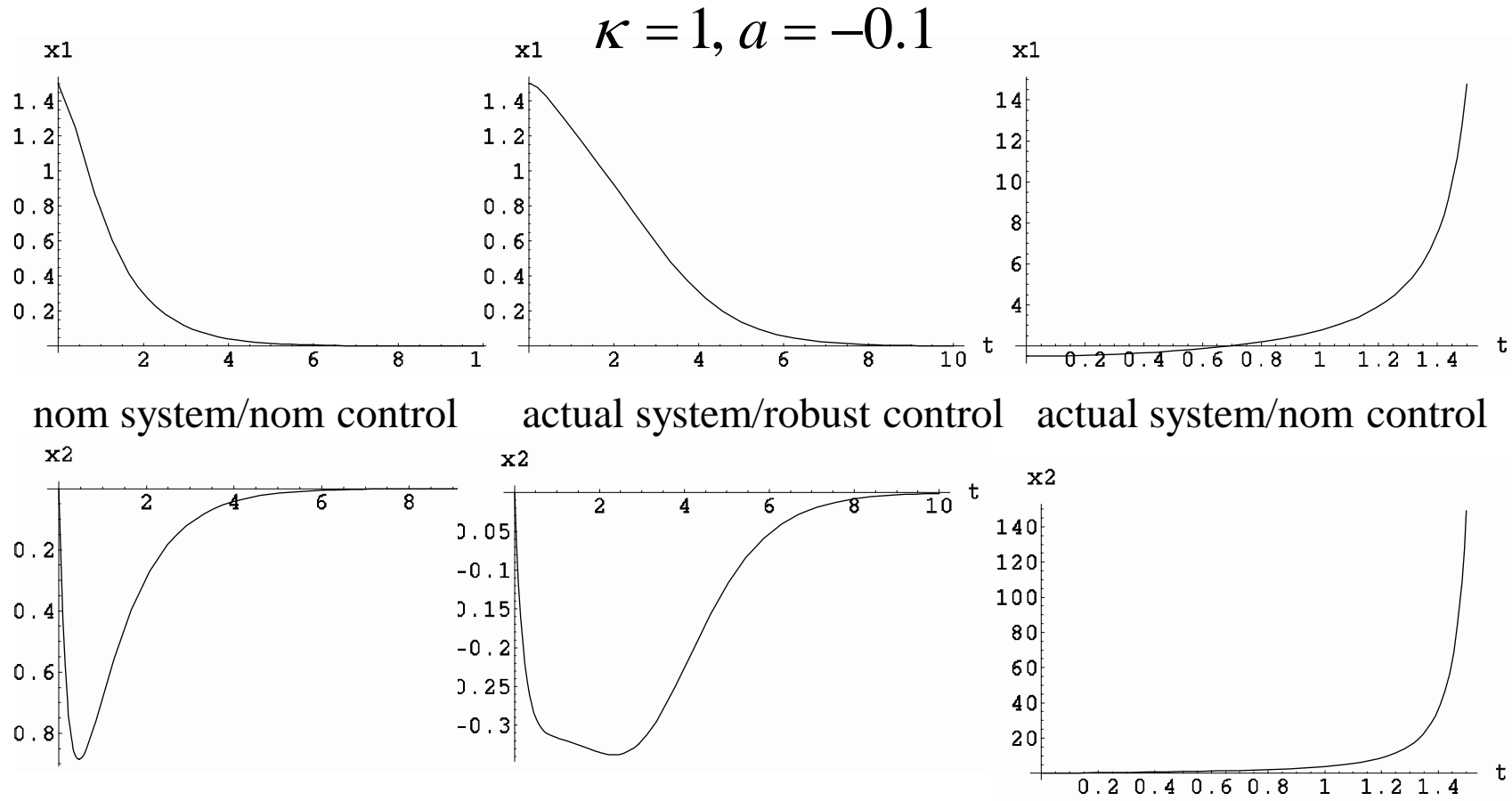
$$u^* = \rho^{-1} \left( -\alpha + \begin{bmatrix} -1 & -2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \right) = -2x_1 - 1.9x_2 - x_1^3 / 2$$

$$\Delta = \kappa x_1^3 + a \left( -2x_1 - \frac{x_1^3}{2} - 1.9x_1 \right) + a\mu$$

$$\sigma^2 = 0.0761 + 1.1025x_1^4, \quad k = a$$

$$u = u^* + \mu = -5.02114x_1 - \frac{x_1^3}{2} - 0.30625x_1^5 - 4.92114x_2 - 0.30625x_1^4x_2$$

# Example ~ Results



# Setup for Backstepping

- SISO system
- $r = n$
- strict triangularity assumption

$$\dot{x}_i = x_{i+1} + \Delta_i(x_1, \dots, x_i, t), \quad 1 \leq i \leq n-1$$

$$\dot{x}_n = \alpha(x) + \rho(x)u + \Delta_n(x, t)$$

$$\det \rho(0) \neq 0$$

$$\Delta(0, t) = 0, \quad |\Delta_i(X_i, t)| \leq \sigma_i(X_i) \|X_i\|, \quad \sigma_i \geq 0$$

$$X_i = \{x_1 \quad \cdots \quad x_i\}$$

# Step 1

$$\begin{aligned} \dot{x}_1 &= v_1 + \Delta_1(x_1, t) \\ y_1 &= x_1 \end{aligned} \quad \Rightarrow \quad \dot{y}_1 = v_1 + \bar{\Delta}_1(y_1, t) := f_1(y_1)$$

$$V_1 := \frac{1}{2} y_1^2$$

$$L_{f_1} V_1 = y_1 (v_1 + \bar{\Delta}_1(y_1, t))$$

$$\text{choose : } v_1 = -k_1 y_1 - \kappa_1(y_1) y_1$$

$\Downarrow$

$$L_{f_1} V_1 = -k_1 y_1^2 - \kappa_1(y_1) y_1^2 + y_1 \bar{\Delta}_1$$

$$\text{take : } \kappa_1(y_1) > \frac{1}{4} \sigma_1^2(x_1), x_1 \neq 0$$

$\Downarrow$

$$L_{f_1} V_1 \leq -(k_1 - 1) y_1^2$$

## Step 2

$$\dot{x}_1 = x_2 + \Delta_1(x_1, t)$$

$$\dot{x}_2 = v_2 + \Delta_2(x_1, x_2, t)$$

$$y_2 = x_2 - v_1(x_1)$$

$$(x_1, x_2) \mapsto (y_1, y_2)$$

⇓

$$\begin{aligned} \dot{y}_1 &= v_1(y_1) + y_2 + \bar{\Delta}_1(y_1, t) \\ \dot{y}_2 &= v_2 + \bar{\Delta}_2(y_1, y_2, t) \end{aligned} \quad := f_1(y_1, y_2)$$

$$\text{define : } V_2 = V_1 + \frac{1}{2} y_2^2$$

$$\text{choose : } v_2 = -y_1 - k_2 y_2 - \kappa_2(y_1, y_2) y_2$$

$$\kappa_2(y_1, y_2) > \frac{1}{4} \bar{\sigma}_2^2(y_1, y_2) \Rightarrow L_{f_2} V_2 \leq -(k_1 - 2) y_1^2 - (k_2 - 1) y_2^2$$