

# **High Relative Degree Control Barrier Functions Under Input Constraints**

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- System:  $\dot{x} = f(x) + g(x)u$ , where  $x \in \mathbb{R}^n, u \in U \subset \mathbb{R}^m$  where  $U$  is compact
- Safe set:  $S = \{x \in \mathbb{R}^n \mid h(x) \leq 0\}$  for  $h : \mathbb{R}^n \rightarrow \mathbb{R}$ 
  - Goal is to render trajectories always inside  $S$
- Assume  $f, g, h$  are  $r$ -times continuously differentiable, where  $r$  is the relative-degree of  $h$  (lowest  $r$  such that  $h^{(r)}$  depends on  $u$ )



- Control Barrier Functions (CBFs) are a method to certify existence of safe control inputs
- $S$  is rendered forward invariant if and only if  $\dot{h}(x, u) = \frac{\partial h(x)}{\partial x} \dot{x} \leq 0$  for all  $x \in \partial S$  (Nagumo's Theorem)
  - In practice, enforce  $\dot{h}(x, u) \leq \alpha(-h(x))$  for all  $x \in S$

### Definition 1

A continuously differentiable function  $h : \mathbb{R}^n \rightarrow \mathbb{R}$  is a Control Barrier Function (CBF) on set  $S$  for control set  $U$  if there exists  $\alpha \in \mathcal{K}$  such that

$$\inf_{u \in U} [\dot{h}(x, u) - \alpha(-h(x))] \leq 0, \quad \forall x \in S.$$

## Example



- Double-integrator with a position constraint

$$\dot{x} = v$$

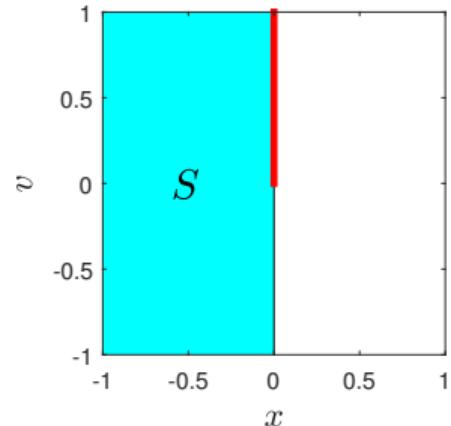
$$\dot{v} = u$$

$$h([x, v]^T) = x$$

$$S = \{(x, v) \in \mathbb{R}^2 \mid x \leq 0\}$$

$$\dot{h}([x, v]^T) = v \rightarrow \exists(x, v) \in \partial S : \dot{h}([x, v]^T) \not\leq 0$$

- $h$  is not a CBF
- $\dot{h}$  does not depend on  $u$ , so  $h$  is of “high-relative-degree” (i.e.  $r > 1$ )



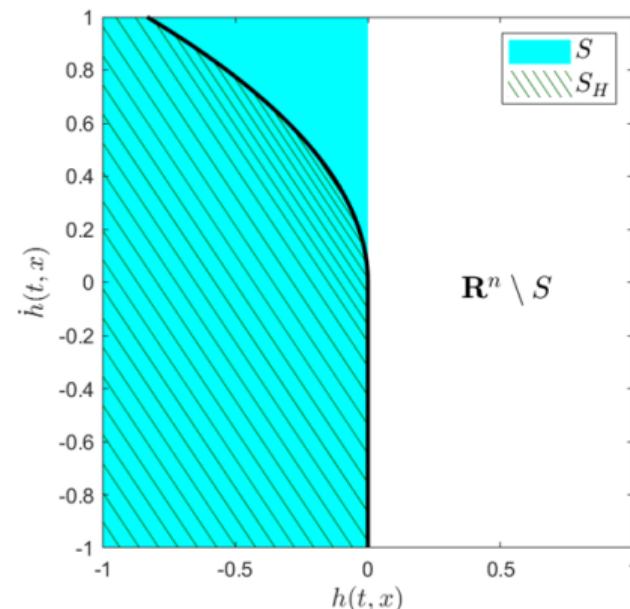


- Fact:  $H = h + \dot{h}$  is a CBF for the prior example, provided no input constraints (i.e.  $U = \mathbb{R}^m$ )
- Prior work on converting high-relative-degree  $h$  to CBFs
  - Backstepping approach (Hsu, Xu, Ames, ACC 2015)
  - Exponential CBFs (Nguyen, Sreenath, ACC 2016)
  - Compositions with  $h^{(r)}$  (Ames, Xu, Grizzle, Tabuada, TAC 2017)
  - Backup Controllers (Squires, Pierpaoli, Egerstedt, CCTA 2018)
  - Higher Order CBFs (Xiao, Belta, CDC 2019)

# Contribution



- Suggest two forms of  $H : \mathbb{R}^n \rightarrow \mathbb{R}$  that are CBFs in the presence of input constraints, where  $H(x) \geq h(x)$  for all  $x \in \mathbb{R}^n$  so that  $S_H = \{x \in \mathbb{R}^n \mid H(x) \leq 0\} \subset S$
- $S_H$  = the “inner safe set” = set of allowable initial conditions
- Existence of a CBF implies we can render  $S_H$  forward invariant



- For some policy  $u^* : \mathbb{R}^n \rightarrow U$ , define  $\psi_x(t; x, u^*) = y(t)$  according to the initial value problem

$$\dot{y} = f(y) + g(y)u^*(y), \quad y(0) = x$$

and  $\psi_h(t; x, u^*) = h(\psi_x(t; x, u^*))$

- E.g.  $u_{\text{ball}}^*(x) = \arg \min_{u \in U} h^{(r)}(x, u) = \arg \min_{u \in U} L_g L_f^{r-1} h(x) u$
- $u^*$  called the “nominal evading maneuver” in [Squires, Pierpaoli, Egerstedt, CCTA 2018]
- We do not need closed-form expressions for  $\psi_x, \psi_h$

- Define  $H(x) \triangleq \sup_{t \geq 0} \psi_h(t; x, u^*)$

### Assumption 1

Assume  $H$  exists and is differentiable everywhere in  $S$ .

### Theorem 1

$H$  is a CBF on the set  $S_H$  for the control set  $U$ , provided  $S_H$  is nonempty.

- We do not need closed-form expressions for  $H$
- The CBF condition  $\dot{H}(x, u) \leq \alpha(-H(x))$  is still control-affine

- Let  $u' : \mathbb{R}^n \rightarrow U$  be a policy such that

$$h^{(r)}(x, u'(x)) = -a_{max}, \forall x \in S$$

for some fixed  $a_{max} \in \mathbb{R}_{>0}$  (provided  $a_{max}$  exists).

- One such  $a_{max}$  is

$$a_{max} \triangleq \max \left( \left\{ a \in \mathbb{R} \mid \forall x \in S, \exists v \in (L_g L_f^{r-1} h(x))^{\perp} : -\frac{(a + L_f^r h(x))(L_g L_f^{r-1} h(x))}{\|L_g L_f^{r-1} h(x)\|^2} + v \in U \right\} \right)$$

- $\psi_h(t; x, u')$  is a polynomial in  $t$

$$\psi_h(t; x, u') = \sum_{i=0}^{r-1} \frac{1}{i!} h^{(i)}(x) t^i - \frac{1}{r!} a_{max} t^r$$

- Define  $H'(x) \triangleq \sup_{t \geq 0} \psi_h(t; x, u')$
- Existence and differentiability of  $H'$  are guaranteed

### Theorem 2

*$H'$  is a CBF on the set  $S_{H'}$  for the control set  $U$ , provided  $S_{H'}$  is nonempty.*

- Method 1 requires propagating a  $n \times 1$  and a  $n \times n$  ordinary differential equation
- Method 2 requires finding the roots of a  $(r - 1)$ -dimensional polynomial

- Double integrator with a spherical exclusion region

$$\dot{x} = \begin{bmatrix} \dot{r} \\ \dot{\vartheta} \end{bmatrix} = \begin{bmatrix} v \\ u \end{bmatrix}$$

$$r, v, \in \mathbb{R}^3, \quad u \in U = \{u \in \mathbb{R}^3 \mid \|u\|_\infty \leq u_{max}\}$$

- $h(x) = \rho - \|r - r_s\|$  for fixed  $r_s \in \mathbb{R}^3$
- $a_{max} = u_{max}$

- Lyapunov function

$$V(x) = \frac{1}{2}||r - r_p||^2 + \frac{1}{2}k_2||v - k_1(r - r_p)||^2$$

where  $r_p \in \mathbb{R}^3$  is a target location.

- Control law:

$$u(x) = \arg \min_{u \in U, \delta \in \mathbb{R}} u^T u + J\delta^2 \quad \text{such that}$$

$$L_f H(x) + L_g H(x)u \leq \alpha(-H(x))$$

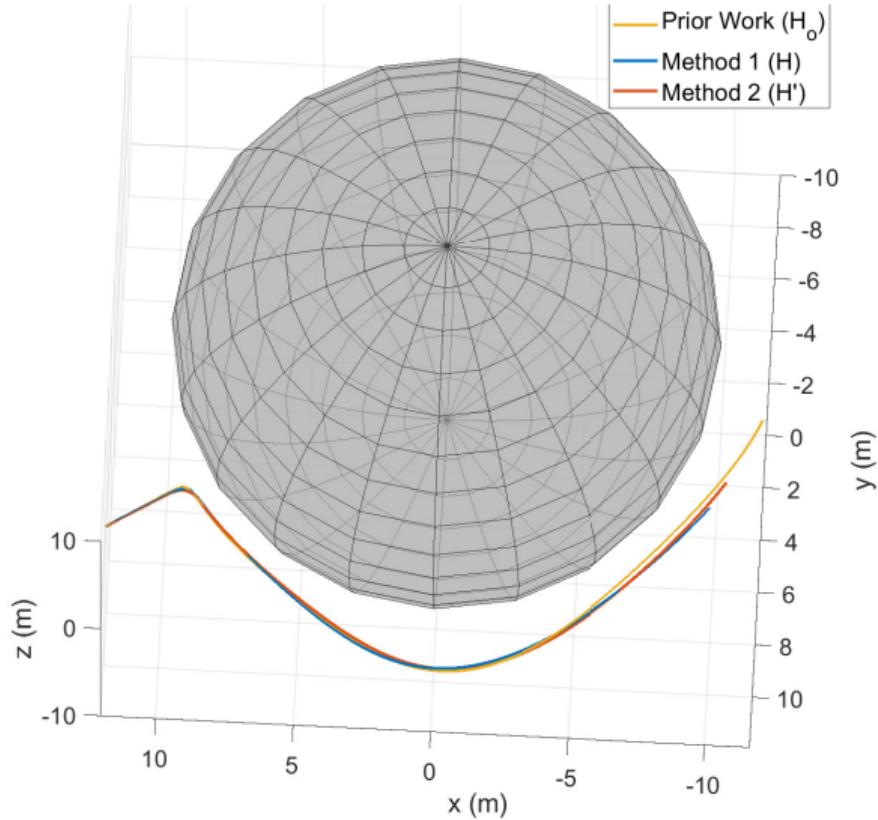
$$L_f V(x) + L_g V(x)u + \delta \leq -k_3 V(x)$$

# Application 1 Results



- Comparison CBF (no guarantee of input constraint satisfaction) from [Ames, Xu, Grizzle, Tabuada, TAC 2017]

$$H_o(x) = \left( \arctan(\dot{h}(x)) + \frac{\pi}{2} \right) h(x)$$



# Application 1 Results

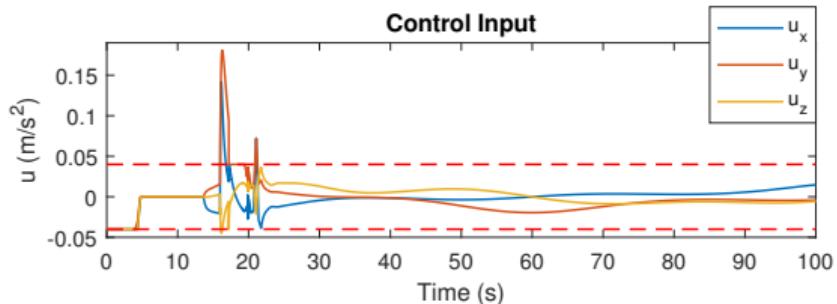


Figure: The control input using  $H_o(x)$  from prior work, which necessitates using control inputs outside the prescribed bounds (dashed red lines) for the QP to have a solution

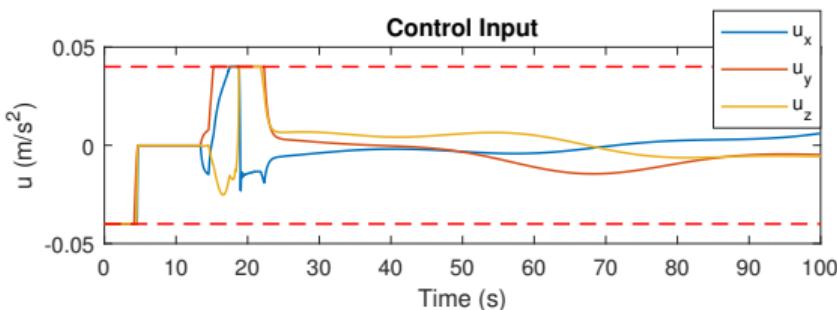


Figure: The control input using  $H(x)$  as in Method 1

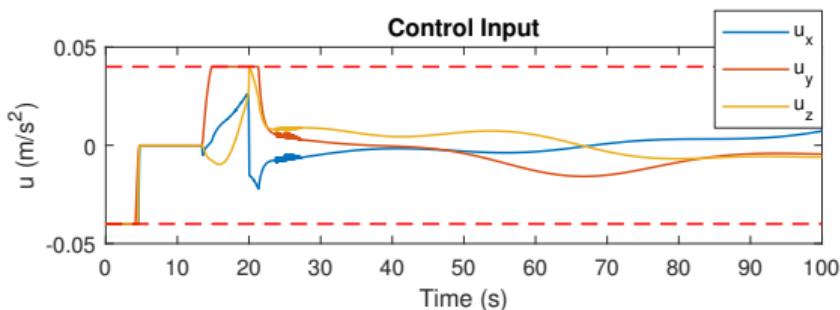


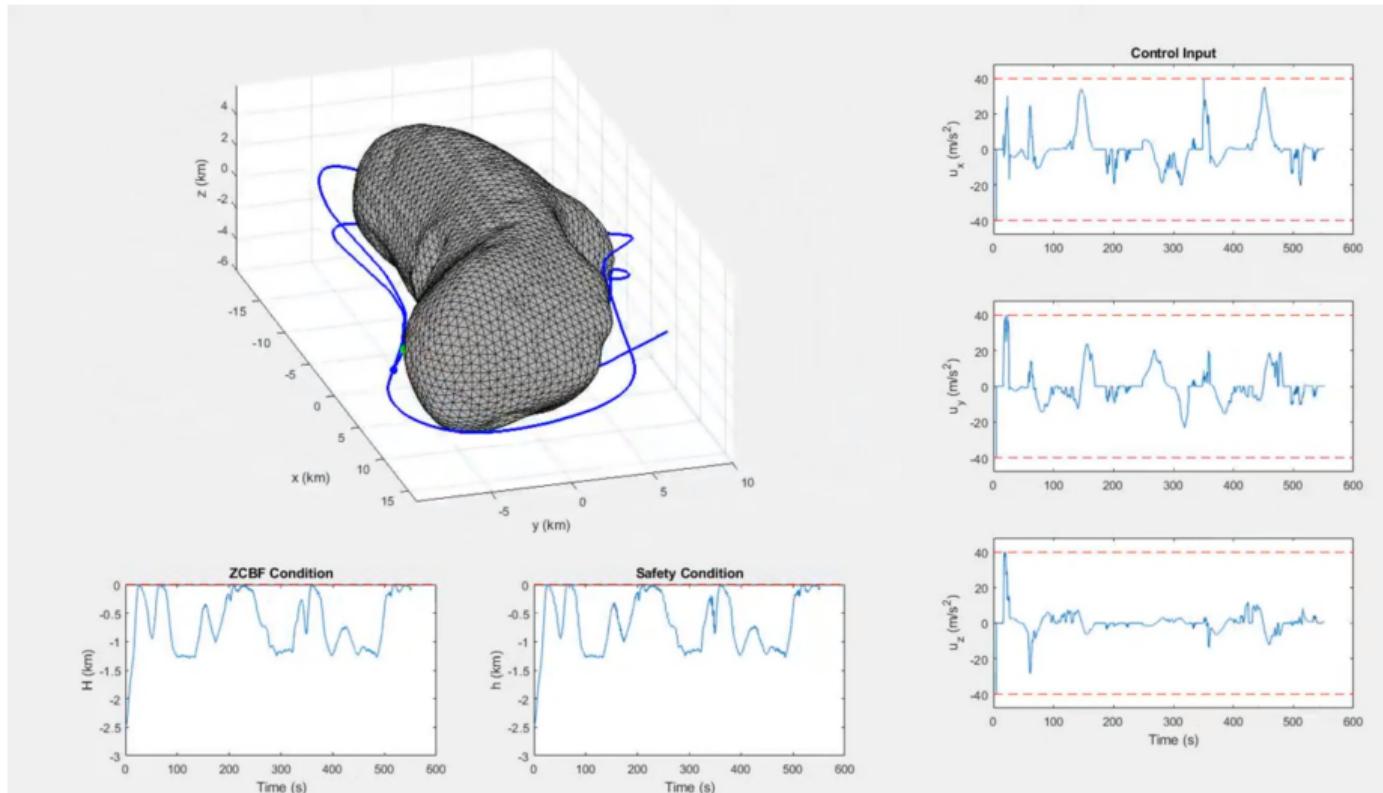
Figure: The control input using  $H'(x)$  as in Method 2

- Dynamics

$$\dot{x} = \begin{bmatrix} \dot{r} \\ \dot{v} \end{bmatrix} = \begin{bmatrix} v \\ f_\mu(r) + u \end{bmatrix}$$

- Simulated using only  $H'$  to reduce computations
- $a_{max} < u_{max}$  ( $a_{max} \approx \frac{1}{2}u_{max}$  in this simulation)
- Collection of CBFs  $\{H'_i(x)\}_{i=1}^{7790}$  for point cloud model  $\{r_{s,i}\}_{i=1}^{7790}$
- $r_p = r_p(x)$  moving target (for Lyapunov function)

# Application 2 Results



<https://youtu.be/JKj3PUrYnEg>



- Presented two explicit methods for constructing CBFs with input constraints
- Feasibility of  $\dot{H} \leq \alpha(-H)$  under input constraints is guaranteed within the zero sublevel sets of both CBFs
- Expanded utility of CBFs as an online control methodology
- Current/future work
  - Input constraints + disturbances + sampled-data dynamics
  - Fuel-optimality/planning
  - Multi-agent space systems

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