

# GLOBAL STABILIZATION FOR SYSTEMS EVOLVING ON MANIFOLDS



**MICHAEL MALISOFF**

Department of Mathematics  
Louisiana State University



Joint with **Mikhail Krichman** and **Eduardo Sontag**

SIAM Conference on Control and Its Applications  
New Orleans, LA, July 11-14, 2005

## STATEMENT of PROBLEM

Given System and Attractor:

$$\dot{x} = f(x, u), \quad x \in \mathcal{X}, \quad u \in \mathbf{U} \quad (\Sigma)$$

Assumptions:  $f$  locally Lipschitz in  $x$  uniformly for  $u$  in compact sets,  
 $\mathcal{X}$  a manifold,  $\mathbf{U}$  a locally compact metric space.

$\mathcal{A} \subseteq \mathcal{X}$ : nonempty, compact, and weakly invariant for  $\Sigma$ .

## STATEMENT of PROBLEM

Given System and Attractor:

$$\dot{x} = f(x, u), \quad x \in \mathcal{X}, \quad u \in \mathbf{U} \quad (\Sigma)$$

Assumptions:  $f$  locally Lipschitz in  $x$  uniformly for  $u$  in compact sets,  $\mathcal{X}$  a manifold,  $\mathbf{U}$  a locally compact metric space.

$\mathcal{A} \subseteq \mathcal{X}$ : nonempty, compact, and weakly invariant for  $\Sigma$ .

**Goal:** Assuming  $\Sigma$  is GAC to  $\mathcal{A}$ , find a state feedback  $u = k(x)$  such that the system  $\dot{x} = f(x, k(x))$  is GAS to  $\mathcal{A}$  in an appropriate sense.

## STATEMENT of PROBLEM

Given System and Attractor:

$$\dot{x} = f(x, u), \quad x \in \mathcal{X}, \quad u \in \mathbf{U} \quad (\Sigma)$$

Assumptions:  $f$  locally Lipschitz in  $x$  uniformly for  $u$  in compact sets,  $\mathcal{X}$  a manifold,  $\mathbf{U}$  a locally compact metric space.

$\mathcal{A} \subseteq \mathcal{X}$ : nonempty, compact, and weakly invariant for  $\Sigma$ .

**Goal:** Assuming  $\Sigma$  is GAC to  $\mathcal{A}$ , find a state feedback  $u = k(x)$  such that the system  $\dot{x} = f(x, k(x))$  is GAS to  $\mathcal{A}$  in an appropriate sense.

**Obstacles:** Virtual (Brockett's Criterion) and topological (Morse Theory)

## STATEMENT of PROBLEM

Given System and Attractor:

$$\dot{x} = f(x, u), \quad x \in \mathcal{X}, \quad u \in \mathbf{U} \quad (\Sigma)$$

Assumptions:  $f$  locally Lipschitz in  $x$  uniformly for  $u$  in compact sets,  $\mathcal{X}$  a manifold,  $\mathbf{U}$  a locally compact metric space.

$\mathcal{A} \subseteq \mathcal{X}$ : nonempty, compact, and weakly invariant for  $\Sigma$ .

**Goal:** Assuming  $\Sigma$  is GAC to  $\mathcal{A}$ , find a state feedback  $u = k(x)$  such that the system  $\dot{x} = f(x, k(x))$  is GAS to  $\mathcal{A}$  in an appropriate sense.

**Obstacles:** Virtual (Brockett's Criterion) and topological (Morse Theory)

**E.g.:** Take  $\mathcal{X} = \mathbb{R}^n$ ,  $\mathcal{A} = \{0\}$ ,  $\dot{x} = g_1(x)u_1 + \dots + g_m(x)u_m$ ,  $m > n$ .

No **continuous** stabilizing  $k(x)$  when  $\text{rank}[g_1(0) \ g_2(0) \ \dots \ g_m(0)] = m$ .

## OBSTACLES to CONTINUOUS STABILIZATION

**Brockett's Criterion (1983):** If  $\dot{x} = f(x, k(x))$  is GAS to 0 on  $\mathbb{R}^n$  where  $k$  is  $C^1$  and satisfies  $k(0) = 0$ , then  $\text{Image}(f)$  contains a ngbd of 0.

## OBSTACLES to CONTINUOUS STABILIZATION

**Brockett's Criterion (1983):** If  $\dot{x} = f(x, k(x))$  is GAS to 0 on  $\mathbb{R}^n$  where  $k$  is  $C^1$  and satisfies  $k(0) = 0$ , then  $\text{Image}(f)$  contains a ngbd of 0.

This provides a **virtual obstacle** to state feedback stabilization.

## OBSTACLES to CONTINUOUS STABILIZATION

**Brockett's Criterion (1983):** If  $\dot{x} = f(x, k(x))$  is GAS to 0 on  $\mathbb{R}^n$  where  $k$  is  $C^1$  and satisfies  $k(0) = 0$ , then  $\text{Image}(f)$  contains a ngbd of 0.

This provides a **virtual obstacle** to state feedback stabilization.

[See EDS/MCT book for proof based on homotopy.]

## OBSTACLES to CONTINUOUS STABILIZATION

**Brockett's Criterion (1983):** If  $\dot{x} = f(x, k(x))$  is GAS to 0 on  $\mathbb{R}^n$  where  $k$  is  $C^1$  and satisfies  $k(0) = 0$ , then  $\text{Image}(f)$  contains a ngbd of 0.

This provides a **virtual obstacle** to state feedback stabilization.

[See EDS/MCT book for proof based on homotopy.]

**Nonholonomic Integrator:** For  $\dot{x}_1 = u_1$ ,  $\dot{x}_2 = u_2$ ,  $\dot{x}_3 = x_2u_1 - x_1u_2$ , no triple  $(0, 0, \varepsilon)$  with  $\varepsilon \neq 0$  is in image of  $f$ , so no  $C^1$  stabilizer.

## OBSTACLES to CONTINUOUS STABILIZATION

**Brockett's Criterion (1983):** If  $\dot{x} = f(x, k(x))$  is GAS to 0 on  $\mathbb{R}^n$  where  $k$  is  $C^1$  and satisfies  $k(0) = 0$ , then  $\text{Image}(f)$  contains a ngbd of 0.

This provides a **virtual obstacle** to state feedback stabilization.

[See EDS/MCT book for proof based on homotopy.]

**Nonholonomic Integrator:** For  $\dot{x}_1 = u_1$ ,  $\dot{x}_2 = u_2$ ,  $\dot{x}_3 = x_2u_1 - x_1u_2$ , no triple  $(0, 0, \varepsilon)$  with  $\varepsilon \neq 0$  is in image of  $f$ , so no  $C^1$  stabilizer.

**Morse Theory:** If  $\dot{x} = f(x, k(x))$  is GAS to  $p \in \mathcal{X}$  with continuous  $k$ , then  $\mathcal{X}$  is homeomorphic to  $\mathbb{R}^n$ . **Topological obstacle** to stabilization.

## OBSTACLES to CONTINUOUS STABILIZATION

**Brockett's Criterion (1983):** If  $\dot{x} = f(x, k(x))$  is GAS to 0 on  $\mathbb{R}^n$  where  $k$  is  $C^1$  and satisfies  $k(0) = 0$ , then  $\text{Image}(f)$  contains a ngbd of 0.

This provides a **virtual obstacle** to state feedback stabilization.

[See EDS/MCT book for proof based on homotopy.]

**Nonholonomic Integrator:** For  $\dot{x}_1 = u_1$ ,  $\dot{x}_2 = u_2$ ,  $\dot{x}_3 = x_2u_1 - x_1u_2$ , no triple  $(0, 0, \varepsilon)$  with  $\varepsilon \neq 0$  is in image of  $f$ , so no  $C^1$  stabilizer.

**Morse Theory:** If  $\dot{x} = f(x, k(x))$  is GAS to  $p \in \mathcal{X}$  with continuous  $k$ , then  $\mathcal{X}$  is homeomorphic to  $\mathbb{R}^n$ . **Topological obstacle** to stabilization.

[Have Lyapunov function for closed loop system i.e. a Morse function with a unique possibly degenerate critical point. Use Milnor's Theorem.]

**DEFINITIONS for SYSTEMS on  $\mathcal{X} = \mathbb{R}^n$**

Comparison Functions  $\alpha : \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{\geq 0}$  and  $\beta : \mathbb{R}_{\geq 0} \times \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{\geq 0}$ :

## DEFINITIONS for SYSTEMS on $\mathcal{X} = \mathbb{R}^n$

Comparison Functions  $\alpha : \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{\geq 0}$  and  $\beta : \mathbb{R}_{\geq 0} \times \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{\geq 0}$ :

- $\alpha \in \mathcal{K}$  :  $\alpha$  is continuous and strictly increasing with  $\alpha(0) = 0$ .
- $\alpha \in \mathcal{K}_{\infty}$  :  $\alpha \in \mathcal{K}$  and is unbounded.
- $\alpha \in \mathcal{N}$  : nondecreasing.
- $\alpha \in \mathcal{L}$  :  $\alpha$  is continuous and asymptotically decreases to 0.
- $\beta \in \mathcal{KL}$  :  $\beta(s, \cdot) \in \mathcal{L} \forall s$  and  $\beta(\cdot, t) \in \mathcal{K} \forall t$ .

## DEFINITIONS for SYSTEMS on $\mathcal{X} = \mathbb{R}^n$

Comparison Functions  $\alpha : \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{\geq 0}$  and  $\beta : \mathbb{R}_{\geq 0} \times \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{\geq 0}$ :

- $\alpha \in \mathcal{K}$  :  $\alpha$  is continuous and strictly increasing with  $\alpha(0) = 0$ .
- $\alpha \in \mathcal{K}_{\infty}$  :  $\alpha \in \mathcal{K}$  and is unbounded.
- $\alpha \in \mathcal{N}$  : nondecreasing.
- $\alpha \in \mathcal{L}$  :  $\alpha$  is continuous and asymptotically decreases to 0.
- $\beta \in \mathcal{KL}$  :  $\beta(s, \cdot) \in \mathcal{L} \forall s$  and  $\beta(\cdot, t) \in \mathcal{K} \forall t$ .

Trajectories of  $\Sigma$ :  $x(\cdot, \xi, \mathbf{u})$  is unique maximally defined trajectory starting at  $\xi$  for the control  $\mathbf{u} \in \mathcal{U}$ .  $|\cdot|_{\mathcal{A}}$  = distance to  $\mathcal{A}$ .

$\mathcal{U}$  = measurable essentially bounded functions  $\mathbf{u} : [0, \infty) \rightarrow \mathbf{U}$ .

## DEFINITIONS for SYSTEMS on $\mathcal{X} = \mathbb{R}^n$

Comparison Functions  $\alpha : \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{\geq 0}$  and  $\beta : \mathbb{R}_{\geq 0} \times \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{\geq 0}$ :

- $\alpha \in \mathcal{K}$  :  $\alpha$  is continuous and strictly increasing with  $\alpha(0) = 0$ .
- $\alpha \in \mathcal{K}_{\infty}$  :  $\alpha \in \mathcal{K}$  and is unbounded.
- $\alpha \in \mathcal{N}$  : nondecreasing.
- $\alpha \in \mathcal{L}$  :  $\alpha$  is continuous and asymptotically decreases to 0.
- $\beta \in \mathcal{KL}$  :  $\beta(s, \cdot) \in \mathcal{L} \forall s$  and  $\beta(\cdot, t) \in \mathcal{K} \forall t$ .

Trajectories of  $\Sigma$ :  $x(\cdot, \xi, \mathbf{u})$  is unique maximally defined trajectory starting at  $\xi$  for the control  $\mathbf{u} \in \mathcal{U}$ .  $|\cdot|_{\mathcal{A}} = \text{distance to } \mathcal{A}$ .

$\mathcal{U} = \text{measurable essentially bounded functions } \mathbf{u} : [0, \infty) \rightarrow \mathbf{U}$ .

**GAC** of  $\Sigma$ :  $\exists \beta \in \mathcal{KL}$  and  $\sigma \in \mathcal{N}$  s.t.  $\forall \xi \in \mathcal{X}$ , there exists  $\mathbf{u} \in \mathcal{U}$  for which  $\|\mathbf{u}\| \leq \sigma(|\xi|_{\mathcal{A}})$  and  $|x(t, \xi, \mathbf{u})|_{\mathcal{A}} \leq \beta(|\xi|_{\mathcal{A}}, t)$  for all  $t \geq 0$ .

**GAS** of  $\Sigma$ : GAC with no controls.

## SAMPLE SOLUTIONS on $\mathcal{X} = \mathbb{R}^n$

**Partitions:**  $\pi : 0 = t_0 < t_1 < t_2 < \dots < t_n \rightarrow +\infty$ . (Sample times.)

$\bar{\mathbf{d}}(\pi) = \sup_{i \geq 0} (t_{i+1} - t_i)$  is **(upper) diameter** of  $\pi$ . **Par** = all partitions.

## SAMPLE SOLUTIONS on $\mathcal{X} = \mathbb{R}^n$

**Partitions:**  $\pi : 0 = t_0 < t_1 < t_2 < \dots < t_n \rightarrow +\infty$ . (Sample times.)

$\bar{d}(\pi) = \sup_{i \geq 0} (t_{i+1} - t_i)$  is **(upper) diameter** of  $\pi$ . **Par** = all partitions.

**$\pi$ -Trajectories:** The  **$\pi$ -trajectory**  $x_\pi(\cdot, \xi, k)$  for  $\Sigma$ ,  $\pi \in \text{Par}$ ,  $\xi \in \mathbb{R}^n$ , and feedback  $k$  is recursively defined by:

- Solving  $\dot{x}(t) = f(x(t), k(x(t_i)))$  on  $[t_i, t_{i+1}]$  (or as far as trajectory can be extended on that interval) with  $x(0) = \xi$ .
- Using the final value on previous subinterval as initial value on next subinterval (if trajectory defined on whole previous interval).

## SAMPLE SOLUTIONS on $\mathcal{X} = \mathbb{R}^n$

**Partitions:**  $\pi : 0 = t_0 < t_1 < t_2 < \dots < t_n \rightarrow +\infty$ . (Sample times.)

$\bar{d}(\pi) = \sup_{i \geq 0} (t_{i+1} - t_i)$  is **(upper) diameter** of  $\pi$ . **Par** = all partitions.

**$\pi$ -Trajectories:** The  **$\pi$ -trajectory**  $x_\pi(\cdot, \xi, k)$  for  $\Sigma$ ,  $\pi \in \text{Par}$ ,  $\xi \in \mathbb{R}^n$ , and feedback  $k$  is recursively defined by:

- Solving  $\dot{x}(t) = f(x(t), k(x(t_i)))$  on  $[t_i, t_{i+1}]$  (or as far as trajectory can be extended on that interval) with  $x(0) = \xi$ .
- Using the final value on previous subinterval as initial value on next subinterval (if trajectory defined on whole previous interval).

**Main Points:**

- Introduced by Clarke-Ledyaev-EDS-Subbotin (CLSS), IEEE/TAC'97. [Aka **sample-and-hold solutions with fast sampling.**]
- Only sample in controller  $k(x)$ .
- Allows  $k(x)$  to be discontinuous hence sample stabilization.

**CLSS THEOREM on  $\mathcal{X} = \mathbb{R}^n$**

$s$ -Stabilizing  $\Sigma$  to  $\mathcal{A}$ : This means  $\forall(r, R)$  with  $0 < r < R$ , there exist

- (i)  $M = M(R) > 0$  with  $M(R) \rightarrow 0$  as  $R \rightarrow 0$
- (ii)  $\delta = \delta(r, R) > 0$ , and  $T = T(r, R) > 0$

such that  $\forall \pi \in \text{Par}$  with  $\bar{\mathbf{d}}(\pi) < \delta$  and  $\xi \in \mathcal{B}_R(\mathcal{A})$ , the corresponding sample trajectory  $x(t) := x_\pi(t, \xi, k)$  for  $\Sigma$  satisfies:

- (a)  $|x(t)|_{\mathcal{A}} \leq r$  for all  $t \geq T$  and
- (b)  $|x(t)|_{\mathcal{A}} \leq M(R)$  for all  $t \geq 0$ .

## CLASS THEOREM on $\mathcal{X} = \mathbb{R}^n$

$s$ -Stabilizing  $\Sigma$  to  $\mathcal{A}$ : This means  $\forall(r, R)$  with  $0 < r < R$ , there exist

- (i)  $M = M(R) > 0$  with  $M(R) \rightarrow 0$  as  $R \rightarrow 0$
- (ii)  $\delta = \delta(r, R) > 0$ , and  $T = T(r, R) > 0$

such that  $\forall \pi \in \text{Par}$  with  $\bar{\mathbf{d}}(\pi) < \delta$  and  $\xi \in \mathcal{B}_R(\mathcal{A})$ , the corresponding sample trajectory  $x(t) := x_\pi(t, \xi, k)$  for  $\Sigma$  satisfies:

- (a)  $|x(t)|_{\mathcal{A}} \leq r$  for all  $t \geq T$  and
- (b)  $|x(t)|_{\mathcal{A}} \leq M(R)$  for all  $t \geq 0$ .

In other words,  $k$  stabilizes all initial values on  $\{r \leq |x|_{\mathcal{A}} \leq R\}$  to the smaller ball  $\{|x|_{\mathcal{A}} \leq r\}$  without making any ‘large excursions’.

## CLSS THEOREM on $\mathcal{X} = \mathbb{R}^n$

*s*-Stabilizing  $\Sigma$  to  $\mathcal{A}$ : This means  $\forall(r, R)$  with  $0 < r < R$ , there exist

- (i)  $M = M(R) > 0$  with  $M(R) \rightarrow 0$  as  $R \rightarrow 0$
- (ii)  $\delta = \delta(r, R) > 0$ , and  $T = T(r, R) > 0$

such that  $\forall \pi \in \text{Par}$  with  $\bar{\mathbf{d}}(\pi) < \delta$  and  $\xi \in \mathcal{B}_R(\mathcal{A})$ , the corresponding sample trajectory  $x(t) := x_\pi(t, \xi, k)$  for  $\Sigma$  satisfies:

- (a)  $|x(t)|_{\mathcal{A}} \leq r$  for all  $t \geq T$  and
- (b)  $|x(t)|_{\mathcal{A}} \leq M(R)$  for all  $t \geq 0$ .

In other words,  $k$  stabilizes all initial values on  $\{r \leq |x|_{\mathcal{A}} \leq R\}$  to the smaller ball  $\{|x|_{\mathcal{A}} \leq r\}$  without making any ‘large excursions’.

**Theorem (CLSS’97):** If  $\Sigma$  is GAC to  $\mathcal{A}$  on  $\mathcal{X} = \mathbb{R}^n$ , then  $\Sigma$  admits a feedback that *s*-stabilizes the system to  $\mathcal{A}$ .

## MAIN CONTRIBUTION: CLSS THEOREM ON MANIFOLDS

### Main Steps for Generalization to Manifolds:

- S1** Embed manifold  $\mathcal{X}$  as closed submanifold  $g(\mathcal{X}) \subseteq \mathbb{R}^k$ .
- S2** Extend  $f$  to  $\mathbb{R}^k$  so that  $g(\mathcal{X})$  is invariant for extended system and  $\mathbb{R}^k \setminus g(\mathcal{X})$  can be controlled to a tubular neighborhood of  $\mathcal{A}$ .
- S3** Apply original CLSS Theorem to embedded system on  $\mathbb{R}^k$  to get s-stabilizing feedback  $k(x)$ .
- S4** Restrict  $k(x)$  to  $g(\mathcal{X})$  to get desired feedback.

## MAIN CONTRIBUTION: CLSS THEOREM ON MANIFOLDS

### Main Steps for Generalization to Manifolds:

- S1** Embed manifold  $\mathcal{X}$  as closed submanifold  $g(\mathcal{X}) \subseteq \mathbb{R}^k$ .
- S2** Extend  $f$  to  $\mathbb{R}^k$  so that  $g(\mathcal{X})$  is invariant for extended system and  $\mathbb{R}^k \setminus g(\mathcal{X})$  can be controlled to a tubular neighborhood of  $\mathcal{A}$ .
- S3** Apply original CLSS Theorem to embedded system on  $\mathbb{R}^k$  to get s-stabilizing feedback  $k(x)$ .
- S4** Restrict  $k(x)$  to  $g(\mathcal{X})$  to get desired feedback.

**N.B.:** Must redefine GAC and s-stabilization in terms of neighborhoods instead of  $|\cdot|_{\mathcal{A}}$  for general manifold that may lack a distance!

## MAIN CONTRIBUTION: CLSS THEOREM ON MANIFOLDS

### Main Steps for Generalization to Manifolds:

- S1** Embed manifold  $\mathcal{X}$  as closed submanifold  $g(\mathcal{X}) \subseteq \mathbb{R}^k$ .
- S2** Extend  $f$  to  $\mathbb{R}^k$  so that  $g(\mathcal{X})$  is invariant for extended system and  $\mathbb{R}^k \setminus g(\mathcal{X})$  can be controlled to a tubular neighborhood of  $\mathcal{A}$ .
- S3** Apply original CLSS Theorem to embedded system on  $\mathbb{R}^k$  to get s-stabilizing feedback  $k(x)$ .
- S4** Restrict  $k(x)$  to  $g(\mathcal{X})$  to get desired feedback.

**N.B.:** Must redefine GAC and s-stabilization in terms of neighborhoods instead of  $|\cdot|_{\mathcal{A}}$  for general manifold that may lack a distance!

### Technical Tools:

- T1** Whitey Embedding Theorem.  $k = 2n + 1$ .
- T2** Tubular neighborhood theorem.

**ILLUSTRATION with  $\mathcal{A} = \{\pm\text{NP}\} \subseteq \mathbf{S}^2$**

Dynamics:

$$\dot{x} = A_1(x)u_1 + A_2(x)u_2 \in T_x(\mathcal{X}), \quad u \in \mathbb{R}^2, \quad \mathcal{X} = S^2 \quad (\Sigma')$$

where  $A_1(x) = M_1(x)B_1(x)$ ,  $A_2(x) = B_2(x)$ ,  $B_1(x) = q - (x \cdot q)x$ ,  $B_2(x) = x \times q$ ,  $q = \text{NP}$ , and  $M_1$  introduces a set of zeros in a geodesic rectangle covering part of the equator in  $\{x \in S^2 : x_1 > 0, x_2 > 0\}$ .

## ILLUSTRATION with $\mathcal{A} = \{\pm\text{NP}\} \subseteq S^2$

Dynamics:

$$\dot{x} = A_1(x)u_1 + A_2(x)u_2 \in T_x(\mathcal{X}), \quad u \in \mathbb{R}^2, \quad \mathcal{X} = S^2 \quad (\Sigma')$$

where  $A_1(x) = M_1(x)B_1(x)$ ,  $A_2(x) = B_2(x)$ ,  $B_1(x) = q - (x \cdot q)x$ ,  $B_2(x) = x \times q$ ,  $q = \text{NP}$ , and  $M_1$  introduces a set of zeros in a geodesic rectangle covering part of the equator in  $\{x \in S^2 : x_1 > 0, x_2 > 0\}$ .

**Main Points:**  $\{B_1, B_2\} = \text{basis for } T_x(\mathcal{X}) = \{x\}^\top$  outside  $\mathcal{A}$ .  $M_1$  limits motions. We assume  $M$  is valued in  $[0, 1]$ . **Discontinuous stabilizer:**

$$u = K(x) = \begin{cases} \langle 1, 0 \rangle, & \text{if } x_3 \geq 0 \text{ and } M_1(x) = 1 \\ -\langle 1, 0 \rangle, & \text{if } x_3 < 0 \text{ and } M_1(x) = 1 \\ \langle 0, 1 \rangle, & \text{if } M_1(x) < 1 \end{cases}$$

## ILLUSTRATION with $\mathcal{A} = \{\pm\text{NP}\} \subseteq \mathbf{S}^2$

Dynamics:

$$\dot{x} = A_1(x)u_1 + A_2(x)u_2 \in T_x(\mathcal{X}), \quad u \in \mathbb{R}^2, \quad \mathcal{X} = S^2 \quad (\Sigma')$$

where  $A_1(x) = M_1(x)B_1(x)$ ,  $A_2(x) = B_2(x)$ ,  $B_1(x) = q - (x \cdot q)x$ ,  $B_2(x) = x \times q$ ,  $q = \text{NP}$ , and  $M_1$  introduces a set of zeros in a geodesic rectangle covering part of the equator in  $\{x \in S^2 : x_1 > 0, x_2 > 0\}$ .

**Main Points:**  $\{B_1, B_2\} = \text{basis for } T_x(\mathcal{X}) = \{x\}^\top$  outside  $\mathcal{A}$ .  $M_1$  limits motions. We assume  $M$  is valued in  $[0, 1]$ . **Discontinuous stabilizer:**

$$u = K(x) = \begin{cases} \langle 1, 0 \rangle, & \text{if } x_3 \geq 0 \text{ and } M_1(x) = 1 \\ -\langle 1, 0 \rangle, & \text{if } x_3 < 0 \text{ and } M_1(x) = 1 \\ \langle 0, 1 \rangle, & \text{if } M_1(x) < 1 \end{cases}$$

A corresponding explicit Lyapunov function can also be constructed.

## LYAPUNOV FUNCTION for SPHERE EXAMPLE

In terms of  $\mathcal{G}(x, x') := \arccos(x \cdot x')$ ,  $q = \text{NP}$ ,  $r = \langle 0, 1, 0 \rangle$ ,  
 $V_q(x) = \min\{\mathcal{G}(x, \bar{q}) : \bar{q} = \pm q\}$ ,  $V_r(x) = \max\{\mathcal{G}(x, \bar{r}) : \bar{r} = \pm r\}$ ,

$$V(x) = V_q(x)[1 + V_r(x)], \quad x \in S^2.$$

## LYAPUNOV FUNCTION for SPHERE EXAMPLE

In terms of  $\mathcal{G}(x, x') := \arccos(x \cdot x')$ ,  $q = \text{NP}$ ,  $r = \langle 0, 1, 0 \rangle$ ,  
 $V_q(x) = \min\{\mathcal{G}(x, \bar{q}) : \bar{q} = \pm q\}$ ,  $V_r(x) = \max\{\mathcal{G}(x, \bar{r}) : \bar{r} = \pm r\}$ ,

$$V(x) = V_q(x)[1 + V_r(x)], \quad x \in S^2.$$

### Main Points:

- $V_r$  part of  $V$  puts high cost on being near  $\pm r$ .
- $V$  continuous, only zero on  $\mathcal{A}$ , nonsmooth
- Control-Lyapunov integral function: There exist  $N > 0$  and  $\alpha_3 \in \mathcal{K}$  satisfying:  $\forall \xi \in \mathcal{X}$ ,  $\exists u \in \mathcal{U}_N$  s.t.  $x(t) := x(t, \xi, u)$  satisfies

$$V(x(t)) - V(\xi) \leq - \int_0^t \alpha_3(|x(s)|_{\mathcal{A}}) ds \quad \forall t \geq 0.$$

- $\mathcal{U}_N$  -  $\{u \in \mathcal{U} : \|u\| \leq N\}$ .

## EXTENSIONS/CONCLUSIONS

- Can also sample **input-to-state stabilize** control **affine systems** with observation errors on manifolds a la [M-Rifford-EDS, SICON'04].
  - Use semiconcave CLF, with Riemannian metric to quantify observation error.
  - Sampling is done sufficiently fast but not *too* fast.
  - See EDS'99/NATO-ASI or COCV paper for  $\mathcal{X} = \mathbb{R}^n$  case.

## EXTENSIONS/CONCLUSIONS

- Can also sample **input-to-state stabilize** control **affine systems** with observation errors on manifolds a la [M-Rifford-EDS, SICON'04].
  - Use semiconcave CLF, with Riemannian metric to quantify observation error.
  - Sampling is done sufficiently fast but not *too* fast.
  - See EDS'99/NATO-ASI or COCV paper for  $\mathcal{X} = \mathbb{R}^n$  case.
- Can make **fully nonlinear systems** s-ISS in weak sense of EDS IEEE/TAC'90 “Further Facts..” paper on Riemannian manifolds.

## EXTENSIONS/CONCLUSIONS

- Can also sample **input-to-state stabilize** control **affine systems** with observation errors on manifolds a la [M-Rifford-EDS, SICON'04].
  - Use semiconcave CLF, Riemannian metric to quantify observation error.
  - Sampling is done sufficiently fast but not *too* fast.
  - See EDS'99/NATO-ASI or COCV paper for  $\mathcal{X} = \mathbb{R}^n$  case.
- Can make **fully nonlinear systems** s-ISS in weak sense of EDS IEEE/TAC'90 “Further Facts..” paper on Riemannian manifolds.
- Alternative Approach
  - No embedding. **Intrinsic**.
  - Build CLFs and feedbacks on manifolds as we did for  $S^2$  example.
  - Ongoing joint research by M with Ludovic Rifford.

## ACKNOWLEDGEMENTS

Malisoff was supported in part by the Louisiana Board of Regents Research and Development Program and the NSF.



Sontag was supported by NSF Grant CCR-0206789.

Krichman's work on this paper was done while he was working for Alphatech Inc. in Burlington, MA.

