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## **Stabilization**

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# Introduction

Stabilization is one of the major themes in control theory. Very often, a primary goal is to ensure stability (or to improve stability properties), since otherwise the system may just explode.

Let us start with linear systems

$$\dot{x}(t) = Ax(t) + Bu(t), \quad u(t) \in \mathbb{R}^m.$$

Controllability guarantees that one can reach  $0 \in \mathbb{R}^d$  (in finite time) from each  $x_0 \in \mathbb{R}^d$  by an appropriate control  $u_{x_0}(\cdot)$ .

However, if  $A$  has eigenvalues with positive real parts, the solution will diverge under arbitrarily small perturbations:

$$\varphi(t, x_0 + \varepsilon x_1, u_{x_0}) = \varepsilon \underbrace{e^{At} x_1}_{\rightarrow \infty \text{ gener.}} + \underbrace{e^{At} x_0 + \int_0^t e^{A(t-s)} B u_{x_0}(s) ds}_{\rightarrow 0}.$$

# State feedbacks

A remedy is to use feedbacks:

State feedback: Find a matrix  $F$  such that with  $u = Fx$

$$\dot{x}(t) = Ax(t) + BFx(t) = (A + BF)x(t).$$

is (asymptotically) stable.

## Some observations:

(i) By coordinate transformation we may assume that

$$A = \begin{bmatrix} A_1 & A_2 \\ 0 & A_3 \end{bmatrix}, \quad B = \begin{bmatrix} B_1 \\ 0 \end{bmatrix} \quad \text{with } (A_1, B_1) \text{ controllable.}$$

(ii) For scalar control and  $(A, b)$  controllable, we may assume

$$A = \begin{bmatrix} 0 & 1 & \cdot & \cdot & 0 \\ \cdot & \cdot & \cdot & & \cdot \\ \cdot & \cdot & & & 1 \\ \alpha_0 & \alpha_1 & \cdot & \cdot & \alpha_{n-1} \end{bmatrix} \quad \text{and} \quad b = \begin{bmatrix} 0 \\ \cdot \\ 0 \\ 1 \end{bmatrix}$$

with  $\chi_A(z) = z^n - \alpha_{n-1}z^{n-1} - \dots - \alpha_1z - \alpha_0$ .

(iii) This can be stabilized by

$$f = (\beta_0 - \alpha_0, \beta_1 - \alpha_1, \dots, \beta_{n-1} - \alpha_{n-1}) \in \mathbb{R}^{1 \times d},$$

since

$$A + bf = A + \begin{bmatrix} 0 \\ \cdot \\ \cdot \\ 1 \end{bmatrix} (\beta_0 - \alpha_0, \dots, \beta_{n-1} - \alpha_{n-1}) = \begin{bmatrix} 0 & 1 & \cdot & 0 \\ \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & & \cdot \\ \beta_0 & \beta_1 & \cdot & \beta_{n-1} \end{bmatrix}.$$

with  $\chi_A(z) = z^n - \beta_{n-1}z^{n-1} - \dots - \beta_1z - \beta_0$ .

# State feedbacks

(iv) Let  $(A, B)$  be controllable and  $b = Bv \neq 0$ .  
Then there is  $F$  such that

$$(A + BF, b) \text{ is controllable.}$$

(ii) - (iv) imply that every controllable pair is stabilizable.

**Theorem.** For  $(A, B)$  let  $\chi$  be a normed polynomial with  $\deg \chi = \dim \langle A | \text{im} B \rangle$ . Then there exists a feedback  $F$  s.t.

$$\chi_{A+BF} = \chi \cdot \chi_{A_3}.$$

This is known as the **pole shifting theorem**.

The theorem also shows that stabilizability is equivalent to asymptotic null controllability.

# Laplace-transforms and poles

For initial condition  $x(0) = 0$ , take Laplace transforms

$$\hat{u}(s) = \int_0^{\infty} e^{-st} u(t) dt, \quad \hat{x}(s) = \int_0^{\infty} e^{-st} x(t) dt.$$

By partial integration

$$\dot{x}^{\wedge}(s) = \int_0^{\infty} e^{-st} \dot{x}(t) dt = s\hat{x}(s) = s \int_0^{\infty} e^{-st} x(t) dt = s\hat{x}(s).$$

Thus

$$\hat{x}(s) = (sI - A)^{-1} B \hat{u}(s).$$

The eigenvalues of  $A$  are the poles of  $(sI - A)^{-1} B$ .

# Stabilization via outputs

Consider  $\dot{x} = Ax + Bu$ ,  $y = Cx$ .

**Static output feedback:** With  $u = Fy = FCx$

$$\dot{x}(t) = Ax(t) + BFCx = (A + BFC)x(t).$$

## Example

$$\dot{x}_1 = x_2, \dot{x}_2 = u \quad y = x_1.$$

This system is controllable and observable, but there is no (as.) stabilizing feedback  $k : \mathbb{R} \rightarrow \mathbb{R}$

$$\dot{x}_1 = x_2, \dot{x}_2 = k(y) = k(x_1).$$

In fact,

$$V(x_1, x_2) = (x_2)^2 - 2 \int_0^{x_1} k(s) ds$$

is constant along trajectories with  $V(0, 0) = 0$  and  $V(0, \alpha) = \alpha^2 \neq 0$  for  $\alpha \neq 0$ .

Instead of this static output feedback use dynamic output feedback:

Separate the output stabilization problem into two subproblems:

- (i) find a stabilizing state feedback;
- (ii) estimate the state and use this estimate in (i).



# A dynamic observer

ad (ii) For  $\dot{x} = Ax + Bu, y = Cx$  find  $L$  such that  $A + LC$  is stable.

Then, by linearity, the dynamic observer

$$\dot{z} = (A + LC)z - Ly + Bu$$

satisfies

$$\|z(t) - x(t)\| \rightarrow 0 \text{ for } t \rightarrow \infty.$$

In fact: the error  $e(t) = z(t) - x(t)$  converges to 0, since

$$\begin{aligned}\dot{e} &= \dot{z} - \dot{x} = (A + LC)z - Ly + Bu - Ax - Bu \\ &= (A + LC)z - LCx - Ax \\ &= (A + LC)(z - x) \\ &= (A + LC)e.\end{aligned}$$

**Theorem.** If  $(A, B)$  and  $(A^\top, C^\top)$  are stabilizable (i.e., asymptotic null controllability and asymptotic observability hold), then there are  $F$  and  $L$  such that following the dynamic output feedback stabilizes the system,

$$u = Fz,$$

where

$$\dot{z} = (A + LC)z + BFz - LCx$$

# Compensator

Now use the estimate  $z(t)$  instead of the state  $x(t)$  in the state feedback:

Assume that  $(A, B)$  and  $(A^\top, C^\top)$  are stabilizable.

Then the system is stabilized by  $u = Fz$ , since the following coupled system is stable,

$$\begin{aligned}\dot{x} &= Ax + BFz \\ \dot{z} &= (A + LC)z + BFz - LCx.\end{aligned}$$

In fact, it turns out that the system matrix

$$\begin{bmatrix} A & BF \\ -LC & A + LC + BF \end{bmatrix}$$

is stable.

# Linear-quadratic optimal control

This is an efficient (and intensely studied) method to construct stabilizing feedbacks. Consider

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ z(t) &= Cx(t) + Du(t).\end{aligned}$$

Here  $z(t)$  is the output which is to be controlled. This can be done by minimizing for given initial state  $x_0$  over  $u$

$$J(x_0; u) = \int_0^\infty \left[ \|Cx(t)\|^2 + \|Du(t)\|^2 \right] dt.$$

More generally, minimize with  $Q \geq 0$  and  $N > 0$ ,

$$J(x_0; u) = \int_0^\infty \left[ x(t)^\top Qx(t) + u(t)^\top Nu(t) \right] dt.$$

For  $Q > 0$ ,  $x(t) \rightarrow 0$  for  $t \rightarrow \infty$  if there is  $u$  with  $J(x_0; u) < \infty$ .

**Goal:** Show that the optimal controls can be written as feedback  $u = Fx$ .

This problem is closely related to positive semidefinite solutions of the algebraic matrix Riccati equation

$$A^T P + PA - PBB^T P + Q = 0. \quad (\text{ARE})$$

A typical result:

**Theorem.** Assume that  $(A, B)$  is stabilizable and  $\text{spec}(A) \cap i\mathbb{R} = \emptyset$ .

- (i) There is a smallest positive semidefinite solution  $P^-$  of ARE.
- (ii) For every input  $u$

$$J(x_0; u) = x_0^T P^- x_0 + \int_0^\infty \left\| u(t) + B^T P^- x(t) \right\|^2 dt.$$

- (iii) The optimal input is given by the feedback

$$u(t) = -B^T P^- x(t).$$

The **proof** uses the finite time problem and completion of squares.

# An example

Stabilize an inverted pendulum on a flying quadcopter.

The complete system is described by a 16-dimensional system of differential equations (12 for the quadcopter + 4 for the pendulum) with 4 control inputs.

After simplification to 13 dimensions and linearization in the equilibrium a linear-quadratic optimal control problem is solved.

Critical is the measurement of the states which is done by an infrared motion tracking system.

HEHN AND D'ANDREA, IEEE TRANS. AUT. CONTROL (2011)

# Further problems

The  $H^\infty$ -problem for

$$\begin{aligned}\dot{x} &= Ax + Bu + Ed \\ z &= Cx + Du\end{aligned}$$

**Goal:** Given  $\gamma > 0$  find  $F$  such that  $A + BF$  is stable and (for  $x_0 = 0$ )

$$\|z\|_2 \leq \gamma \|d\|_2 \text{ for all perturbations } d \in L^2(0, \infty, \mathbb{R}^\ell).$$

This is possible for  $\gamma > \|G_F\|$  with

$$G_F : L^2(0, \infty) \rightarrow L^2(0, \infty), d(\cdot) \mapsto z(\cdot) = \int_0^\cdot Ce^{(A+BF)(t-\tau)} Ed(\tau) d\tau.$$

(well defined for  $A + BF$  stable)

This again leads to LQ-optimal control (without positive definiteness).

Note that for stable  $A$  and

$$G : L^2(0, \infty) \rightarrow L^2(0, \infty), d(\cdot) \mapsto z(\cdot) = \int_0^\cdot C e^{A(t-\tau)} E d(\tau) \, d\tau$$

and

$$G(s) = C(sI - A)^{-1} E$$

one has

$$\|G\| = \sup \left\{ \frac{\|G(d)\|_2}{\|d\|_2} \mid 0 \neq d \in L^2 \right\} = \sup_{\omega \in \mathbb{R}} \|G(i\omega)\|,$$

where  $\|G(i\omega)\|$  denotes the largest singular value. This is the  $H^\infty$ -norm of matrix-valued functions which are holomorphic on the open right half plane.



# Nonlinear stabilization at an equilibrium

Consider

$$\dot{x}(t) = f(x(t), u(t))$$

and let  $x^*$  be an equilibrium  $f(x^*, u^*) = 0$ . Linearization in  $(x^*, u^*)$  yields

$$\dot{y}(t) = f_x(x^*, u^*)y(t) + f_u(x^*, u^*)v(t)$$

and write  $A = f_x(x^*, u^*)$  and  $B = f_u(x^*, u^*)$ .

Then a stabilizing feedback  $F$  for the linearized system is locally stabilizing for the nonlinear system

$$\dot{x}(t) = f(x(t), F(x(t) - x^*)).$$

(use a Lyapunov function)

# Brockett's necessary condition

**Theorem.** Consider  $\dot{x} = f(x, u)$ ,  $u \in U$  open. If there is a locally stabilizing locally Lipschitz feedback  $F : \mathbb{R}^d \rightarrow U$ , then  $f(\mathbb{R}^d, U)$  is a neighborhood of 0.

**Example** (Brockett's nonholonomic integrator)

$$\dot{x}_1 = u_1$$

$$\dot{x}_2 = u_2$$

$$\dot{x}_3 = x_2 u_1$$

This is a simple model for a vehicle with angle  $\theta = x_1$  in forward direction and position

$$(z_1, z_2) = (x_2 \cos \theta + x_3 \sin \theta, x_2 \sin \theta - x_3 \cos \theta).$$

No point  $(0, r, \varepsilon)$  with  $\varepsilon \neq 0$  and  $r \in \mathbb{R}$  is in the image of  $f$ . The system is asymptotically null controllable.

# Control-Lyapunov functions

Asymptotic controllability to an equilibrium and stabilization can be dealt with using control-Lyapunov functions which decrease along trajectories for appropriate controls..

Roughly,

- asymptotic controllability to an equilibrium holds if there exists a continuous control-Lyapunov function
- stabilizability with continuous feedback holds if there exists a smooth control-Lyapunov function.

cf. Sontag (1998)

# Coron's return method: time-varying feedbacks

**Theorem.** Consider a driftless control system

$$\dot{x} = \sum_{i=1}^m u_i(t) f_i(x)$$

and assume that

$$\{g(x) \mid g \in \mathcal{LA}(f_1, \dots, f_m)\} \text{ for all } x \neq 0.$$

Then for every  $T > 0$  there exists  $u \in C^\infty(\mathbb{R}^d \times \mathbb{R})$  with

$$u(0, t) = 0, \quad u(x, t + T) = u(x, t) \text{ for all } t \in \mathbb{R}, x \in \mathbb{R}^d$$

such that 0 is globally asymptotically stable for

$$\dot{x} = \sum_{i=1}^m u_i(x, t) f_i(x).$$

# Example

Nonholonomic integrator

$$\dot{x}_1 = u_1, \dot{x}_2 = u_2, \dot{x}_3 = x_1 u_2 - x_2 u_1.$$

Here

$$f_1(x) = \begin{bmatrix} 1 \\ 0 \\ -x_2 \end{bmatrix}, f_2(x) = \begin{bmatrix} 0 \\ 1 \\ x_1 \end{bmatrix}.$$

Brockett's necessary condition is violated, but the Lie algebra rank condition is satisfied. Hence it can be globally asymptotically stabilized by means of periodic time-varying feedback.

# Stabilization with piecewise constant controls

Continuous stirred tank reactor

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} -x_1 - a(x_1 - x_c) + B\alpha(1 - x_2)e^{x_1} \\ -x_2 + \alpha(1 - x_2)e^{x_1} \end{bmatrix} + u(t) \begin{bmatrix} x_c - x_1 \\ 0 \end{bmatrix},$$

where  $x_1$  is the coolant temperature and  $x_2$  is the product concentration,  $x_c$  is the coolant temperature and the control affects the heat transfer coefficient with parameters

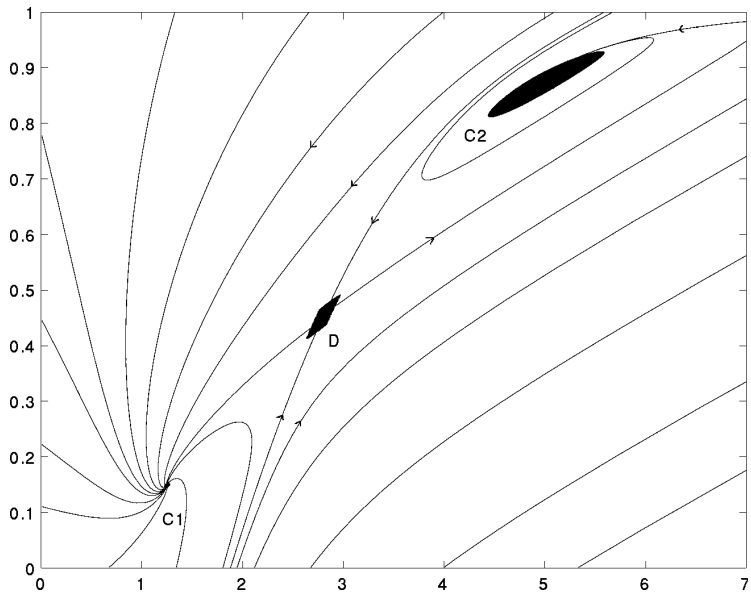
$$a = 0.95, \alpha = 0.05, B = 10.0, c_c = 1.0$$

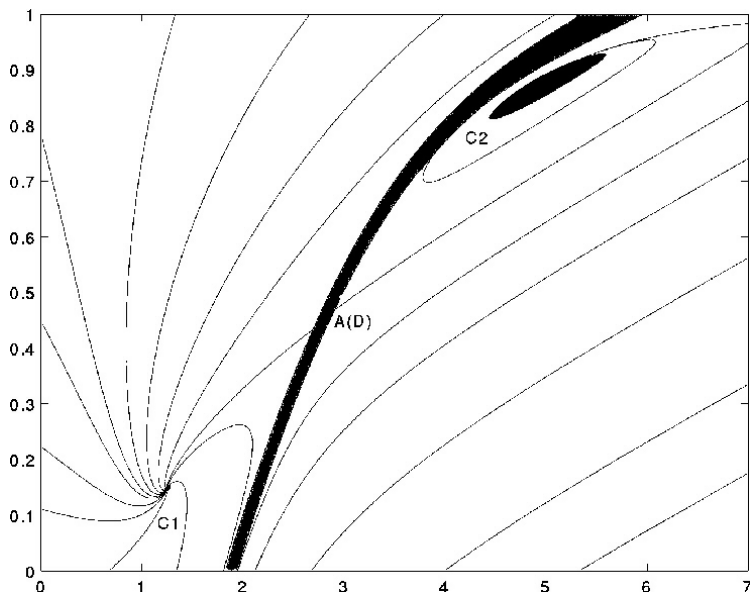
and control range

$$\Omega = [-0.15, 0.15].$$

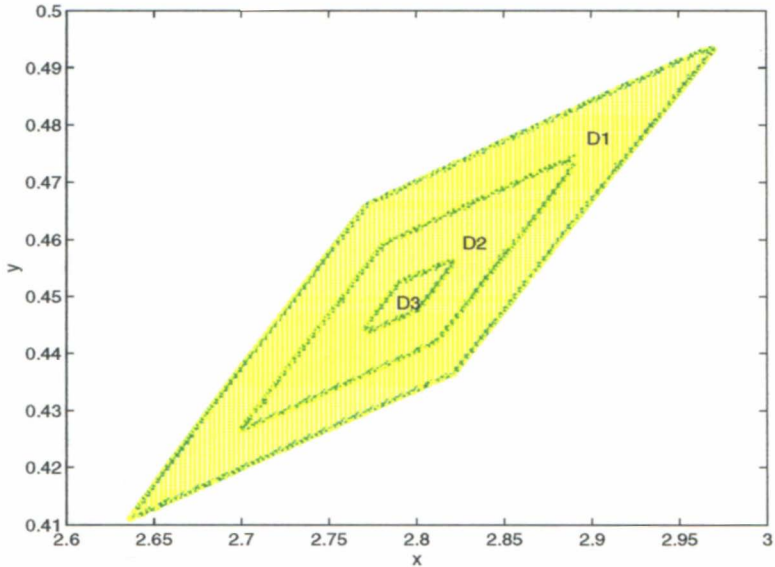
The uncontrolled system has an unstable fixed point at

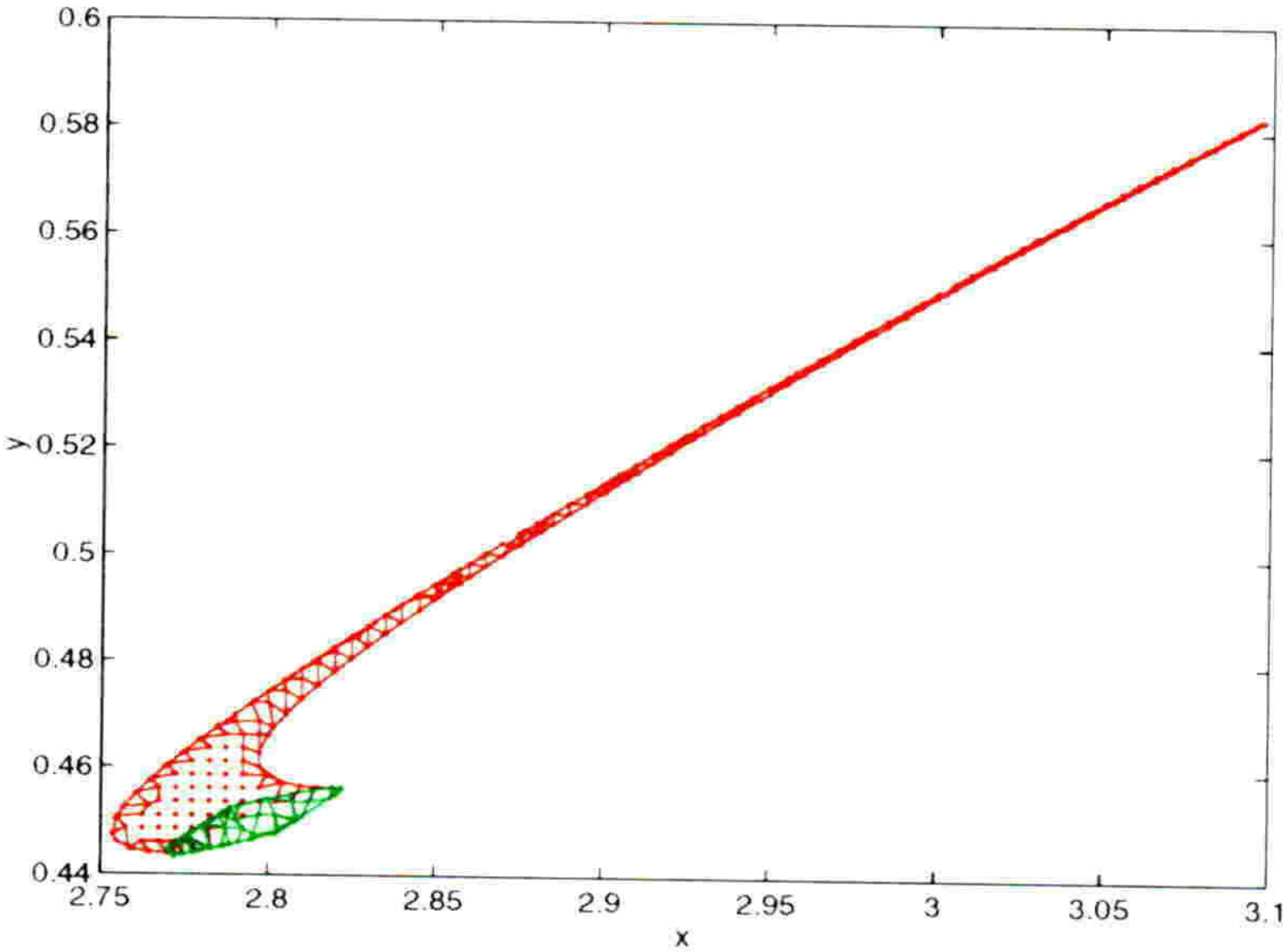
$$(x_1^*, x_2^*) \sim (2.8, 0.45) \in D$$

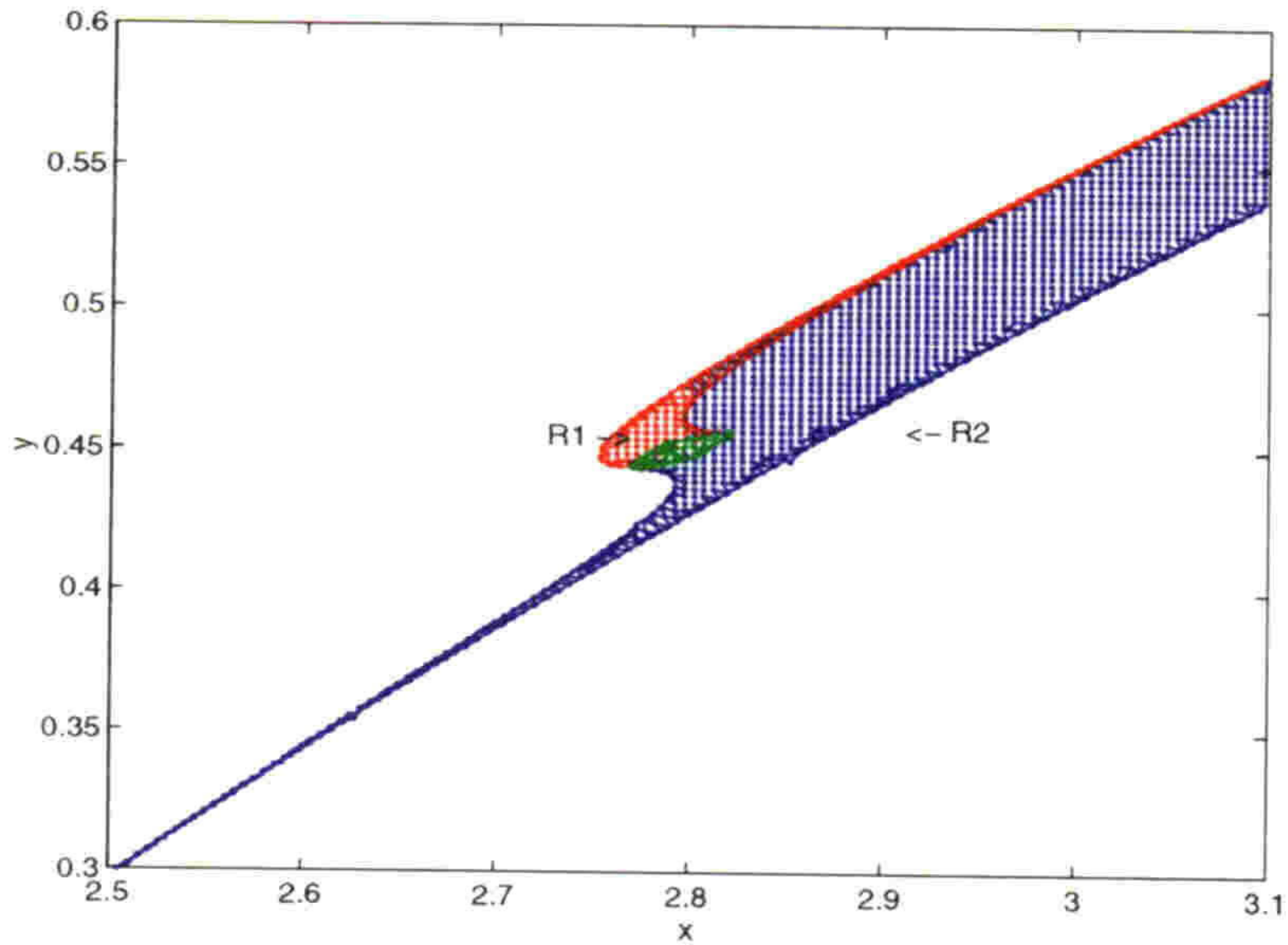




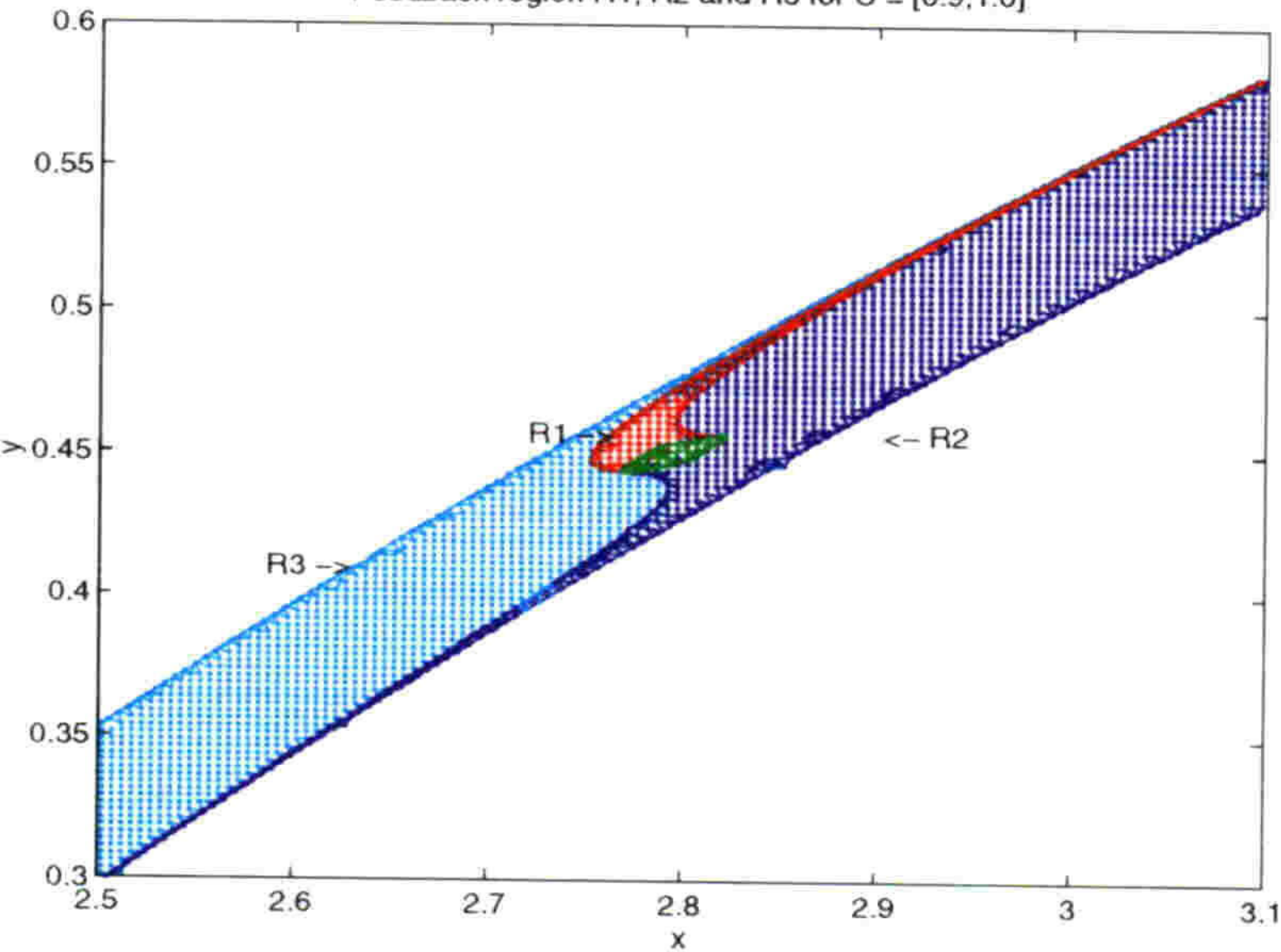








Feedback region R1, R2 and R3 for  $U = [0.9, 1.0]$



# Final remarks

Since asymptotic stabilization is a basic problem in control, there is a multitude of algorithms to achieve it, in addition to the examples presented here.

- Backstepping
- Model-predictive control (receding horizon optimal control)
- ...

Note that in applications stability is only one goal among others including, in particular, robustness properties with respect to perturbations.