



#### **Adaptive Control and Applications**

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# Control of Uncertain Constrained Nonlinear Systems with Barrier Lyapunov Function

in Collaboration with K.P. Tee, I2R, A\*Star, Singapore



#### The Chernobyl Incident, 1986





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## Objectives



- To develop constructive and systematic methods of designing adaptive controllers for uncertain constrained nonlinear systems
- To provide technically rigorous framework for the use of Barrier Lyapunov Functions in ensuring constraint satisfaction
- To illustrate practical relevance through an application study





## **Barrier Lyapunov Functions**



#### Barrier Functions

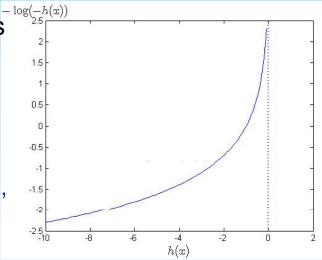


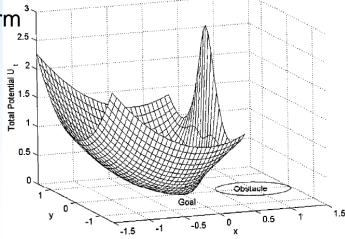
A barrier function is a continuous function b(x)
 whose value increases to infinity as x approaches
 the boundary of the feasible region

- Constrained optimization problems
  - (Nocedal & Wright, 1999)
- Collision avoidance problem in robotics
  - (Rimon and Koditschek, 1992), (Ge and Cui, 2000),
     (Gazi and Passino, 2004), (Do, 2007)



- (Ngo, Mahony & Jiang, 2005)
- Considered known system in Brunowski normal form subjected to state constraint
- States assumed to be constrained before being proved so





#### Barrier Lyapunov Function (BLF)



#### **Definition:**

A BLF is a scalar function V(x), defined with respect to

$$\dot{x} = f(x)$$
,  $x(0) \in D \subset \mathbb{R}^n$  (1)

on an open region D containing the origin, with the properties:

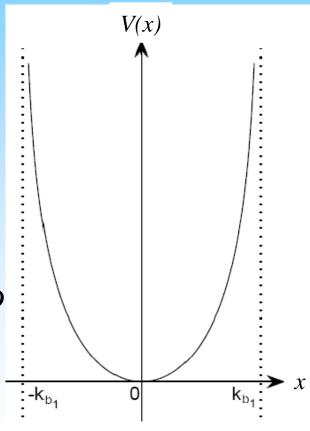
- continuous, positive definite, and continuously differentiable at every point of  ${\it D}$ 

$$- V(x) \rightarrow \infty$$

as x approaches the boundary of D, and

$$- V(x(t)) \le b \quad \forall t > 0$$

along the solution of (1), where b is a positive constant.



#### Technical Lemma



Suppose that there exist continuously differentiable positive definite functions  $U: R^l \to R_+$  and  $V_i: Z_i \to R_+$ , i = 1,...,n, such that

$$V_i(z_i) \to \infty$$
 as  $z_i \to -k_{a_i}$  or  $z_i \to k_{b_i}$   $\gamma_1(\|w\|) \le U(w) \le \gamma_2(\|w\|)$ 

where  $\gamma_1$ ,  $\gamma_2$  are class  $K_{\infty}$  functions. Let  $V = \sum_{i=1}^n V_i(z_i) + U(w)$ .

For initial conditions  $z_i(0)$  starting from  $z_i \in (-k_{a_i}, k_{b_i}), i = 1,...,n$ , if the following inequality holds

$$\dot{V} = \frac{\partial V}{\partial z} h \le 0$$

then  $z_i(t)$  remains in the open set  $z_i \in (-k_{a_i}, k_{b_i}), \forall t \in [0, \infty)$ .

#### Technical Lemma



#### **Proof Logic:**

- i) From conditions on h, existence and uniqueness of solution z(t) is ensured on  $t \in [0, \tau_{\text{max}})$ , based on (Sontag, 1998).
- ii) From  $\dot{V} \leq 0$ , we know that each  $V_i(z_i(t))$  is bounded for  $t \in [0, \tau_{\text{max}})$ , which implies that

$$-k_{a_i} < z_i(t) < k_{b_i}$$
 for  $t \in [0, \tau_{\text{max}})$ .

iii) Finally, show that  $\tau_{\max} = \infty$ . From the boundedness of  $V(\eta(t))$ , we can show that  $\eta(t)$  belongs to a compact set for  $t \in [0, \tau_{\max})$ . Based on (Sontag, 1998),  $\eta(t)$  is defined for all  $t \in [0, \infty)$ . Therefore,  $-k_{a_i} < z_i(t) < k_{b_i}$  for  $t \in [0, \infty)$ .

9





# Control of Output-Constrained Nonlinear Systems

### Output Constraint Problem



#### Consider nonlinear systems in strict feedback form

$$\dot{x}_{i} = \theta^{T} \psi_{i}(\overline{x}_{i}) + g_{i}(\overline{x}_{i}) x_{i+1}, \qquad i = 1, 2, ..., n-1$$

$$\dot{x}_{n} = \theta^{T} \psi_{n}(\overline{x}_{n}) + g_{n}(\overline{x}_{n}) u$$

$$y = x_{1}$$

#### where

 $\overline{x}_i := [x_1, x_2, ..., x_i]^T \in R^i$  are the states  $\theta$  is a vector of uncertain parameters  $\psi_i$  and  $g_i$  are known smooth functions  $u \in R$  is the input,  $y \in R$  is the output

#### **Control Objectives:**

- 1. Track desired trajectory  $y_d(t)$
- 2. Ensure output constraint  $|y| < k_{c_1}$  is met for all time
- 3. All closed loop signals bounded

### Assumptions



A.1.1. There exist positive constants  $\underline{Y}_0, \overline{Y}_0, A_0, Y_1, \dots, Y_n$  such that

$$\begin{aligned} & \max\{\underline{Y}_0, \overline{Y}_0\} \leq \mathsf{A}_0 < k_{c_1} \\ & -\underline{Y}_0 \leq y_d(t) \leq \overline{Y}_0 \\ & |\dot{y}_d(t)| \leq Y_1, \quad \dots, \quad |y_d^{(n)}(t)| \leq Y_n \end{aligned}$$

A.1.2. The control gain functions  $g_i(\cdot)$ , i = 1, 2, ..., n, are known, and there exists a positive constant  $g_0$  such that

$$0 < g_0 \le |g_i(\cdot)|, \qquad i = 1, 2, ..., n$$

Without loss of generality, we further assume that  $g_i(\cdot)$  are positive.

#### Motivating Example: Second Order System



Consider the system:

$$\dot{x}_1 = f_1(x_1) + g_1(x_1)x_2$$

$$\dot{x}_2 = f_2(x_1, x_2) + g_1(x_1, x_2)u$$

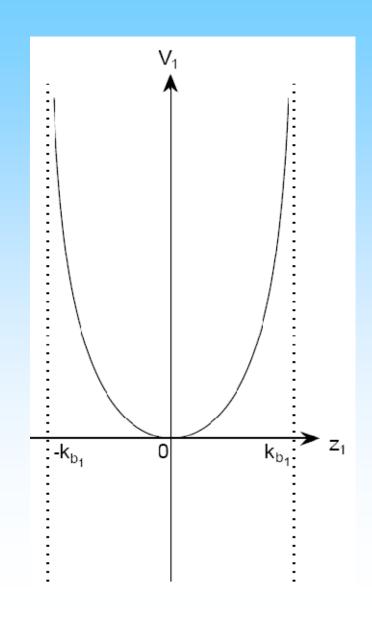
Step 1: Denote  $z_1 := x_1 - y_d$ .

Barrier Lyapunov Function candidate:

$$V_1 = \frac{1}{2} \log \frac{k_{b_1}^2}{k_{b_1}^2 - z_1^2}$$

Design stabilizing function:

$$\alpha_{1} = \frac{1}{g_{1}} \left( -f_{1} - \left( k_{b_{1}}^{2} - z_{1}^{2} \right) \kappa_{1} z_{1} + \dot{y}_{d} \right)$$



Known

#### Motivating Example: Known Second Order System



#### Step 2:

Lyapunov Function candidate:

$$V_2 = V_1 + \frac{1}{2}z_2^2$$

repels  $z_1$  away from barrier

Control law:

$$u = \frac{1}{g_2} \left( -f_2 + \dot{\alpha}_1 - \kappa_2 z_2 - \frac{g_1 z_1}{k_{b_1}^2 - z_1^2} \right)$$

Derivative of  $V_2$  along closed loop trajectories:

$$\dot{V}_2 = -\sum_{i=1}^2 \kappa_i z_i^2$$

#### Motivating Example: Second Order System

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#### Closed loop error dynamics:

$$\dot{z}_1 = -\left(k_{b_1}^2 - z_1^2\right) \kappa_1 z_1 + g_1 z_2$$

$$\dot{z}_2 = -\kappa_2 z_2 - \frac{g_1 z_1}{k_{b_1}^2 - z_1^2}$$

R.H.S. is locally Lipschitz in the set  $|z_1| < k_{\rm b1}$ 

Together with  $\dot{V_2} \leq 0$  and  $|z_1(0)| < k_{\rm b1}$ , we can invoke the Technical Lemma to show

Known

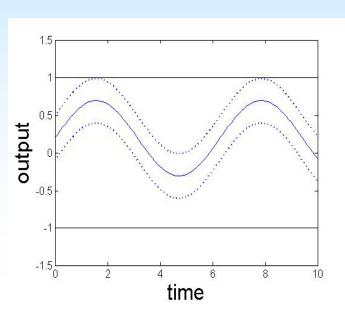
$$|z_1(t)| < k_{b_1} \text{ for } t \in [0, \infty)$$

Transform bounds to  $x_1$  coordinates:

$$|x_1(t)| \le |y_d(t)| + |z_1(t)|$$

$$\le A_0 + k_{b_1}$$

$$< k_{c_1} \quad \text{for } t \in [0, \infty)$$



#### Adaptive Control Design



Adaptive Backstepping Step 1: Barrier Lyapunov function candidate

$$V_{1} = \frac{1}{2} \log \frac{k_{b_{1}}^{2}}{k_{b_{1}}^{2} - z_{1}^{2}} + \frac{1}{2} \tilde{\theta}^{T} \Gamma^{-1} \tilde{\theta}, \quad \text{where } k_{b_{1}} = k_{c_{1}} - A_{0}$$

Step 2 to n: Quadratic Lyapunov function candidates

$$V_i = V_{i-1} + \frac{1}{2}z_i^2, \quad i = 2,...,n$$

Design stabilizing functions  $\alpha_i$ , i=1,...,n-1, control and adaptation laws to obtain

$$\dot{V}_n \leq 0$$

Constraint Satisfaction

Invoke Technical Lemma to show that  $|z_1(t)| < k_{b1}$ 

From Assumption 1.1, it follows that  $|x_1(t)| < k_{c1}$ 

#### Main Results: Symmetric BLF



Consider the closed loop system under the proposed control.

Given that  $z_1(0) < k_{b_1}$ , we have the results:

i) The signals  $z, \hat{\theta}$  remain in the compact sets:

$$\begin{split} &\Omega_z = \left\{z \in R^n: \ |z_1| \leq k_{b_1} \sqrt{1 - e^{-2\overline{V_n}}}, \ \|z_{2:n}\| \leq \sqrt{2\overline{V_n}}\right\} \\ &\Omega_{\hat{\theta}} = \left\{\hat{\theta} \in R^l: \ |\hat{\theta}| \leq \theta_M + \sqrt{\frac{2\overline{V_n}}{\lambda_{\min}(\Gamma^{-1})}}\right\} \end{split}$$

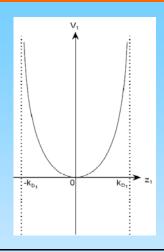
where

$$\overline{V}_{n} = \frac{1}{2} \log \frac{k_{b_{1}}^{2}}{k_{b_{1}}^{2} - z_{1}^{2}(0)} + \frac{1}{2} \sum_{j=2}^{n} z_{j}^{2}(0) + \frac{1}{2} \lambda_{\max}(\Gamma^{-1}) (\|\hat{\theta}(0)\| + \theta_{M})$$

- ii) The output is constrained  $|y| \le D_{z_1} + A_0 < k_{c_1}$ ,  $\forall t > 0$
- iii) All closed loop signals are bounded
- iv) Asymptotic output tracking  $y(t) \rightarrow y_d(t)$  as  $t \rightarrow \infty$

#### Symmetric vs Asymmetric BLFs

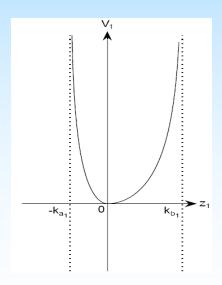




#### **Symmetric**

$$V_1 = \frac{1}{2} \log \frac{k_{b_1}^2}{k_{b_1}^2 - z_1^2}$$

yields smooth stabilizing functions  $\alpha_i$ , i=1,...,n



Asymmetric
$$V_{1} = \frac{1}{p} q(z_{1}) \log \frac{k_{b_{1}}^{p}}{k_{b_{1}}^{p} - z_{1}^{p}} + \frac{1}{p} (1 - q(z_{1})) \log \frac{k_{a_{1}}^{p}}{k_{a_{1}}^{p} - z_{1}^{p}}$$

$$q(\square) = \begin{cases} 1, & \text{if } \square > 0 \\ 0, & \text{if } \square \le 0 \end{cases}$$

To ensure differentiability of  $\alpha_i$ , i=1,...,n:

- choose even integer  $p \ge n$
- careful design of  $\alpha_1$  and  $\alpha_2$

#### Asymmetric BLF-based Design



#### Stabilizing functions and control law:

$$\alpha_{1} = \frac{1}{g_{1}} \left[ -f_{1} - \left( q(k_{b_{1}}^{p} - z_{1}^{p}) + (1 - q)(k_{a_{1}}^{p} - z_{1}^{2}) \right) \kappa_{1} z_{1}^{m} + \dot{y}_{d} \right]$$

$$\alpha_{2} = \frac{1}{g_{2}} \left[ -f_{2} - \kappa_{2} z_{2} + \dot{\alpha}_{1} - \left( \frac{q}{k_{b_{1}}^{p} - z_{1}^{p}} + \frac{1 - q}{k_{a_{1}}^{p} - z_{1}^{p}} \right) g_{1} z_{1}^{p-1} \right]$$

$$\alpha_{i} = \frac{1}{g_{i}} \left( -f_{i} - \kappa_{i} z_{i} + \dot{\alpha}_{i-1} - g_{i-1} z_{i-1} \right), \quad i = 3, ..., n$$

$$u = \alpha_{n}$$

Choosing  $\int \text{even integer } p \ge n$  odd integer  $m \ge \max\{3, n\}$ 

#### ensures that

$$\alpha_i(\overline{x}_i, \overline{z}_i, \overline{y}_{d_i})$$
 is  $C^{n-i}$  in the set  $z_1 \in (-k_{a_i}, k_{b_i})$ 

#### Main Results: Asymmetric BLF



Consider the closed loop system under the proposed control.

Given that  $-k_{a_1} < z_1(0) < k_{b_1}$ , we have the results:

i) The signals  $z, \hat{\theta}$  remain in the compact sets:

$$\Omega_{z} = \left\{ z \in R^{n} : -\underline{D}_{z_{1}} \leq z_{1} \leq \overline{D}_{z_{1}}, \|z_{2:n}\| \leq \sqrt{2\overline{V}_{n}} \right\}$$

$$\Omega_{\hat{\theta}} = \left\{ \hat{\theta} \in R^{l} : |\hat{\theta}| \leq \theta_{M} + \sqrt{\frac{2\overline{V}_{n}}{\lambda_{\min}(\Gamma^{-1})}} \right\}$$

where 
$$\overline{V_n} := \frac{1}{p} q \log \frac{k_{b_1}^p}{k_{b_1}^p - z_1^p(0)} + \frac{1}{p} (1 - q) \log \frac{k_{a_1}^p}{k_{a_1}^p - z_1^p(0)} + \frac{1}{2} \sum_{i=2}^n z_j^2(0) + \frac{1}{2} \lambda_{\max}(\Gamma^{-1}) (\|\hat{\theta}(0)\| + \theta_M)^2$$

$$\underline{D}_{z_1} := k_{b_1} \left( 1 - e^{-2\overline{V}_n} \right)^{\frac{1}{p}}$$

$$\overline{D}_{z_1} := k_{a_1} \left( 1 - e^{-2\overline{V}_n} \right)^{\frac{1}{p}}$$

- ii) All closed loop signals are bounded
- iii) The output is constrained  $|y(t)| \le D_{z_1} + A_0 < k_{c_1}, \quad \forall t > 0$
- iv) Asymptotic output tracking  $y(t) \rightarrow y_d(t)$  as  $t \rightarrow \infty$

# Comparison with Quadratic Lyapunov Functions



QLF candidates: 
$$V_1 = \frac{1}{2}z_1^2 + \frac{1}{2}\tilde{\theta}_1^T\Gamma_1^{-1}\tilde{\theta}_1$$
 
$$V_i = V_{i-1} + \frac{1}{2}z_i^2 + \frac{1}{2}\tilde{\theta}_i^T\Gamma_i^{-1}\tilde{\theta}_i, \quad i = 2,...,n$$

Standard adaptive backstepping (Krstic et al, 1995):

$$\dot{V}_n = -\sum_{i=1}^n \kappa_i z_i^2$$

$$\dot{Z}_1^2(t) \leq \sum_{i=1}^n \left[ z_i^2(0) + \lambda_{\max} \left( \Gamma^{-1} \right) \left( \left\| \hat{\theta}_i(0) \right\| + \theta_M \right)^2 \right]$$

Sufficient initial conditions to ensure  $|z_1(t)| \le k_{b_1}$ :

$$\begin{cases} \|\overline{z}_{n}(0)\| \leq \sqrt{k_{b_{1}}^{2} - \lambda_{\max}\left(\Gamma^{-1}\right) \sum_{i=1}^{n} \left(\left\|\hat{\theta}_{i}(0)\right\| + \theta_{M}\right)^{2}} \\ k_{b_{1}}^{2} > \lambda_{\max}\left(\Gamma^{-1}\right) \sum_{i=1}^{n} \left(\left\|\hat{\theta}_{i}(0)\right\| + \theta_{M}\right)^{2} \end{cases}$$

More conservative than that of BLF-based design:  $|z_1(0)| \le k_{b_1}$ 

#### Numerical Example



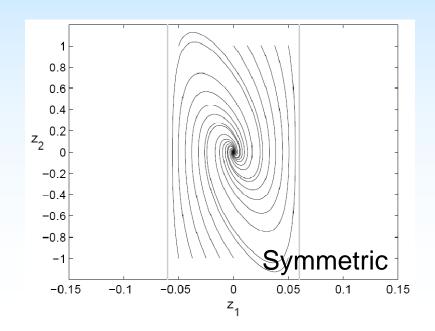
Plant: 
$$\dot{x}_1 = \theta_1 x_1^2 + x_2$$
  
 $\dot{x}_2 = \theta_2 x_1 x_2 + \theta_3 x_1 + (1 + x_1^2) u$ 

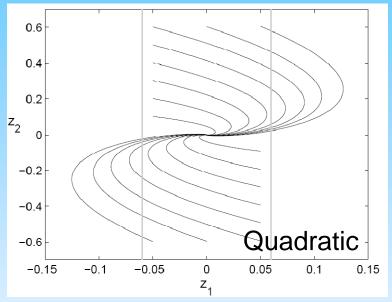
subject to constraint:

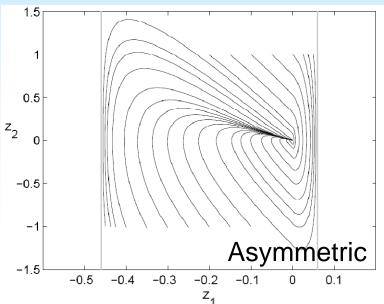
$$|x_1| < 0.56$$

Desired trajectory:

$$y_d(t) = 0.2 + 0.3 \sin t$$



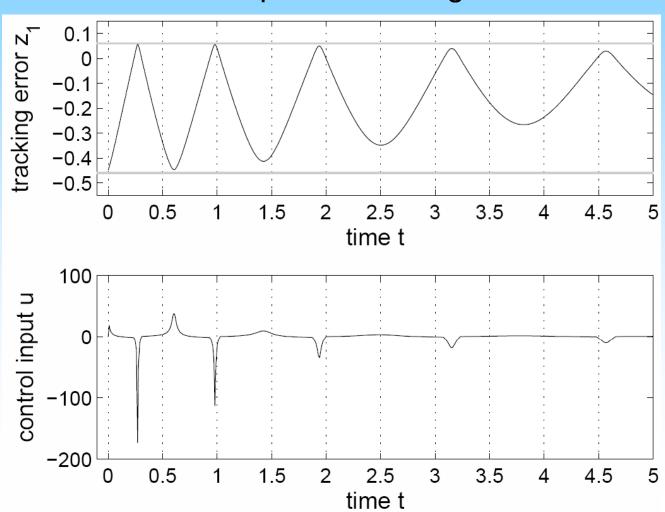




#### Numerical Example



## Control increases rapidly near the barriers to prevent transgression







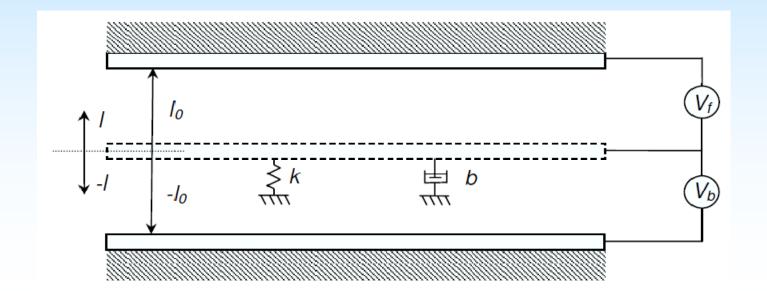
# Application Study: Control of Electrostatic Microactuators

#### Electrostatic Microactuator with Bi-Directional Drive



$$m\ddot{l} + b(l)\dot{l} + kl = \frac{\varepsilon A}{2} \left( \frac{V_f^2}{(l_0 - l)^2} - \frac{V_b^2}{(l_0 + l)^2} \right) =: \frac{\varepsilon A}{2} v$$

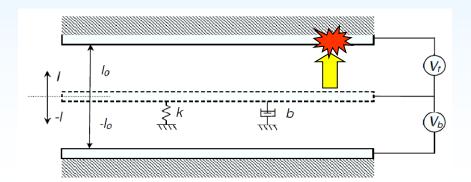
- Parameters unknown
- Track reference trajectory within gap
- Ensure electrodes do not contact (output constraint)



#### Related Works



- Extending travel range
  - (Chu & Pister, 1994), (Seeger & Crary, 1997), (Chan & Dutton, 2000)
- Nonlinear control of MEMs
  - (Zhu et al, 2005), (Maithripala et al, 2005)
- Adaptive control of MEMs
  - (Shkel et al, 1999), (Park and Horowitz, 2003), (Leland, 2001),
     (Piyabongkarn et al, 2005)
- Problem of electrode contact is not rigorously treated



#### Future Research



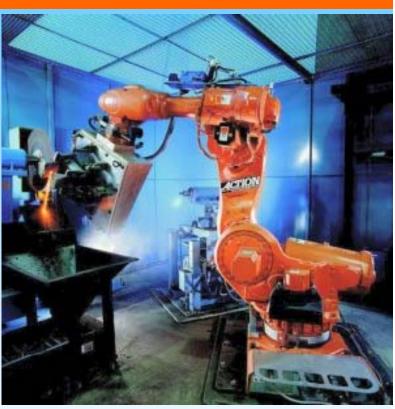
- Different choices of Barrier Lyapunov Functions
- States Constrained nonlinear systems
- Practical applications

### **Industrial Robotics**



- Rigid, Bulky and Heavy
- Steel Cold





#### Modern Robotics



• Intelligent autonomous vehicles

• Semi-intelligent autonomous vehicles







#### Autonomous Robot X1

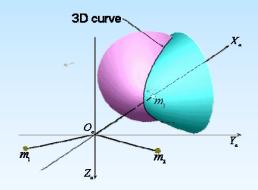
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#### Robust Audio Localization



1. Mask Diffraction







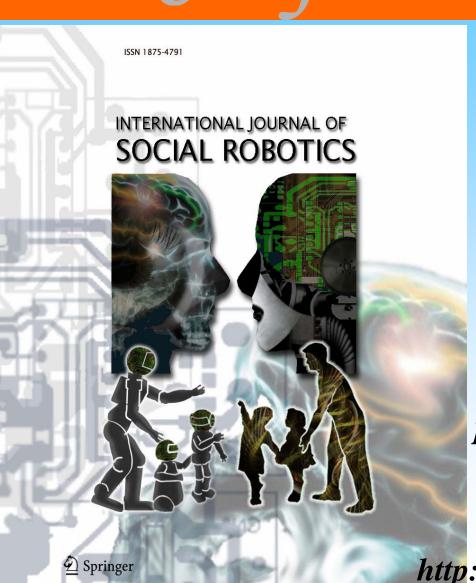




video

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