

Safety of control systems under uncertainty and time delays

Part 2

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Control Barrier Functions and Input-to-State Safety With Application to Automated Vehicles

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IEEE TRANSACTIONS ON INTELLIGENT VEHICLES,

Integrating Safety With Performance in Connected Automated Truck Control: Experimental Validation

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Johaán Chacko Mathew, A. Harvey Bell, and Gábor Orosz⁴, Senior Member, IEEE

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RESEARCH ARTICLE

Safety-Critical Control With Input Delay in Dynamic Environment

Tamas G. Molnar¹, Member, IEEE, Adam K. Kiss², Aaron D. Ames³, Fellow, IEEE,
and Gábor Orosz⁴, Senior Member, IEEE

Control barrier functionals: Safety-critical control for time delay systems

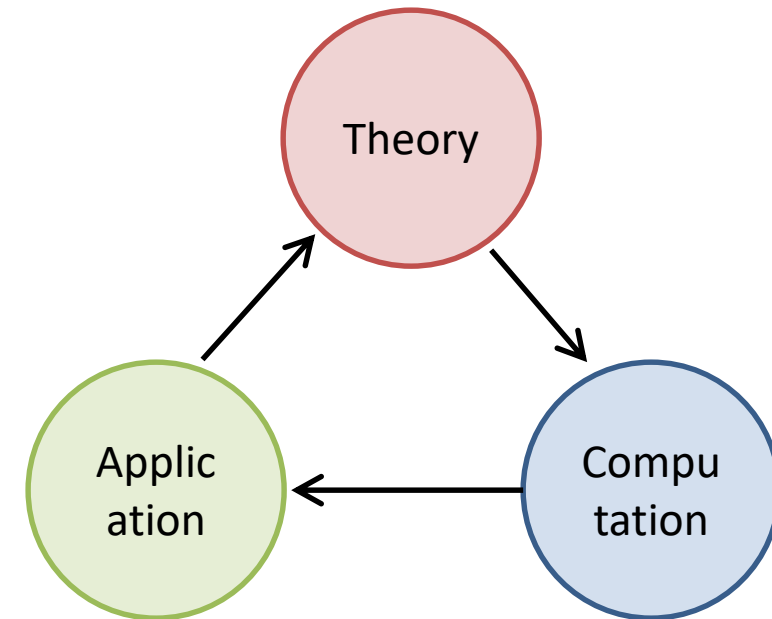
Adam K. Kiss¹ | Tamas G. Molnar² | Aaron D. Ames² | Gabor Orosz³

Motivation

5 advices from **Richard Murray**
(CSS Award Speech, 2017)



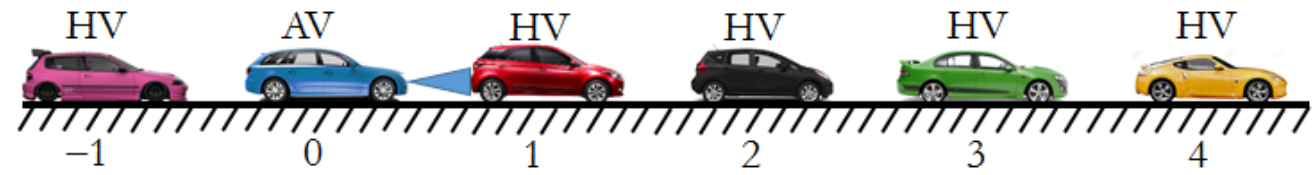
- be increasingly multilingual
- spread the gospel
- embrace diversity
- master the TCA cycle
- get out the box



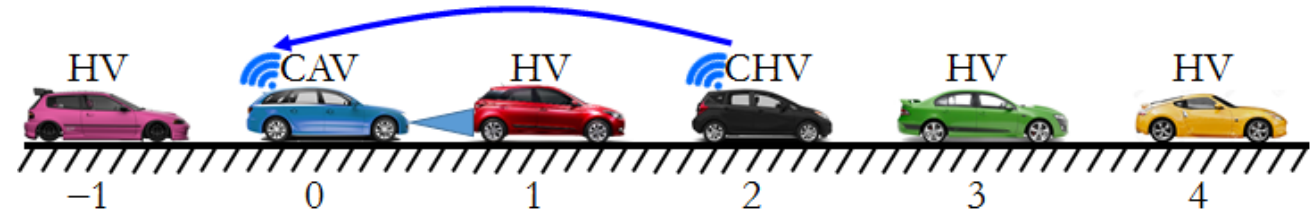
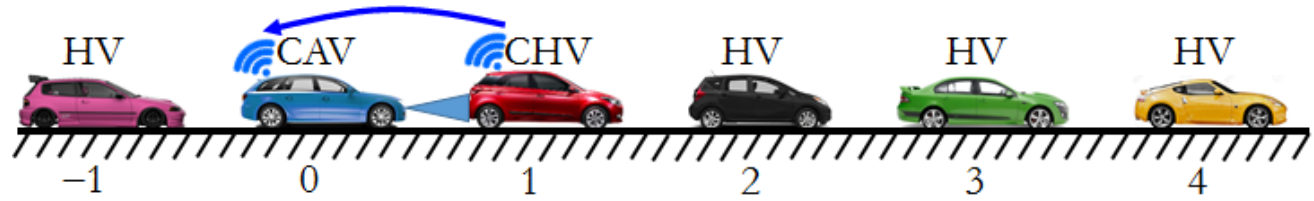
Motivation



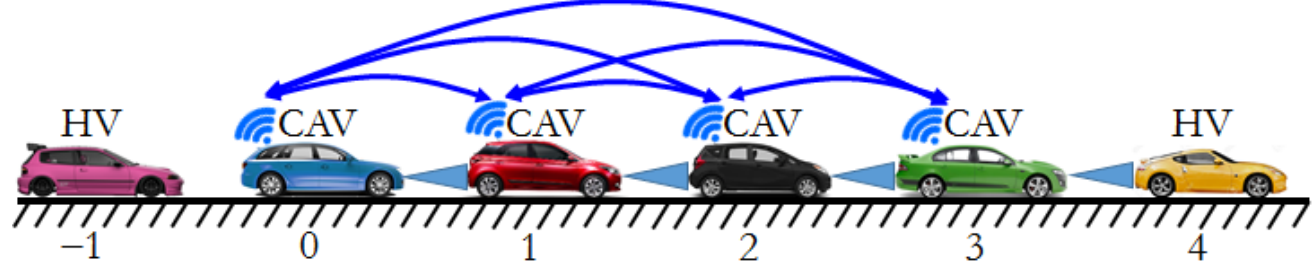
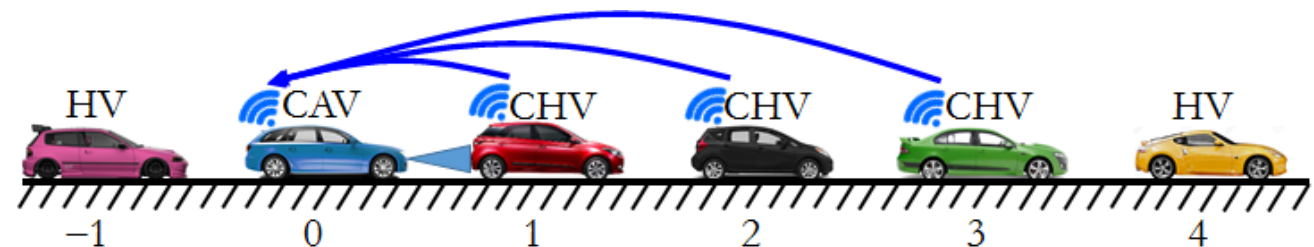
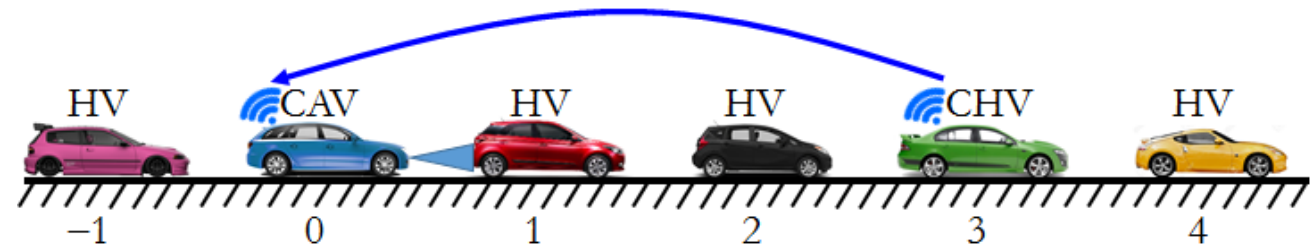
Adaptive
Cruise
Control
(ACC)



Connected
Cruise
Control
(CCC)



Cooperative
Adaptive
Cruise
Control
(CACC)



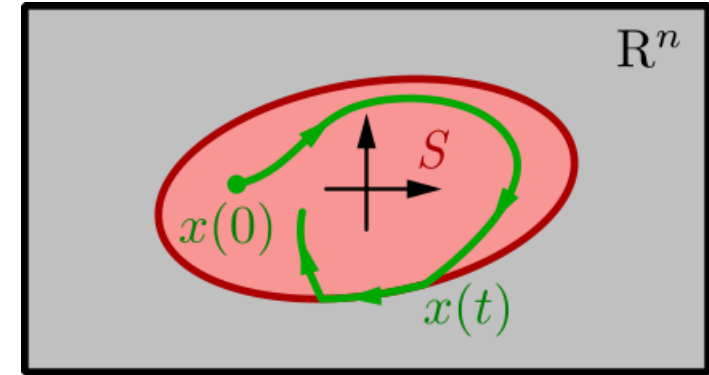
Mixed Traffic

Control Barrier Function 101

Control system:

$$\dot{x} = f(x) + g(x)u \quad \begin{array}{l} x \in \mathbb{R}^n \\ u \in \mathbb{R}^m \end{array}$$

Safe set: $S \subset \mathbb{R}^n$

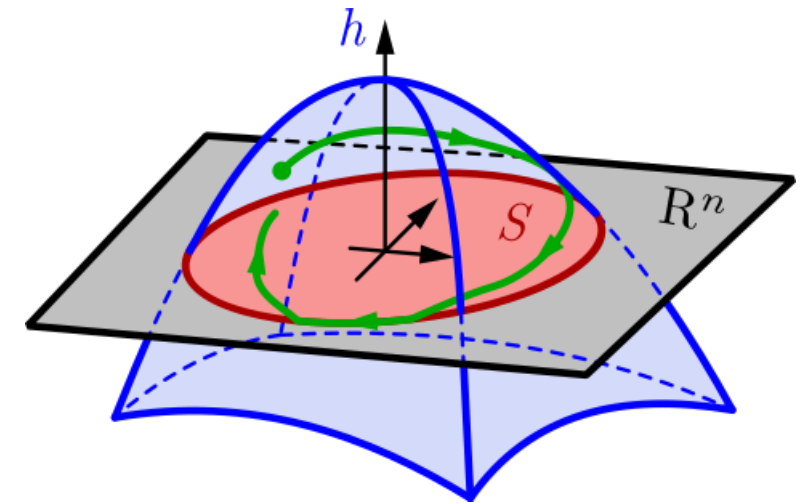


Safety as forward invariance:

$$x(0) \in S \Rightarrow x(t) \in S, \forall t \geq 0$$

Safe set constructed super-level set:

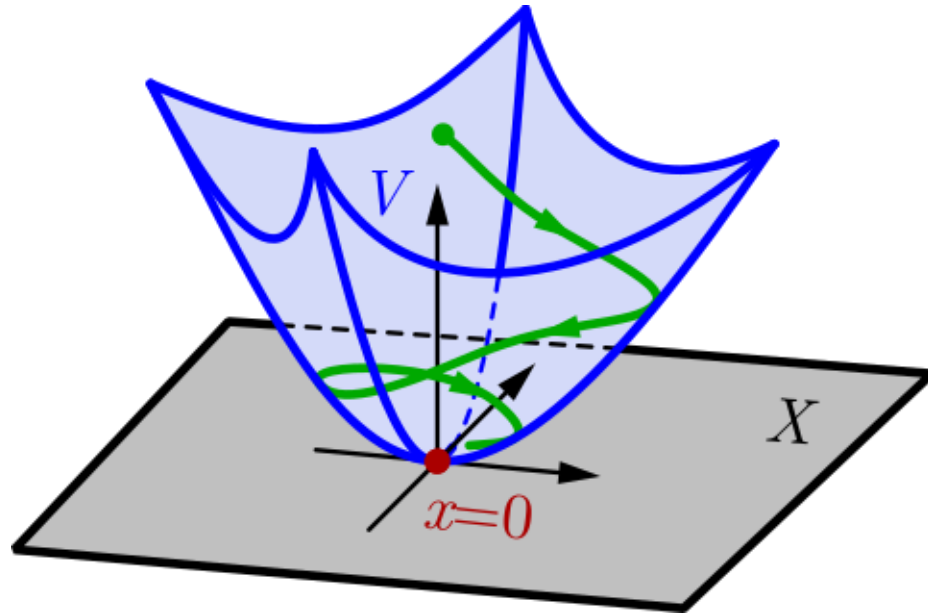
$$S = \{x \in X : h(x) \geq 0\}$$



Goal: synthesize u so that $h(x(t))$ stays nonnegative

Control Barrier Function 101

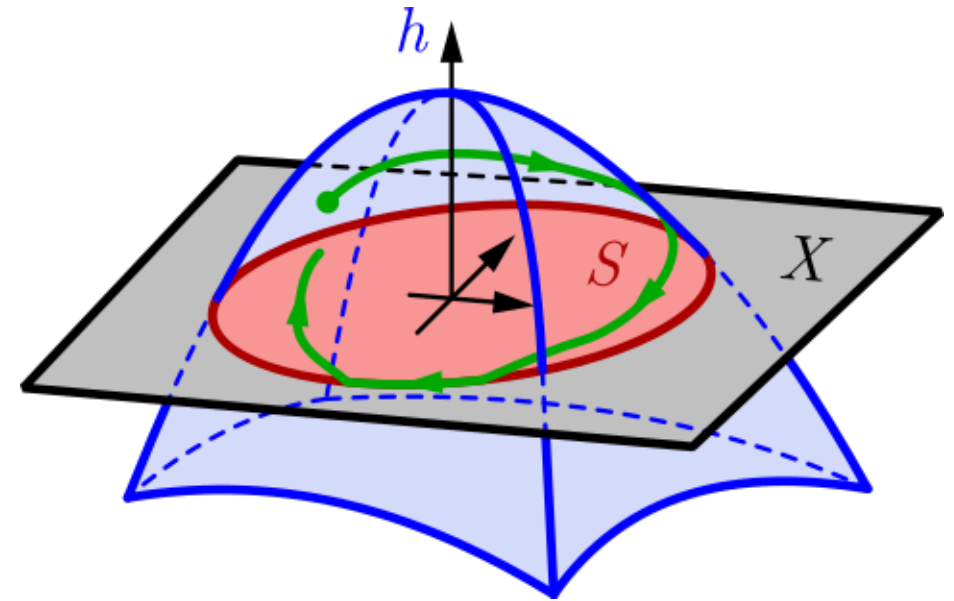
Control Lyapunov functions (CLFs)



$$k_1 \|x\|^c \leq V(x) \leq k_2 \|x\|^c$$
$$\dot{V}(x, u) \leq -\lambda V(x)$$

Approach equilibrium exponentially
with some minimum rate

Control barrier functions (CBFs)



$$S = \{x \in X : h(x) \geq 0\}$$
$$\dot{h}(x, u) \geq -\alpha h(x)$$

Approach safe set boundary
with some maximum rate

Control Barrier Function 101

Control how fast the safe set boundary is approached

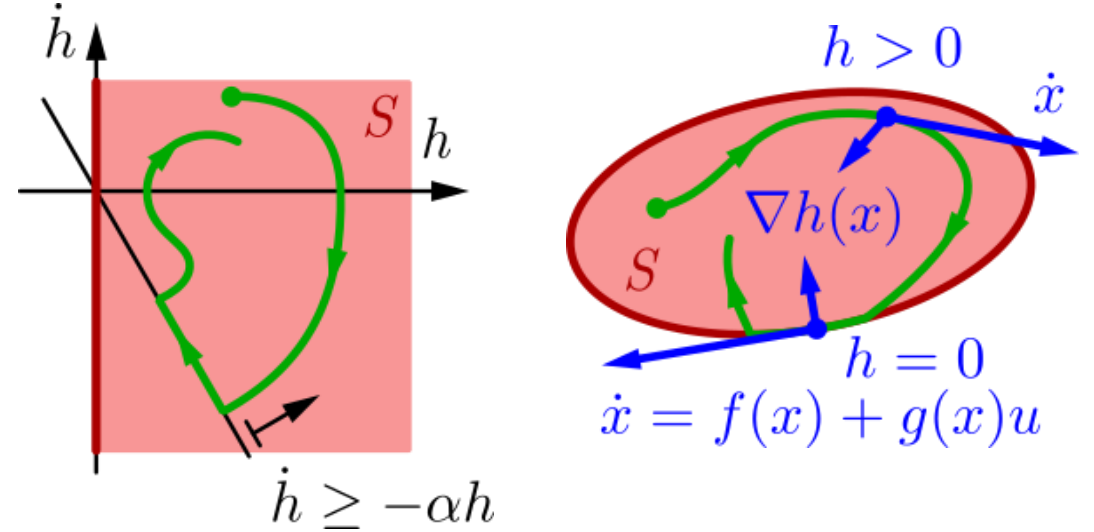
$$\dot{h}(x, u) \geq -\alpha h(x)$$

$$\alpha > 0$$

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

Theorem:

controllers $u = k(x)$
 that satisfy $\dot{h}(x, k(x)) \geq -\alpha h(x)$
 ensure safety: $x(0) \in S \Rightarrow x(t) \in S, \forall t \geq 0$

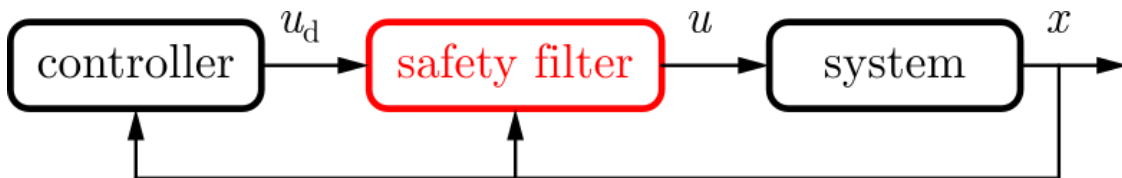


Control synthesis via optimization:

$$k_{QP}(x) = \operatorname{argmin}_{u \in U} \|u - k_n(x)\|^2$$

$$\Rightarrow$$

$$\text{s.t. } \dot{h}(x, u) \geq -\alpha h(x)$$



Analytical solution (single input):

$$k_{QP}(x) = \max \left\{ k_n(x), -\frac{\nabla h(x)f(x) + \alpha h(x)}{\nabla h(x)g(x)} \right\}$$

$$\nabla h(x)g(x) > 0$$

Control Barrier Function 101

Control how fast the safe set boundary is approached

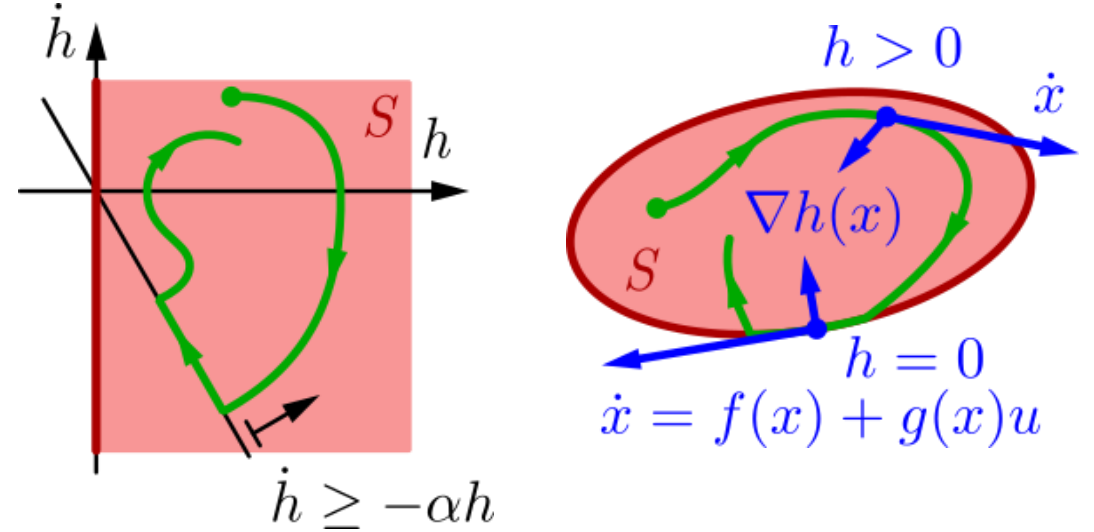
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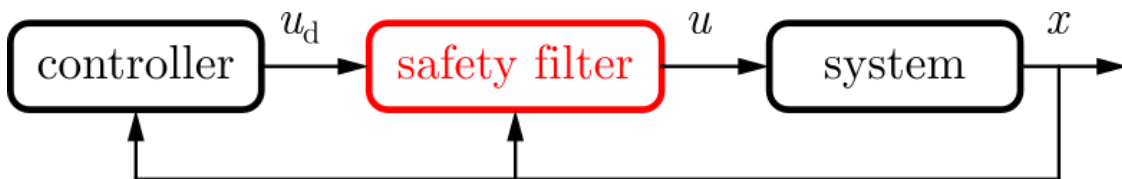


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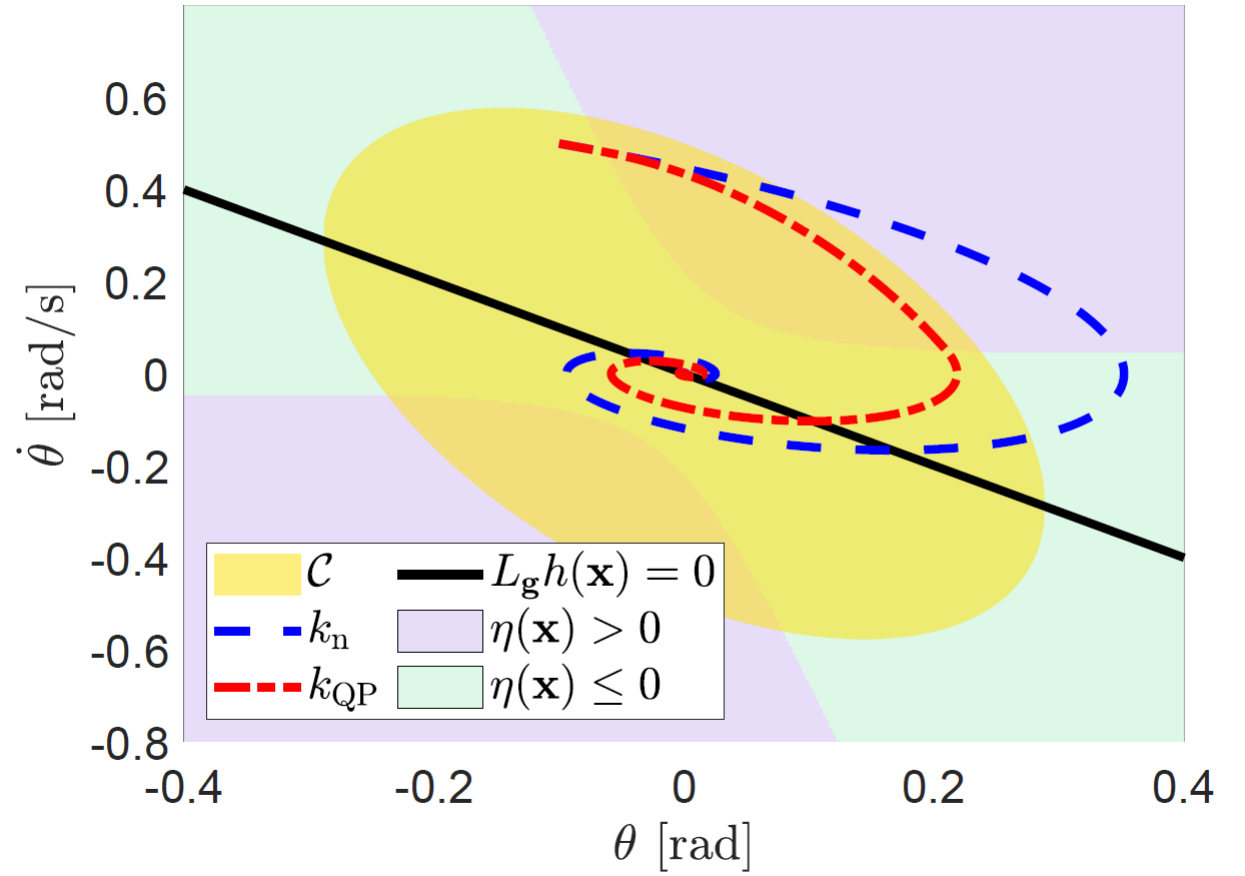
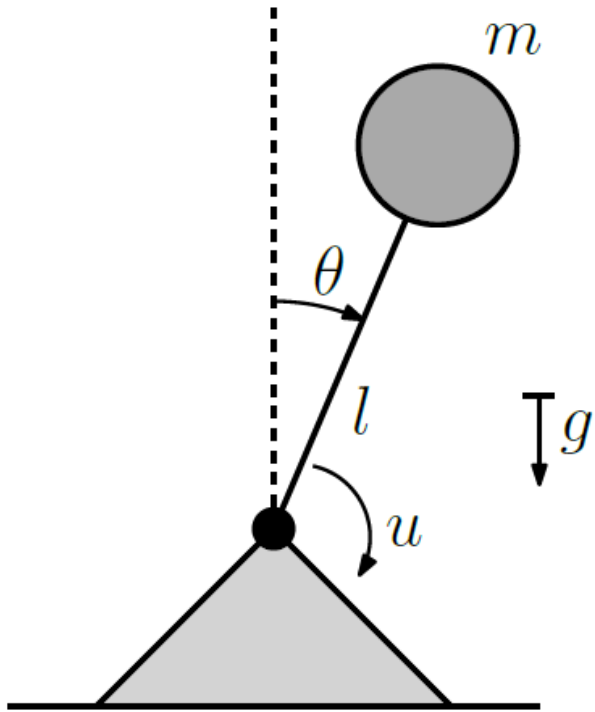


Analytical solution (single input):

$$k_{QP}(x) = \min \left\{ k_n(x), -\frac{\nabla h(x)f(x) + \alpha h(x)}{\nabla h(x)g(x)} \right\}$$

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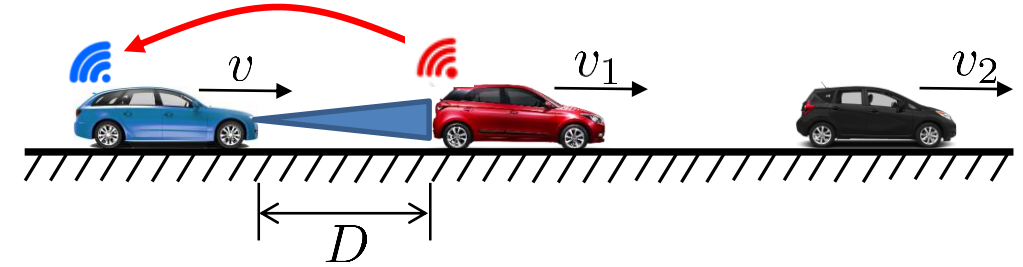
Control Barrier Function 101



CBF Applied to Vehicle Control

Dynamical model:

$$\left. \begin{aligned} \dot{D} &= v_L - v \\ \dot{v} &= u \\ \dot{v}_L &= a_L(t) \end{aligned} \right\} \dot{x} = f(t, x) + g(x)u$$



Connected Cruise Control

Safety measure:

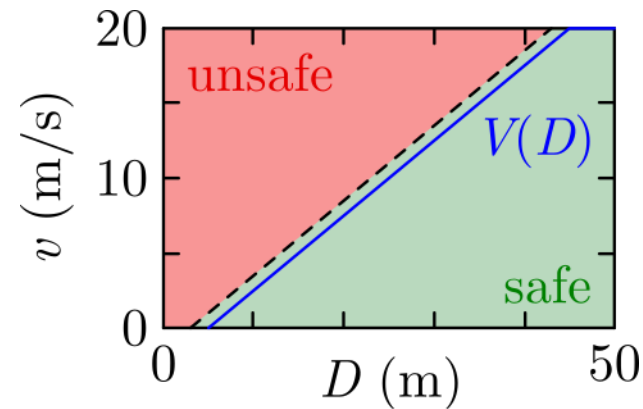
$$h(x) = D - D_{sf} - Tv$$

Safety critical control:

$$\dot{h}(t, x, u) \geq -\alpha h(x)$$

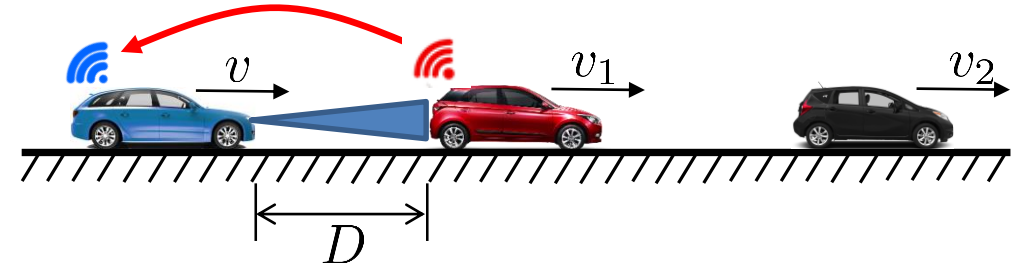
$$u \leq \alpha \left(\frac{1}{T} (D - D_{sf}) - v \right) + \frac{1}{T} (v_L - v)$$

$$u = \alpha \left(\underbrace{\frac{1}{T} (D - D_{st})}_{V(D)} - v \right) + \frac{1}{T} (v_L - v)$$

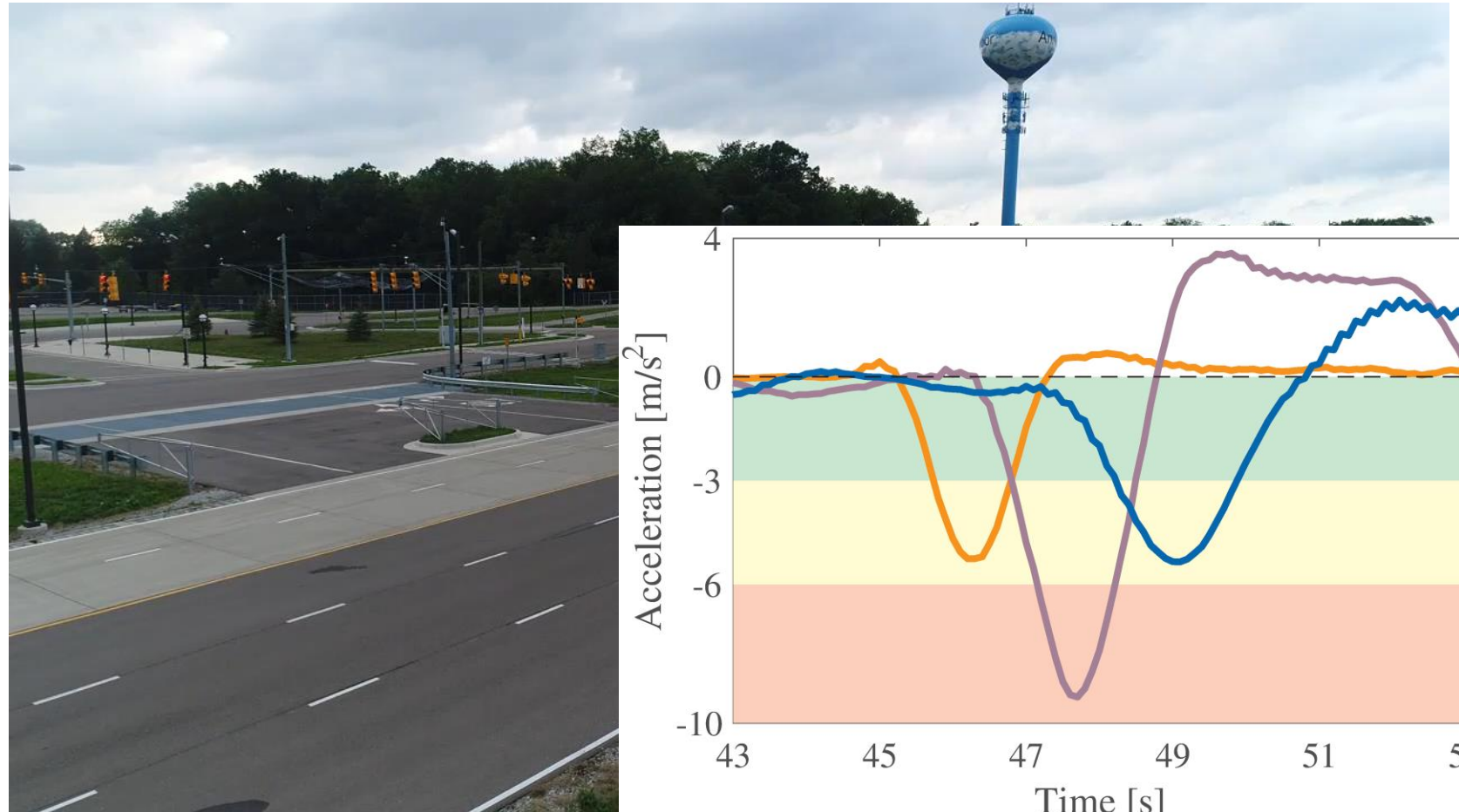
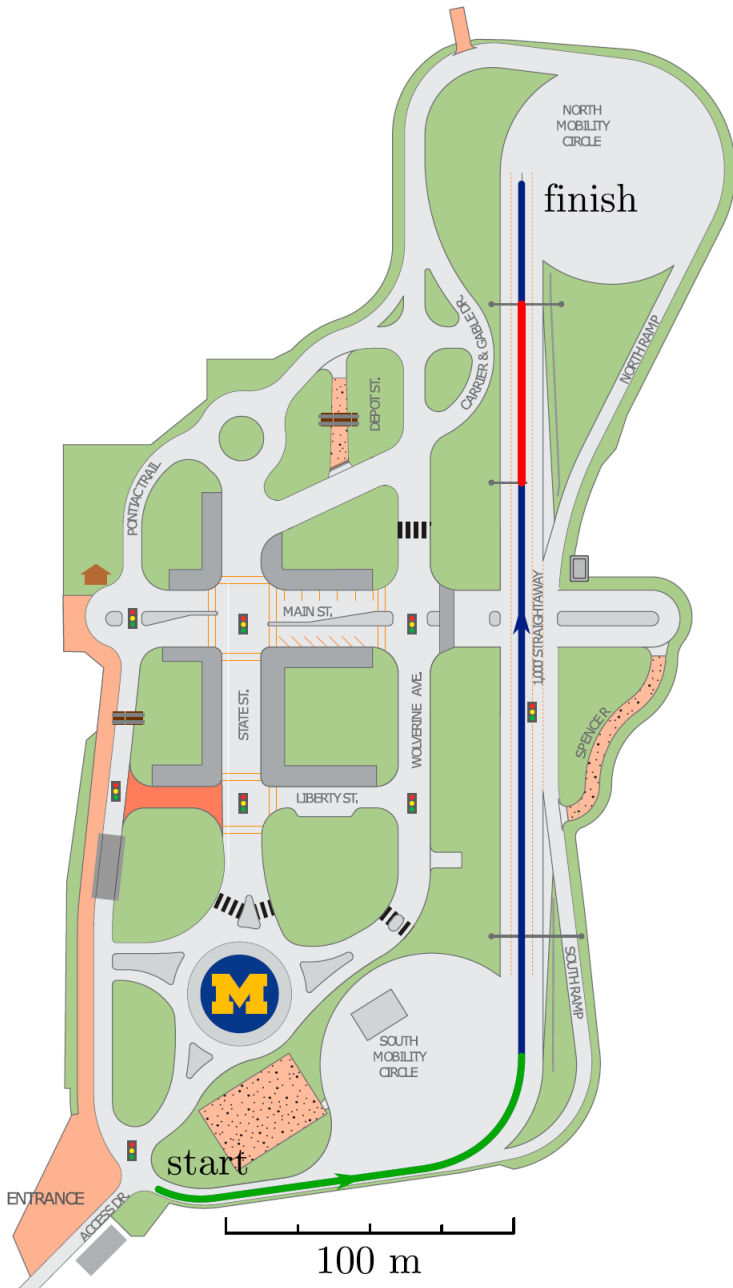


Connected Cruise Control Experiments

- Acceleration
- Constant Speed
- Braking

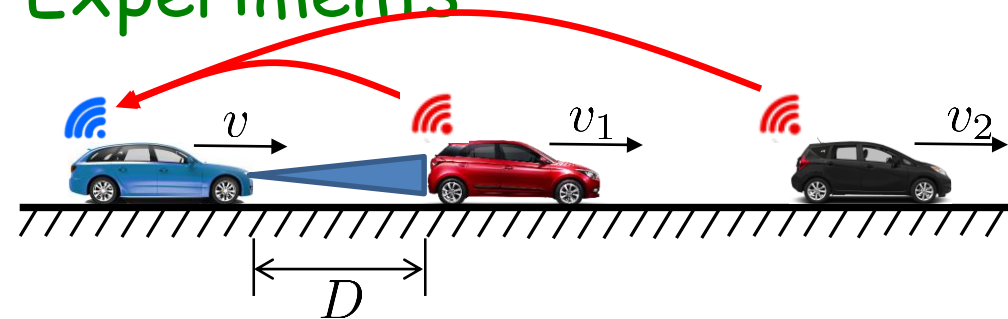


Connected Cruise Control

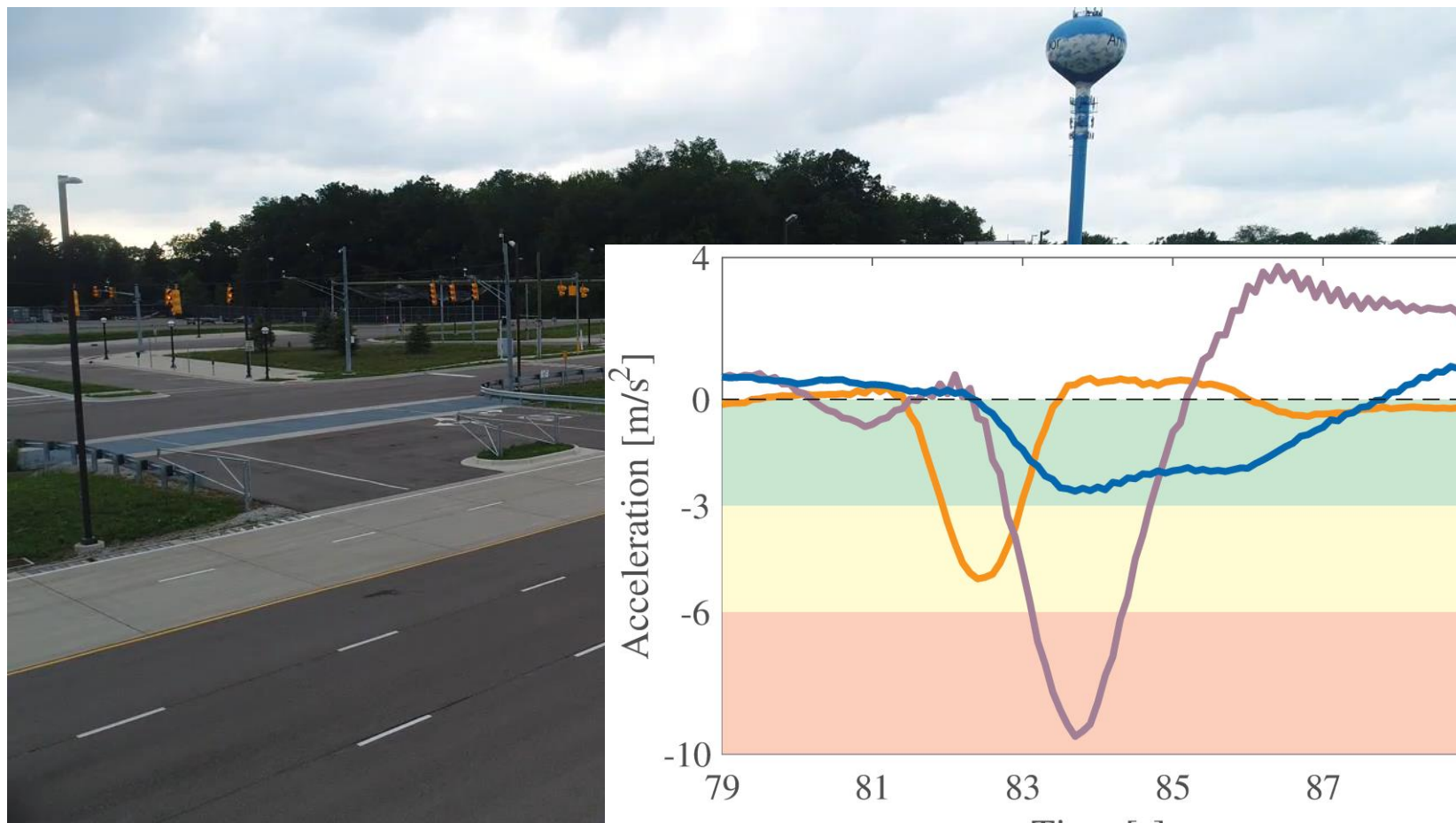
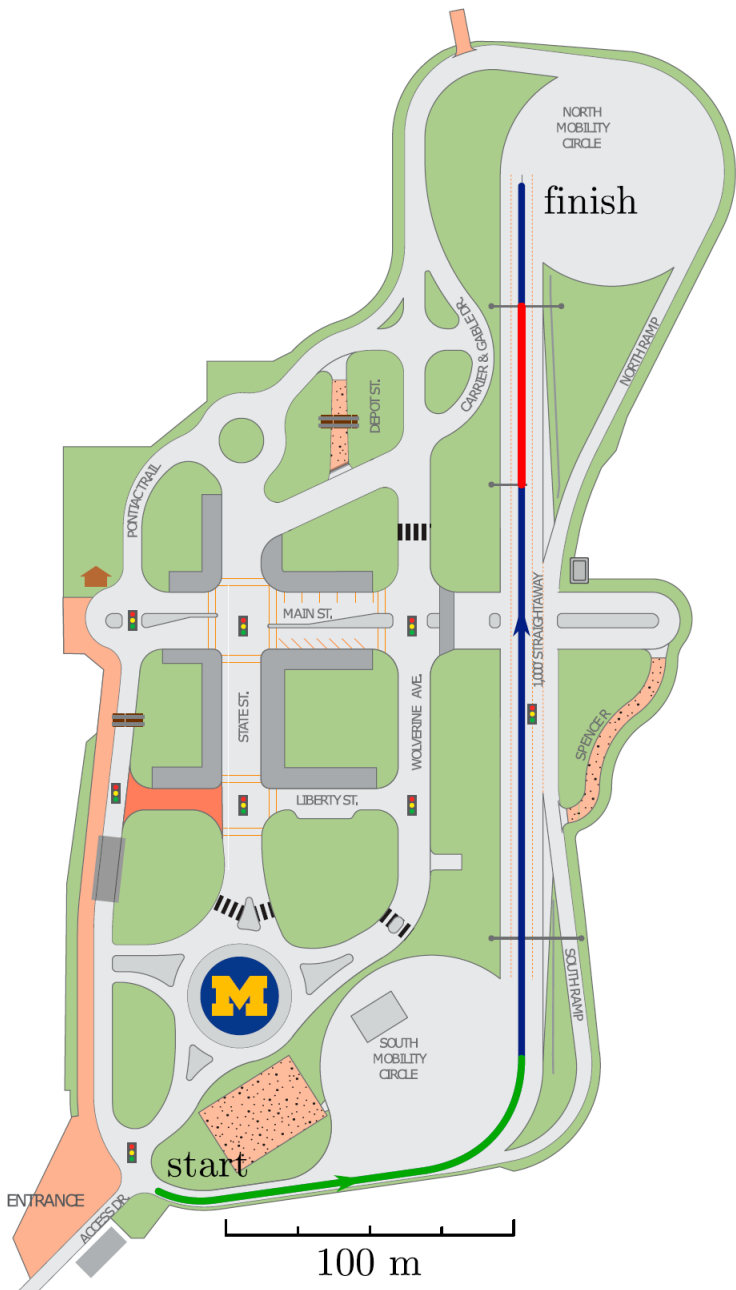


Connected Cruise Control Experiments

- Acceleration
- Constant Speed
- Braking



Connected Cruise Control



CBF Applied to Truck Control

Dynamical model:

$$\left. \begin{aligned} \dot{D} &= v_L - v \\ \dot{v} &= u \\ \dot{v}_L &= a_L(t) \end{aligned} \right\} \dot{x} = f(t, x) + g(x)u$$

Safety measure:

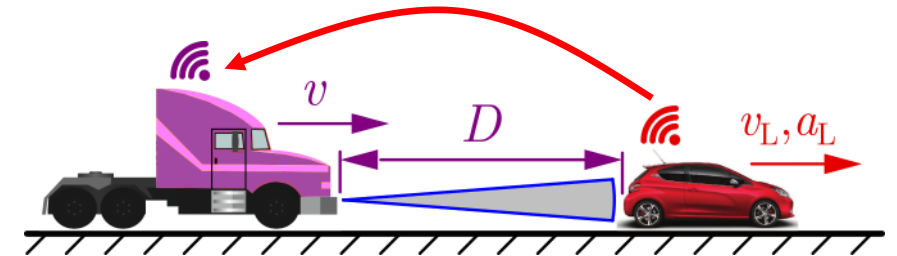
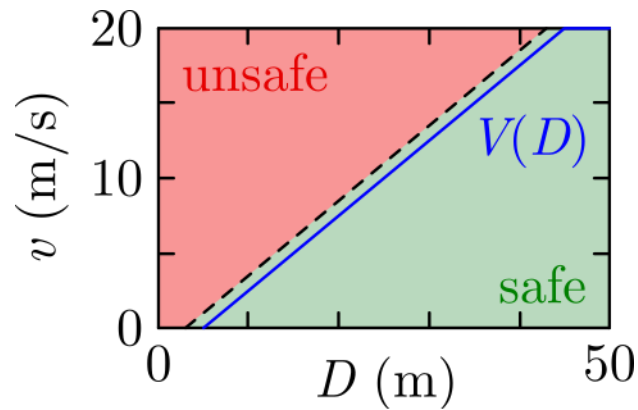
$$h(x) = D - D_{sf} - Tv$$

Safety critical control:

$$\dot{h}(t, x, u) \geq -\alpha h(x)$$

$$u \leq \alpha \left(\frac{1}{T}(D - D_{sf}) - v \right) + \frac{1}{T}(v_L - v)$$

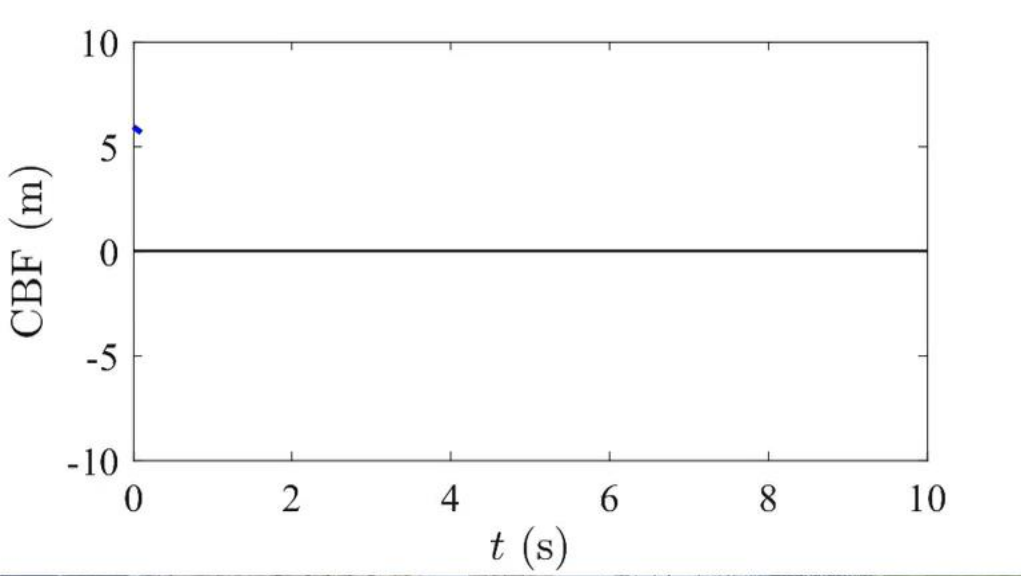
$$u = \alpha \left(\underbrace{\frac{1}{T}(D - D_{st})}_{V(D)} - v \right) + \frac{1}{T}(v_L - v)$$

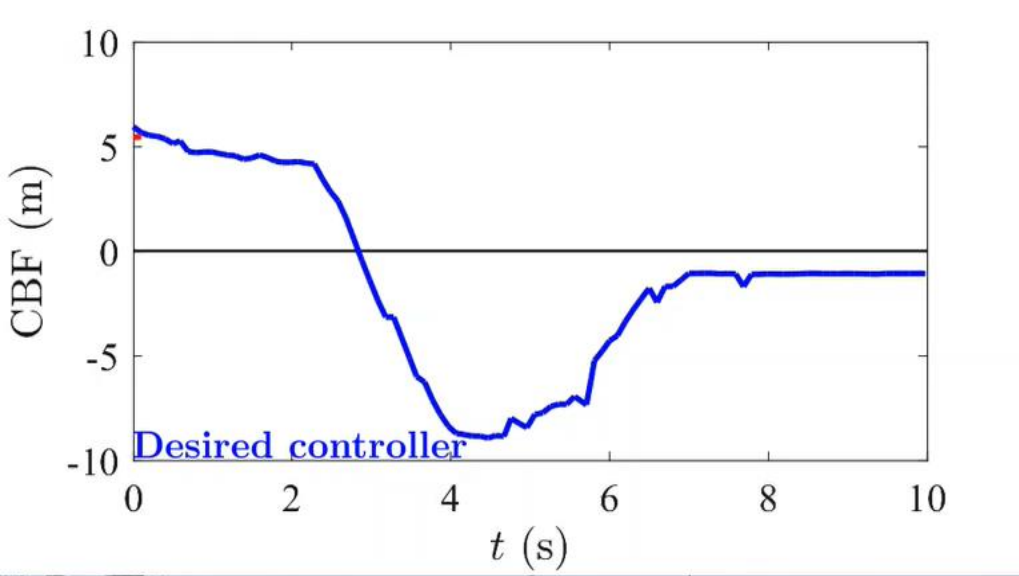


Connected Cruise Control









Delays and Disturbances Destroy Safety

Dynamical model:

$$\left. \begin{aligned} \dot{D} &= v_L - v \\ \dot{v} &= u \\ \dot{v}_L &= a_L(t) \end{aligned} \right\} \dot{x} = f(t, x) + g(x)u$$

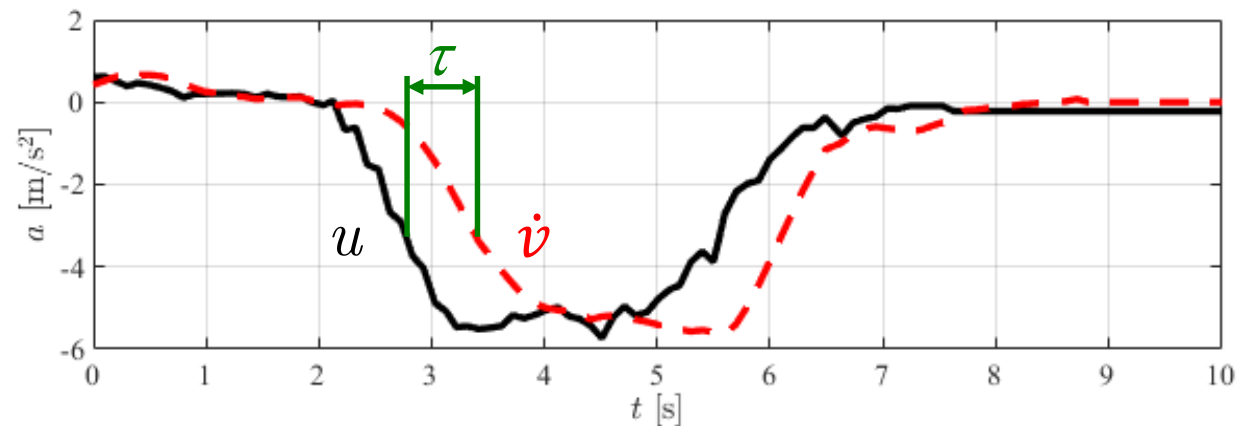
Why did the experiment fail?

Disturbances:

Truck's dynamics are uncertain
(engine, transmission, brakes, tires...)

Delay:

Truck is massive, with large response time



Delays and Disturbances Destroy Safety

Dynamical model:

$$\left. \begin{aligned} \dot{D} &= v_L - v \\ \dot{v} &= u(t - \tau) + d(t) \\ \dot{v}_L &= a_L(t) \end{aligned} \right\} \dot{x} = f(t, x) + g(x)(u(t - \tau) + d(t))$$

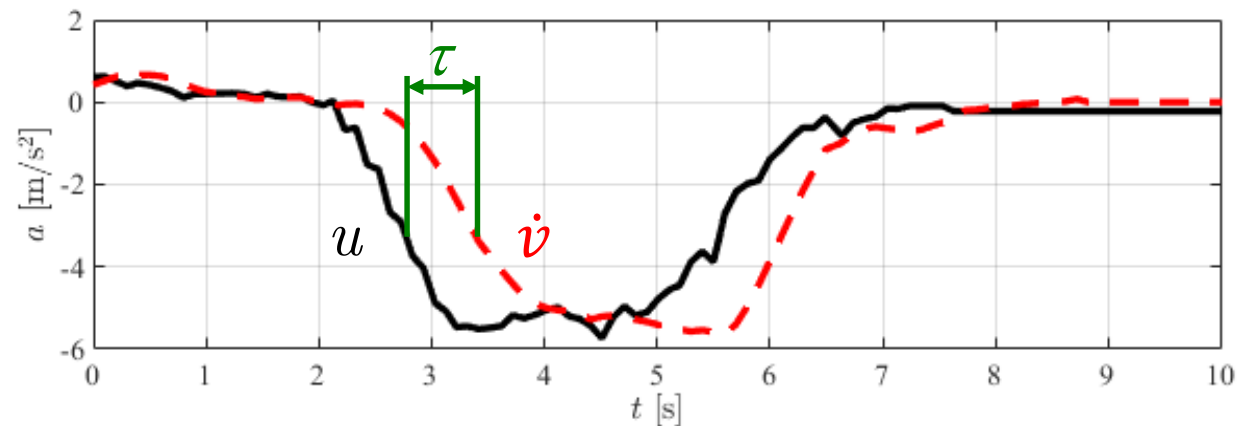
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Disturbances:

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Delay:

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Content

Disturbances/Environment

Input-to-state safe control barrier functions (ISSf-CBF)

Tunable Input-to-state safe control barrier functions (TISSf-CBF)

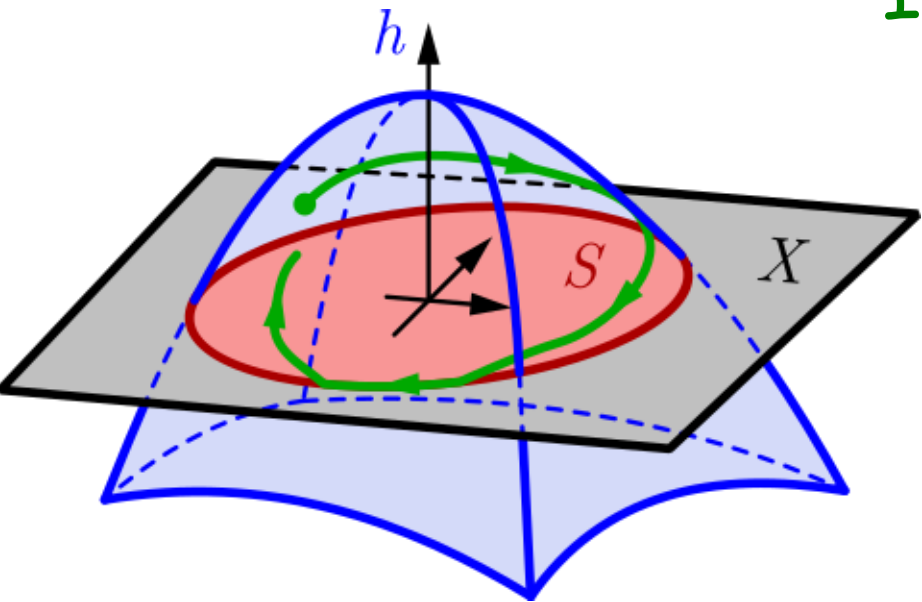
Time delays

Input delay \rightarrow Predictors

Application of safety filters in the real world

Moving from test track to the real world

Input-to-State Safety



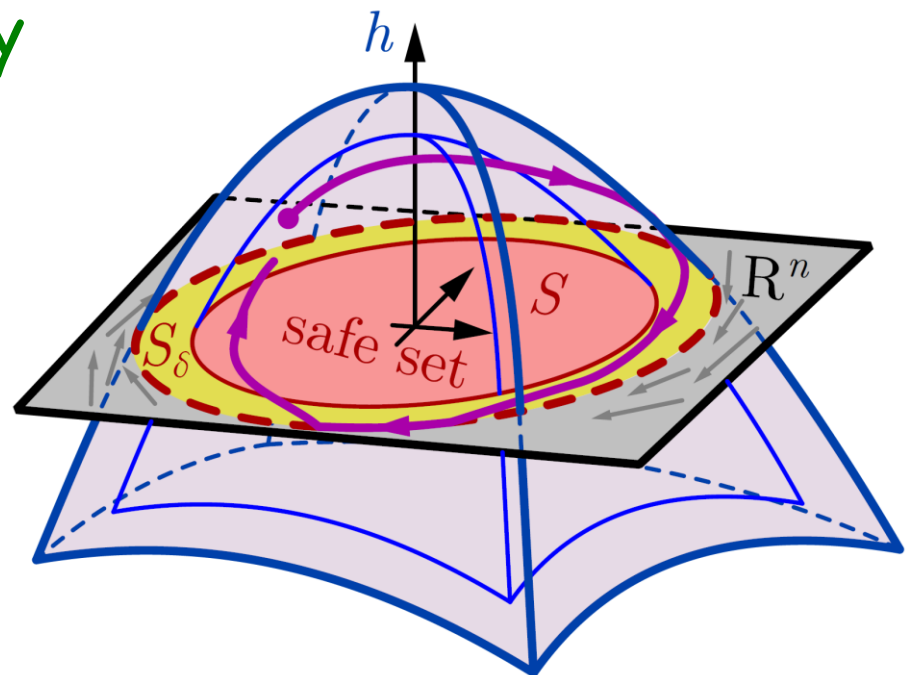
$$S = \{x \in X : h(x) \geq 0\}$$

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

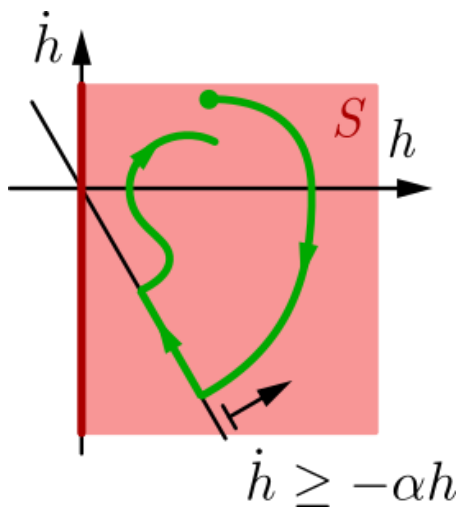
$$\|d\|_\infty \leq \delta$$

$$\dot{h}(x, u) \geq -\alpha h(x) + ???$$

$$\alpha > 0$$



$$S_\delta = \{x \in X : \underbrace{h(x) + \gamma(\delta)}_{h_\delta(x)} \geq 0\}$$

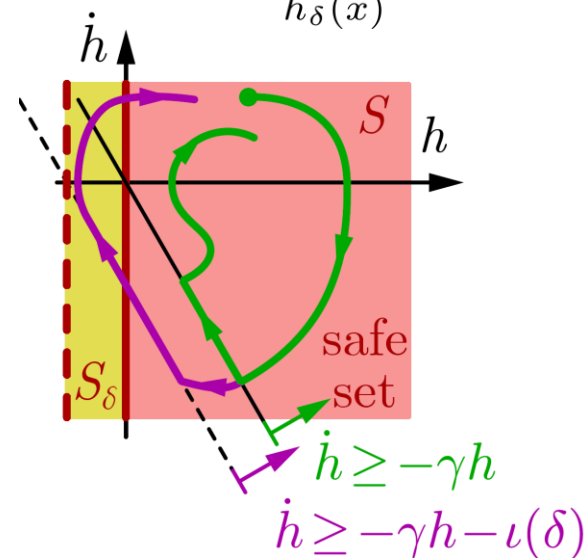


$$\dot{h} \geq -\alpha h$$

Theorem:

$$\dot{h}(x, u) \geq -\alpha h(x) + \sigma \|\nabla h(x)g(x)\|^2$$

+ formulae linking $\sigma, \alpha, \iota, \gamma$



$$\dot{h} \geq -\gamma h$$

$$\dot{h} \geq -\gamma h - \iota(\delta)$$

Input-to-State Safety

Theorem:

$$\dot{h}(x, u) \geq -\alpha h(x) + \sigma \|\nabla h(x)g(x)\|^2$$

If $k(x)$ safe without disturbance:

$$\bar{k}(x) = k(x) + \sigma (\nabla h(x)g(x))^\top$$

Otherwise use QP:

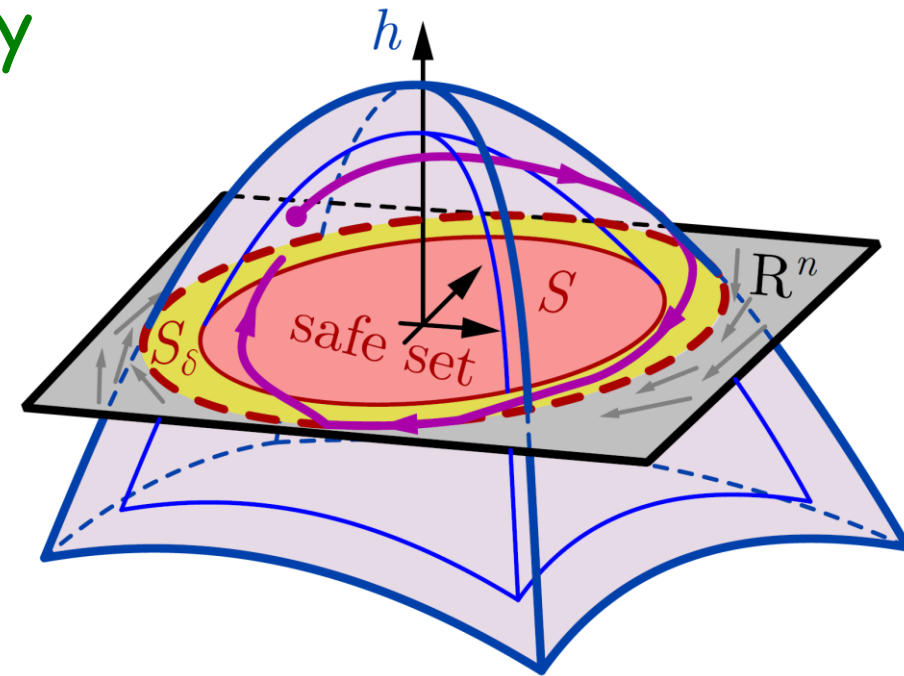
$$\bar{k}_{\text{QP}}(x) = \operatorname{argmin}_{u \in U} \|u - k_n(x)\|^2$$

s.t. $\dot{h}(x, u) \geq -\alpha h(x) + \sigma \|\nabla h(x)g(x)\|^2$

\Rightarrow **Analytical solution** (single input):

$$\bar{k}_{\text{QP}}(x) = \max \left\{ k_n(x), -\frac{\nabla h(x)f(x) + \alpha h(x)}{\nabla h(x)g(x)} + \sigma \nabla h(x)g(x) \right\}$$

$$\nabla h(x)g(x) > 0$$



$$S_\delta = \{x \in X : \underbrace{h(x) + \gamma(\delta)}_{h_\delta(x)} \geq 0\}$$

Input-to-State Safety

Theorem:

$$\dot{h}(x, u) \geq -\alpha h(x) + \sigma \|\nabla h(x)g(x)\|^2$$

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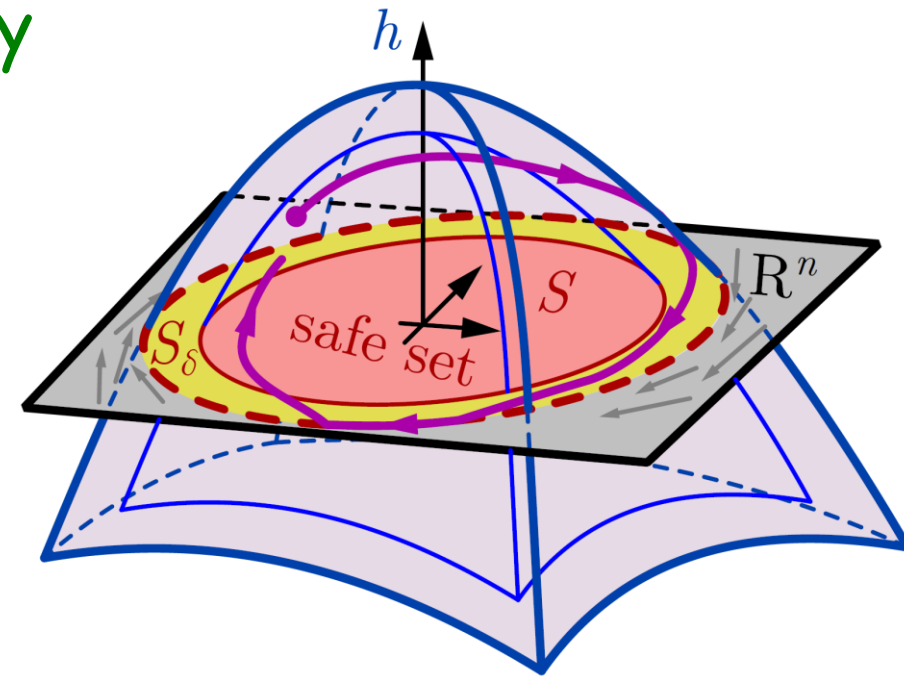
$$\bar{k}_{\text{QP}}(x) = \underset{u \in U}{\operatorname{argmin}} \|u - k_n(x)\|^2$$

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$$\nabla h(x)g(x) < 0$$



$$S_\delta = \{x \in X : \underbrace{h(x) + \gamma(\delta)}_{h_\delta(x)} \geq 0\}$$

Keeps the truck safe but it keeps very large distance all the time!

Tunable Input-to-State Safety

Key idea:

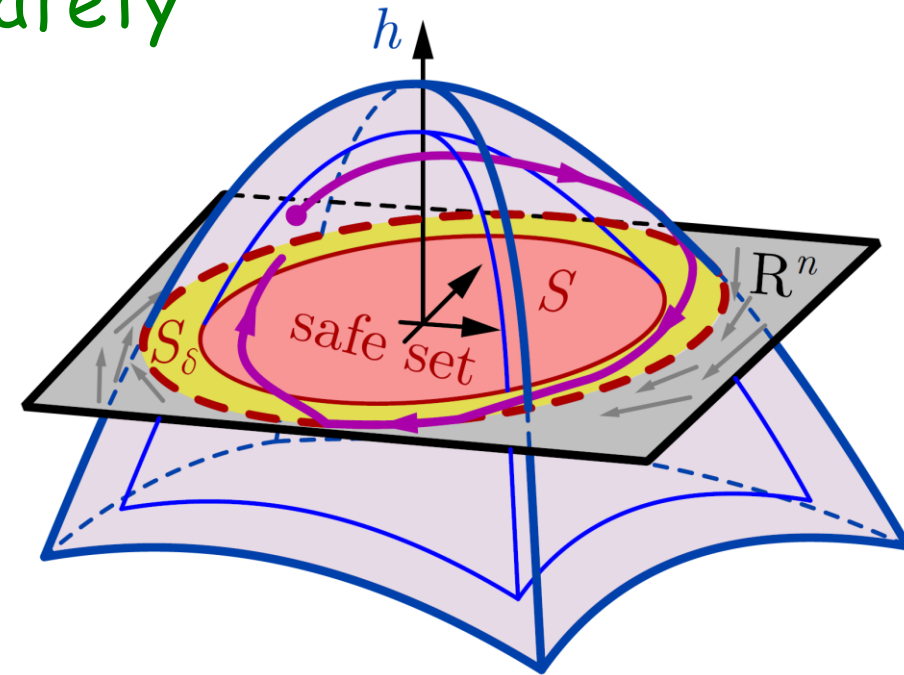
$$\sigma \rightarrow \sigma(h(x))$$

Theorem:

The above construction still works
if $\sigma(r)$ is strictly monotonically decreasing

Example:

$$\sigma(r) = \sigma_0 e^{-\lambda r}$$



$$S_\delta = \{x \in X : \underbrace{h(x) + \gamma(\delta, h(x))}_{h_\delta(x)} \geq 0\}$$

TISSf-CBF Applied to Truck Control

Dynamical model:

$$\left. \begin{aligned} \dot{D} &= v_L - v \\ \dot{v} &= u \\ \dot{v}_L &= a_L(t) \end{aligned} \right\} \dot{x} = f(t, x) + g(x)u$$

Safety measure:

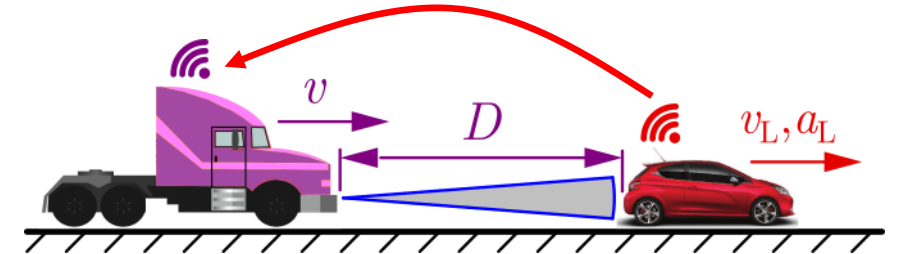
$$h(x) = D - D_{sf} - Tv$$

Safety critical control:

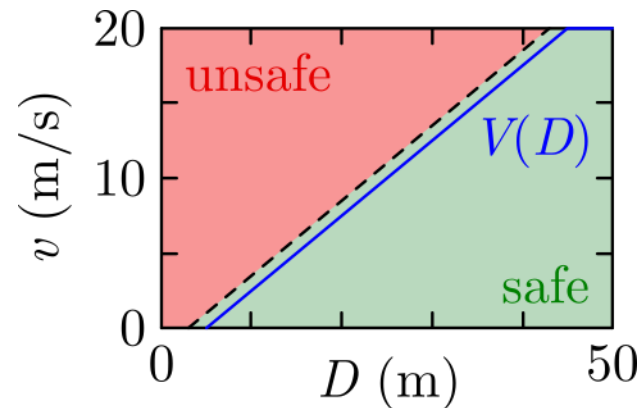
$$\dot{h}(t, x, u) \geq -\alpha h(x) + \sigma(h(x)) \|\nabla h(x)g(x)\|^2 \quad \sigma(r) = \sigma_0 e^{-\lambda r}$$

$$u \leq \alpha \left(\frac{1}{T}(D - D_{sf}) - v \right) + \frac{1}{T}(v_L - v) - T\sigma_0 e^{-\lambda(D - D_{sf} - Tv)}$$

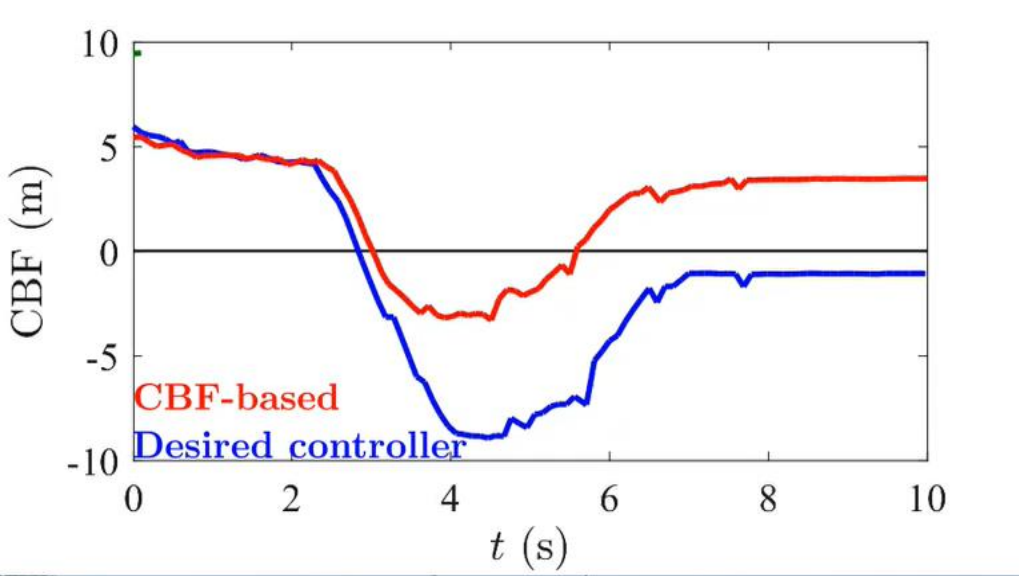
$$u = \alpha \left(\underbrace{\frac{1}{T}(D - D_{st}) - v}_{V(D)} \right) + \frac{1}{T}(v_L - v) - T\sigma_0 e^{-\lambda(D - D_{sf} - Tv)}$$



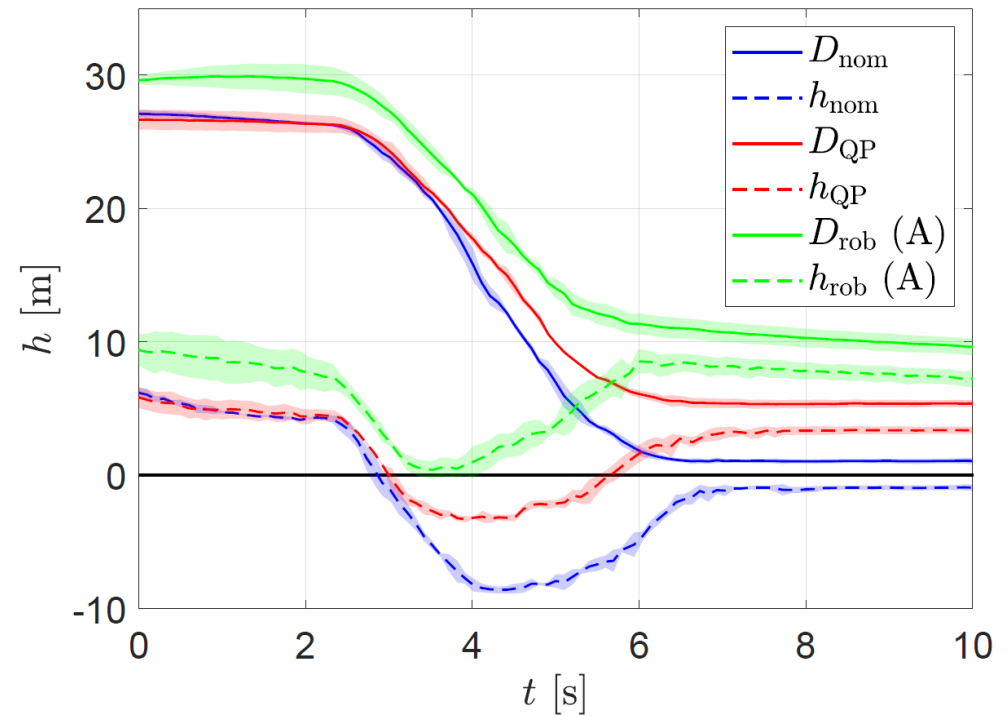
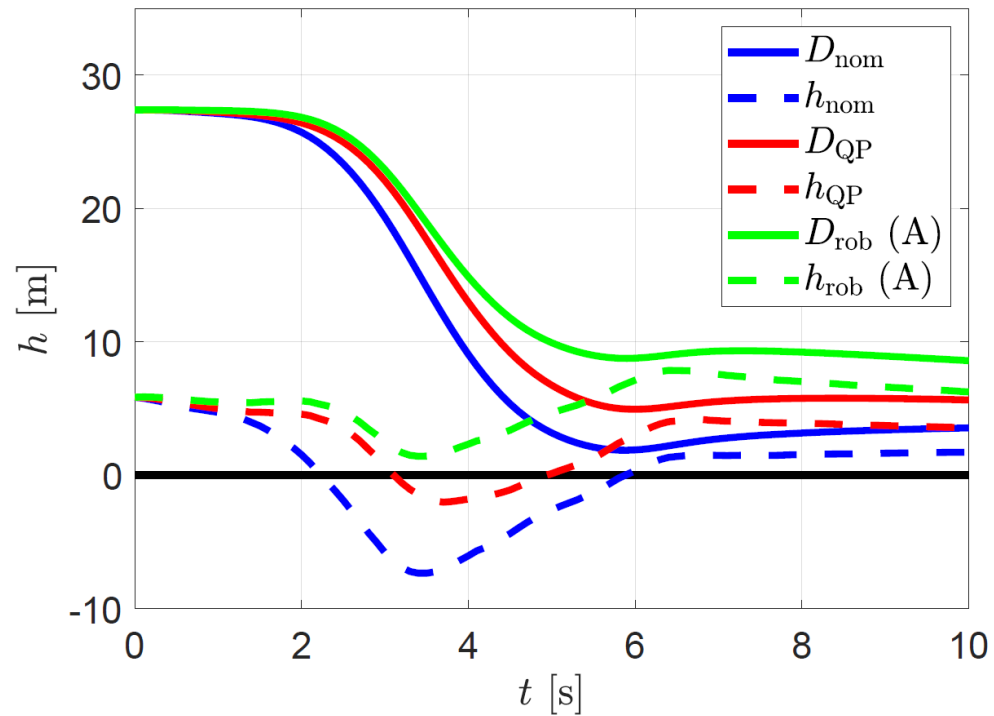
Connected Cruise Control



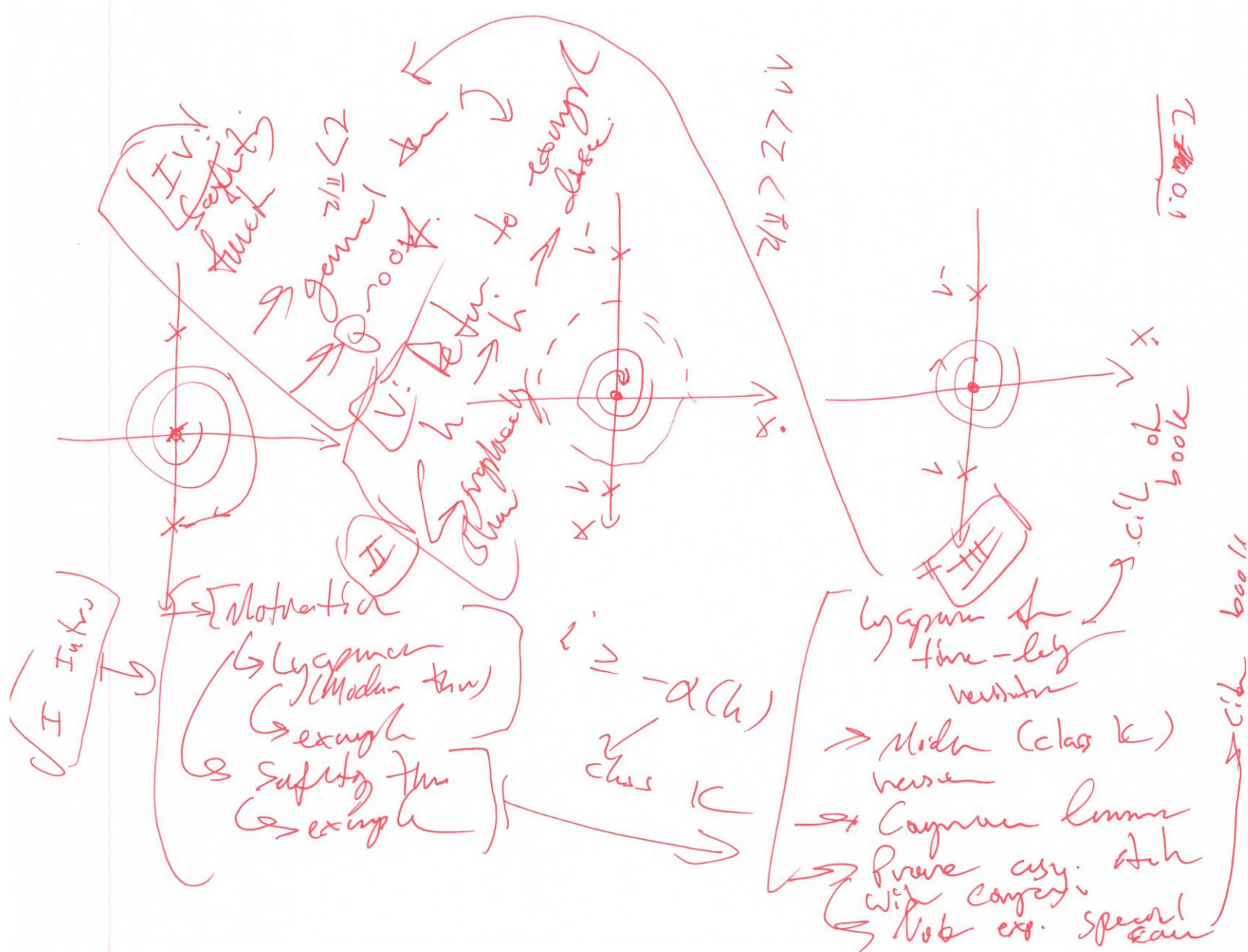




TISSf-CBF Applied to Truck Control



Safety with Time Delay

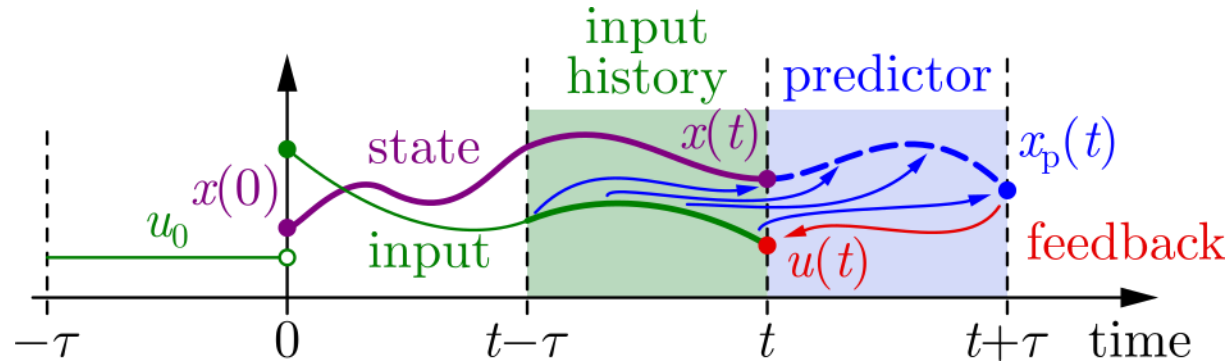
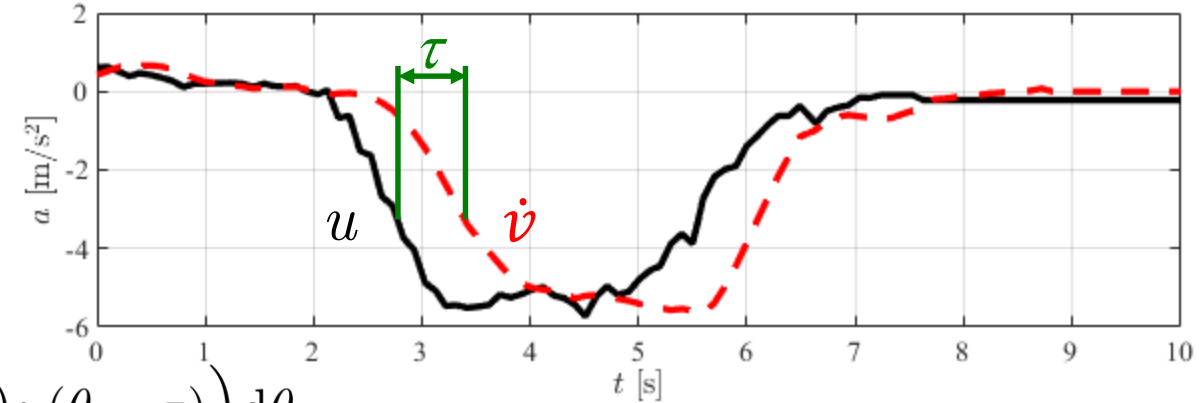


Input Delay

Predictor feedback for delay compensation:

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

$$x_p(t) = x(t + \tau) = x(t) + \int_t^{t+\tau} \left(f(x(\theta)) + g(x(\theta))u(\theta - \tau) \right) d\theta$$



Theorem:

controllers

$$u = k(x_p)$$

that satisfy

$$\dot{h}(x_p, k(x_p)) \geq -\alpha h(x_p)$$

ensure safety:

$$x(\theta) \in S, \forall \theta \in [0, \tau] \Rightarrow x(t) \in S, \forall t \geq 0$$

Predictors Recover Safety Guarantees

Dynamical model:

$$\left. \begin{aligned} \dot{D} &= v_L - v \\ \dot{v} &= u(t - \tau) \\ \dot{v}_L &= a_L(t) \end{aligned} \right\} \dot{x} = f(t, x) + g(x)u(t - \tau)$$

Safety measure:

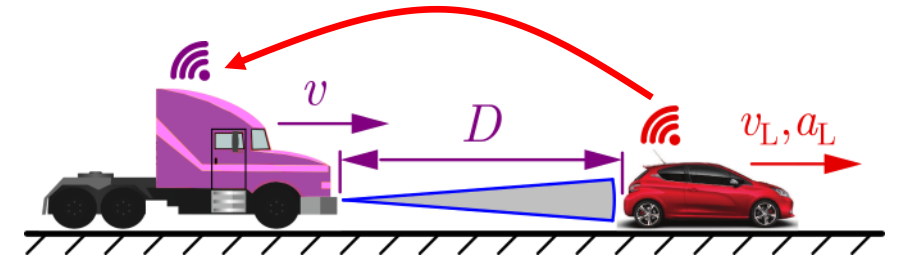
$$h(x) = D - D_{sf} - Tv$$

Safety critical control:

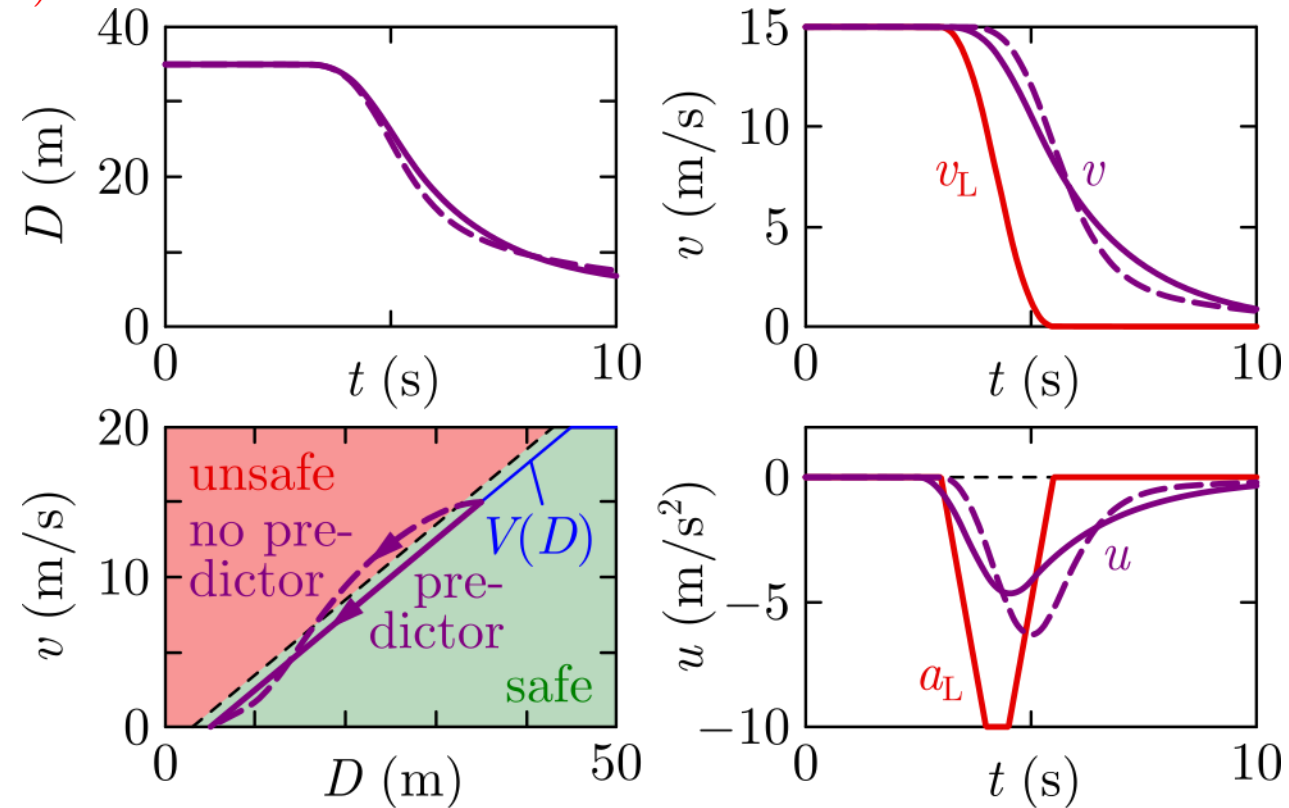
$$\dot{h}(t_p, x_p, u) \geq -\alpha h(x_p)$$

$$u = \alpha \left(\underbrace{\frac{1}{T}(D_p - D_{st}) - v_p}_{V(D_p)} + \frac{1}{T}(v_{L,p} - v_p) \right)$$

↑
intent of preceding vehicle



Connected Cruise Control



Prediction Errors Contribute to Disturbances

Dynamical model:

$$\left. \begin{aligned} \dot{D} &= v_L - v \\ \dot{v} &= u(t - \tau) + d(t) \\ \dot{v}_L &= a_L(t) \end{aligned} \right\} \dot{x} = f(t, x) + g(x)(u(t - \tau) + d(t))$$

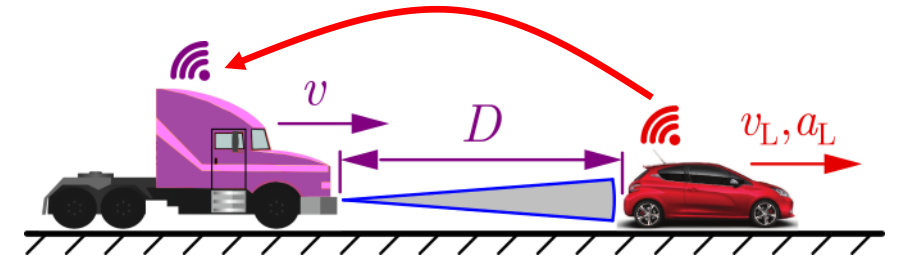
Safety measure:

$$h(x) = D - D_{sf} - Tv$$

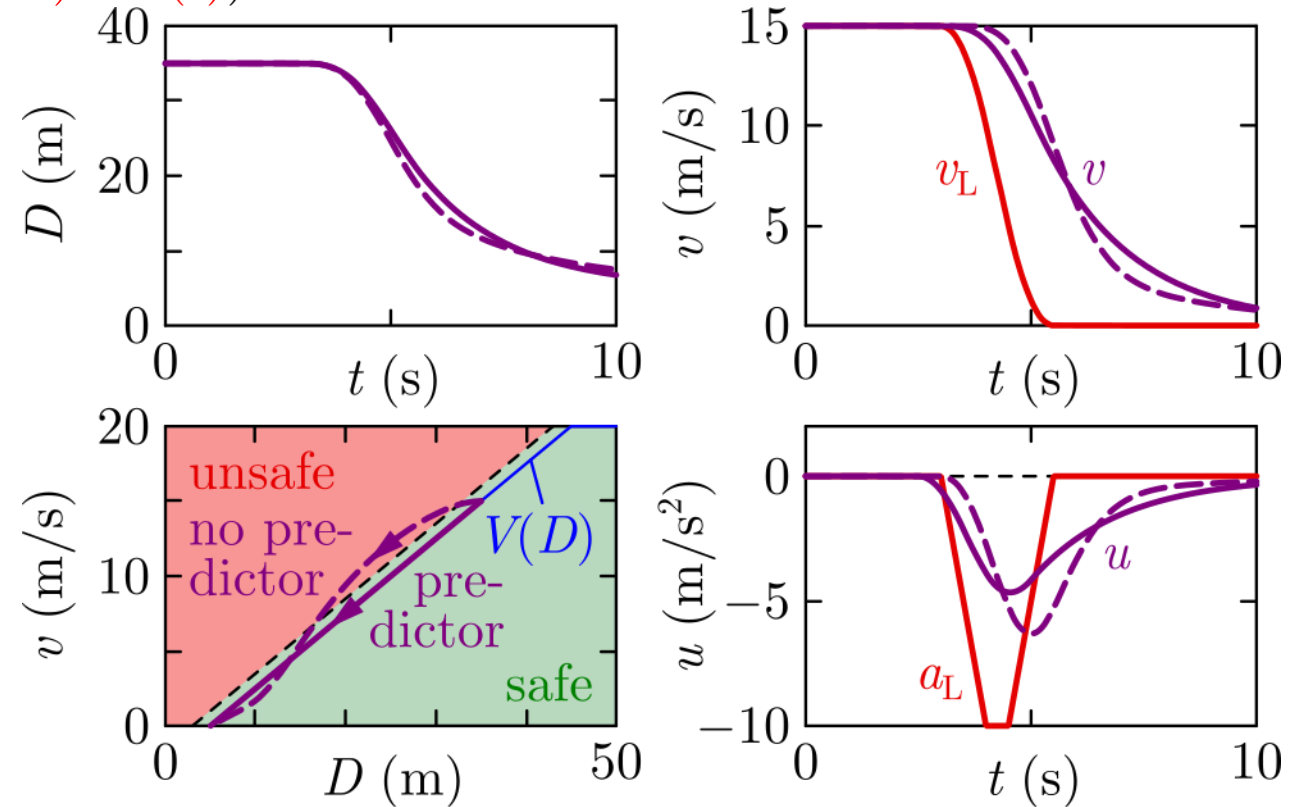
Safety critical control:

$$\dot{h}(t_p, x_p, u) \geq -\alpha h(x_p)$$

$$\hat{u} = \alpha (V(D_p) - v_p) + \frac{1}{T} (\hat{v}_{L,p} - v_p) = u + d$$



Connected Cruise Control



Robust Safety Critical Control is Achieved

Dynamical model:

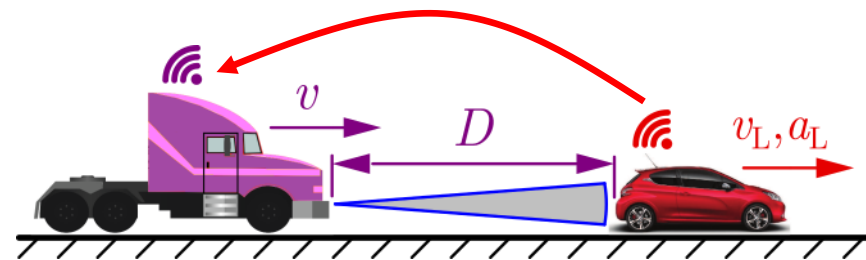
$$\left. \begin{aligned} \dot{D} &= v_L - v \\ \dot{v} &= u(t - \tau) + d(t) \\ \dot{v}_L &= a_L(t) \end{aligned} \right\} \dot{x} = f(t, x) + g(x)(u(t - \tau) + d(t))$$

Disturbance (unmodelled 1st order lag):

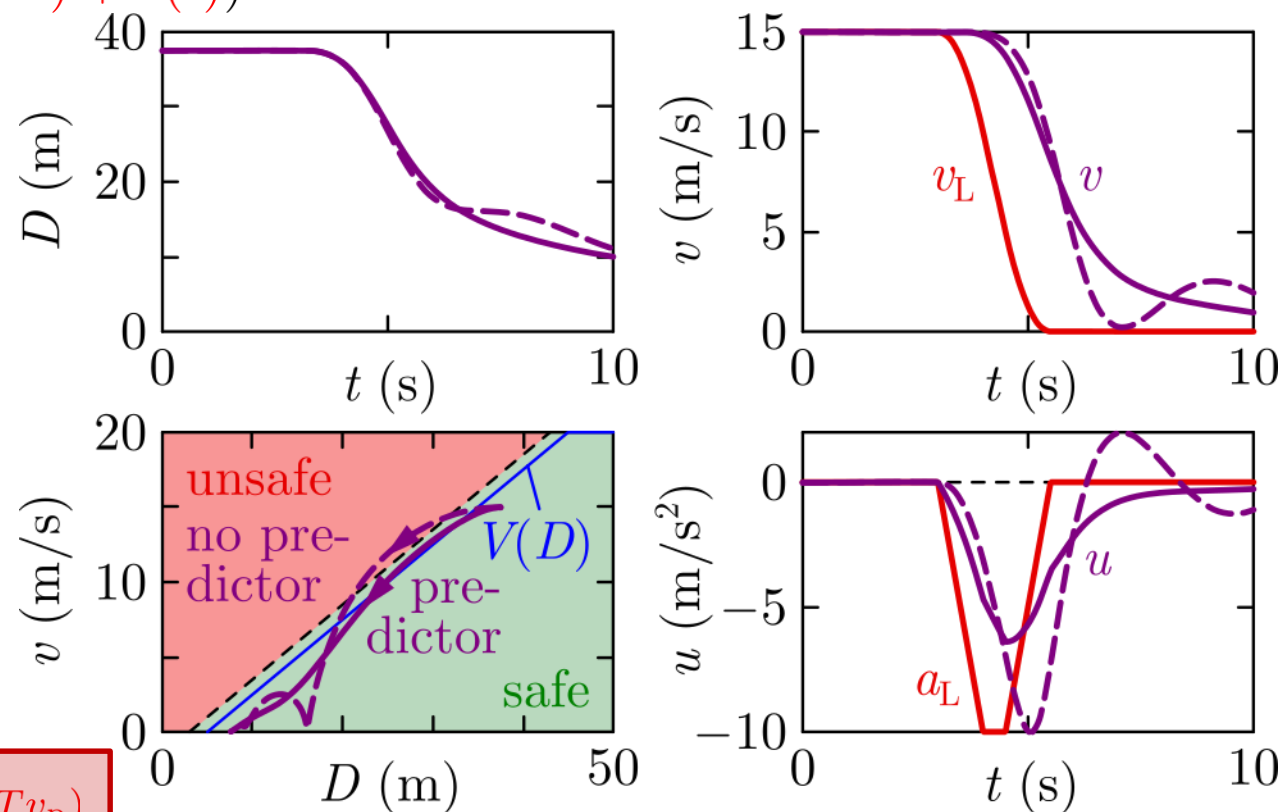
$$\begin{aligned} \dot{a}(t) &= \frac{1}{\xi}(-a(t) + u(t - \tau)), \\ d(t) &= a(t) - u(t - \tau) \end{aligned}$$

Safety critical control:

$$\dot{h}(t_p, x_p, u) \geq -\alpha h(x_p) + \sigma(h(x_p)) \|\nabla h(x_p) g(x_p)\|^2$$

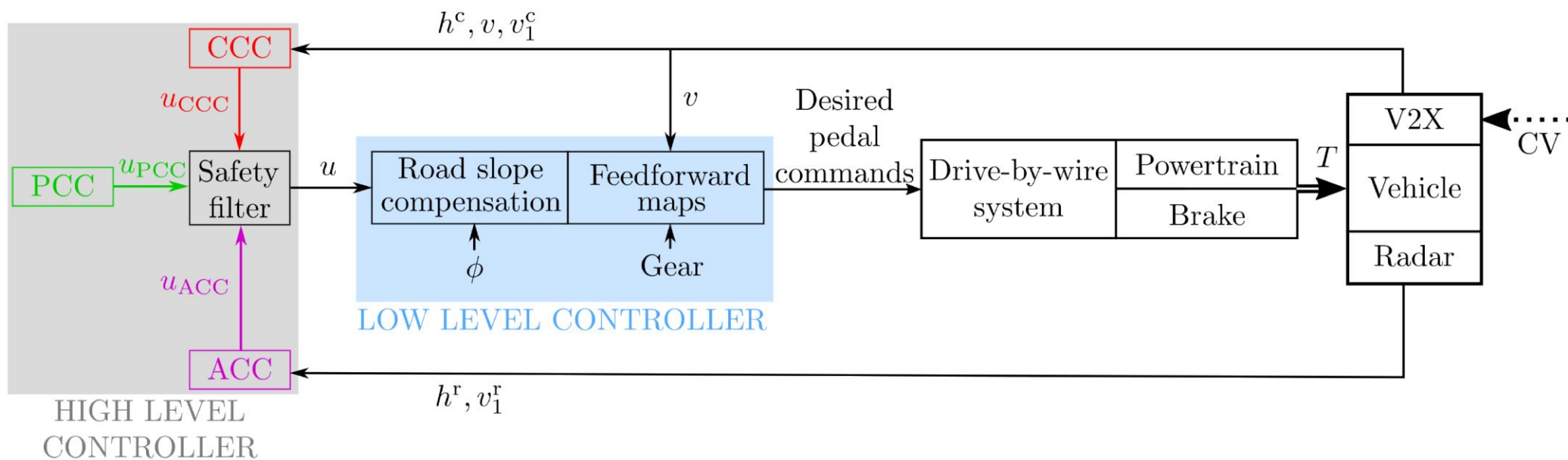
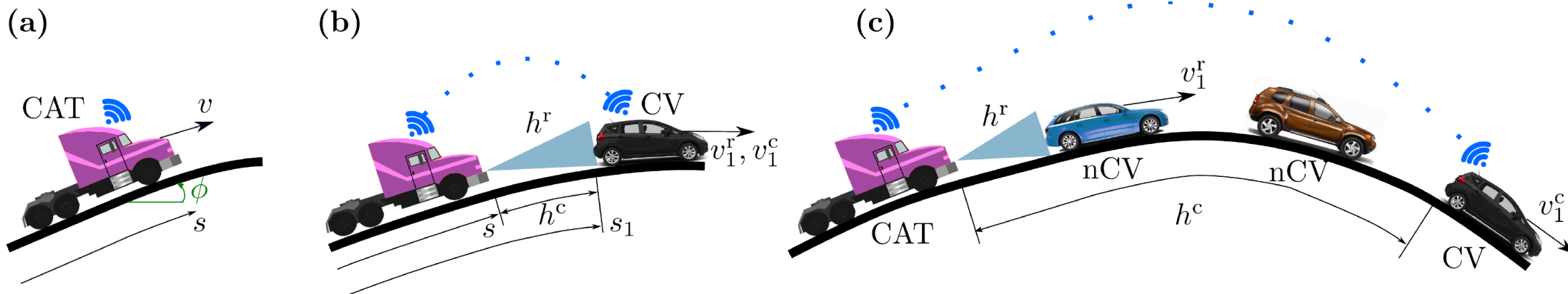


Connected Cruise Control



$$u = \alpha (V(D_p) - v_p) + \frac{1}{T} (\hat{v}_{L,p} - v_p) - T \sigma_0 e^{-\lambda(D_p - D_{sf} - T v_p)}$$

Real-world Experiments

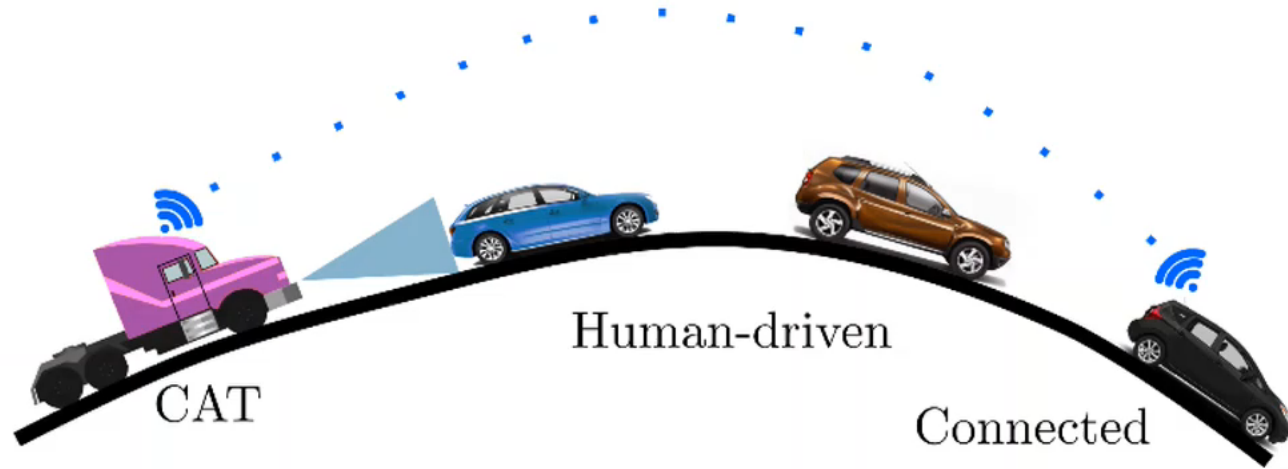


$$u = \min\{u_{PCC}, u_{ACC}, u_{CCC}\}$$

Safe Controller Integration For a Connected Automated Truck (CAT)



NAVISTAR

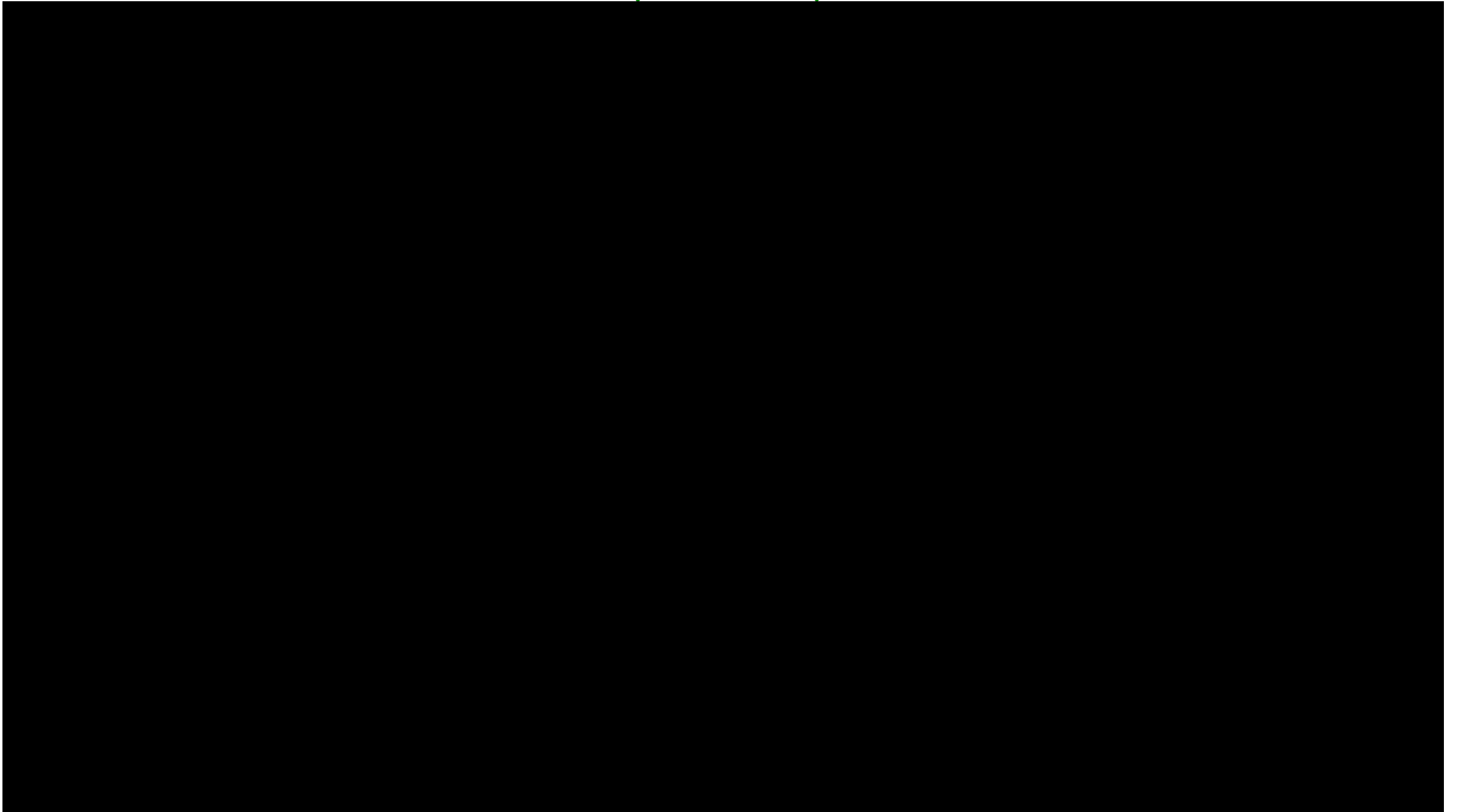


Because Traffic has a Memory



Cheers!

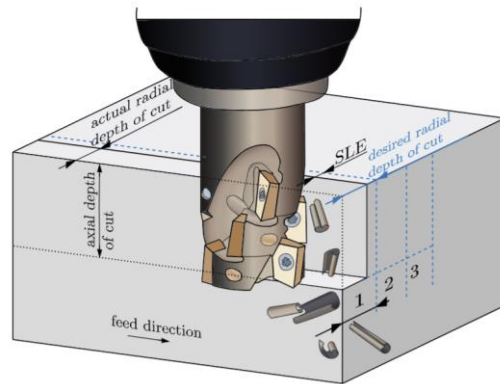
Input Delay



State Delay

$$\dot{x}(t) = f(x(t), x(t - \tau)) + g(x(t), x(t - \tau))u(t)$$

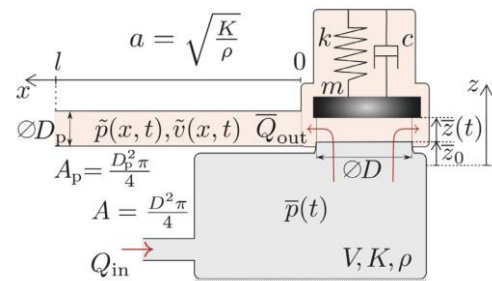
$$h(x(t), x(t - \tau)) \geq 0$$



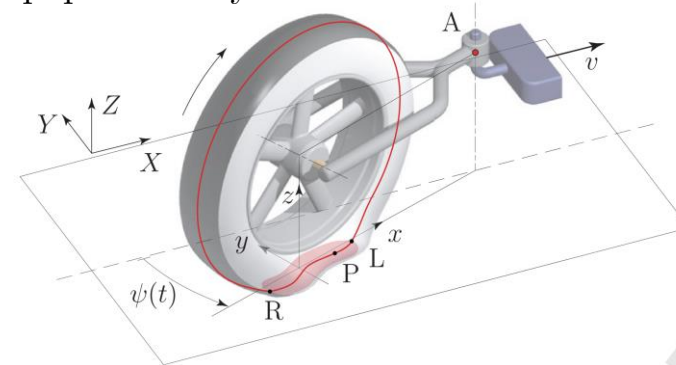
machine tool vibrations



population dynamics



hydraulic systems



tire dynamics

State Delay

$$\dot{x}(t) = f(x(t), x(t - \tau)) + g(x(t), x(t - \tau))u(t)$$

$$h(x(t), x(t - \tau)) \geq 0$$

Infinite dimensional state space

$$x_t(\theta) = x(t + \theta), \quad \theta \in [-\tau, 0]$$

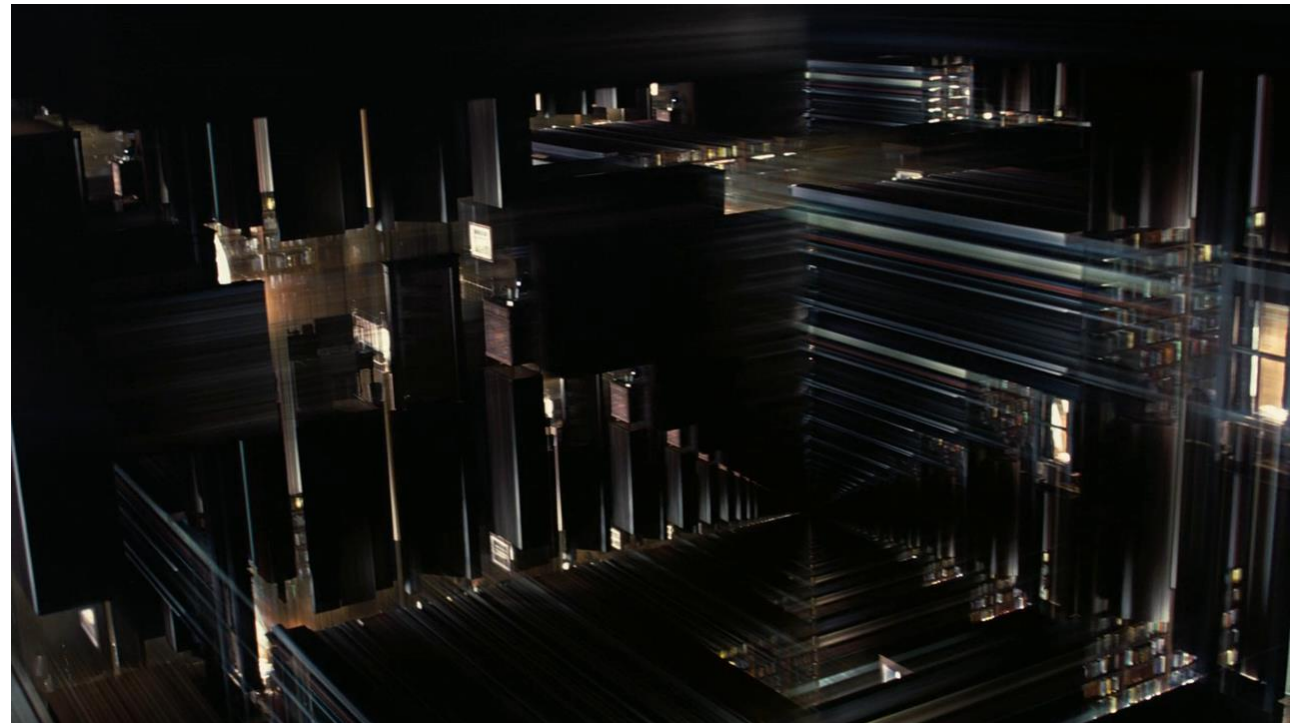
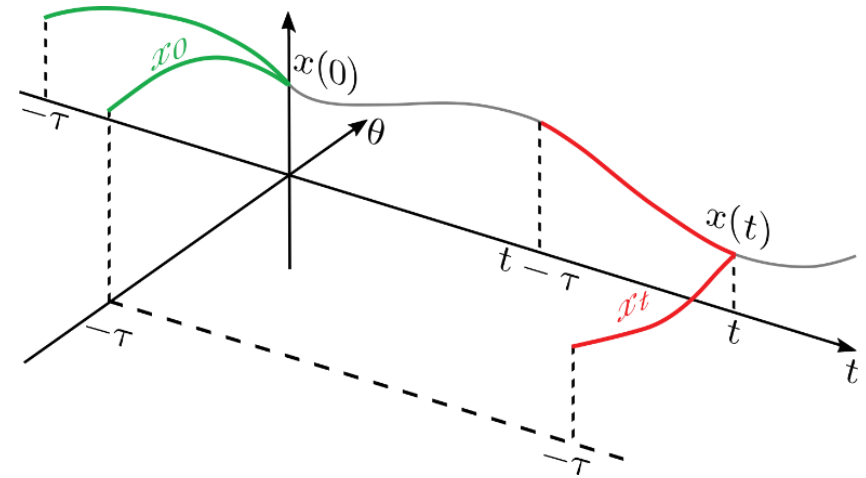
$$x_t \in \mathcal{B} = C([-\tau, 0], \mathbb{R}^n)$$

Functional differential equation

$$\dot{x}(t) = \mathcal{F}(x_t) + \mathcal{G}(x_t)u(t)$$

Control Safety Functional

$$\mathcal{S} = \{x_t \in \mathcal{B} : \mathcal{H}(x_t) \geq 0\}$$

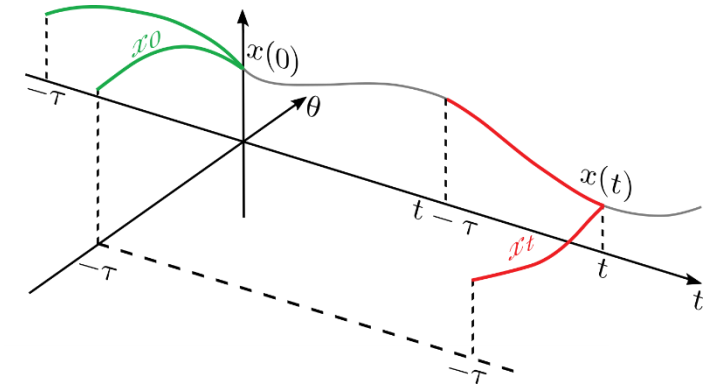


Control Barrier Functionals

Theorem

$$\dot{\mathcal{H}}(x_t, \dot{x}_t) = \int_{-\tau}^0 d_\theta \eta(x_t, \theta) \cdot \dot{x}_t(\theta)$$

Nonlinear functional of x_t
and linear functional of \dot{x}_t



Example

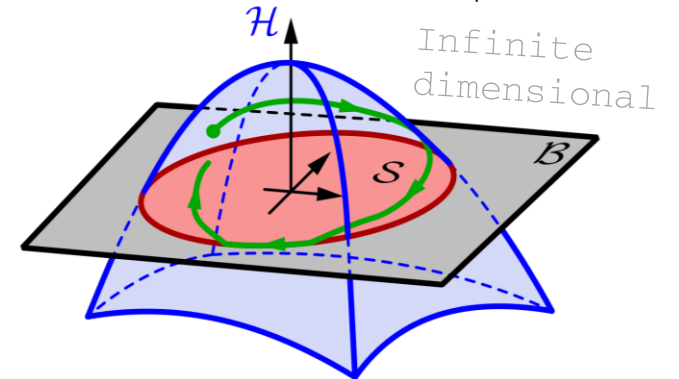
$$\mathcal{H}(x_t) = h\left(x_t(0), x_t(-\tau), \int_{-\sigma_1}^{-\sigma_2} \rho(\vartheta) \kappa(x_t(\vartheta)) d\vartheta\right)$$

headpoint

point delay

distributed delay

$$\dot{\mathcal{H}}(x_t, \dot{x}_t, u) = \underbrace{\nabla_0 h(\cdot)}_{w_0(x_t)} \cdot (\mathcal{F}(x_t) + \mathcal{G}(x_t)u) + \underbrace{\nabla_1 h(\cdot)}_{w_1(x_t)} \cdot \dot{x}_t(-\tau) + \int_{-\sigma_1}^{-\sigma_2} \underbrace{\nabla_2 h(\cdot) \rho(\vartheta) \nabla \kappa(x_t(\vartheta))}_{w_d(x_t, \vartheta)} \cdot \dot{x}_t(\vartheta) d\vartheta$$



Closed Loop Safe System

$$\dot{x}(t) = \mathcal{F}(x_t) + \mathcal{G}(x_t)u(t)$$

No delay in safety condition

$$h(x(t)) \geq 0$$

$$\mathcal{S} = \{x_t \in \mathcal{B} : \mathcal{H}(x_0) \geq 0\}$$

Control law:

$$u = \mathcal{K}(x_t) = k(x(t), x(t - \tau))$$

Functional differential equation:

$$\dot{x}(t) = \mathcal{F}(x_t) + \mathcal{G}(x_t)\mathcal{K}(x_t)$$

Delay in safety condition

$$h(x(t), x(t - \tau)) \geq 0$$

$$\mathcal{S} = \{x_t \in \mathcal{B} : \mathcal{H}(x_t) \geq 0\}$$

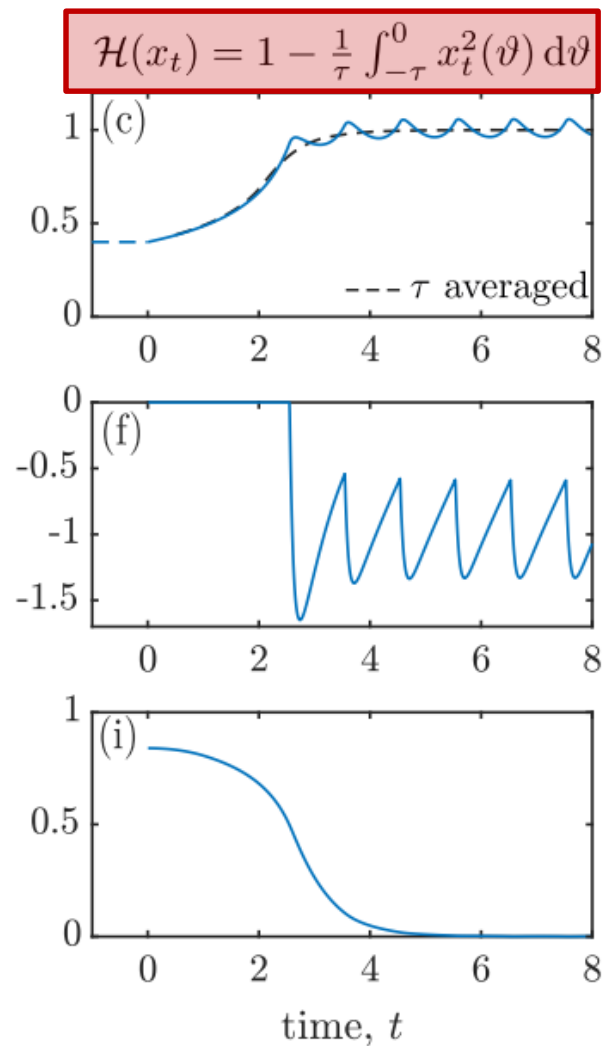
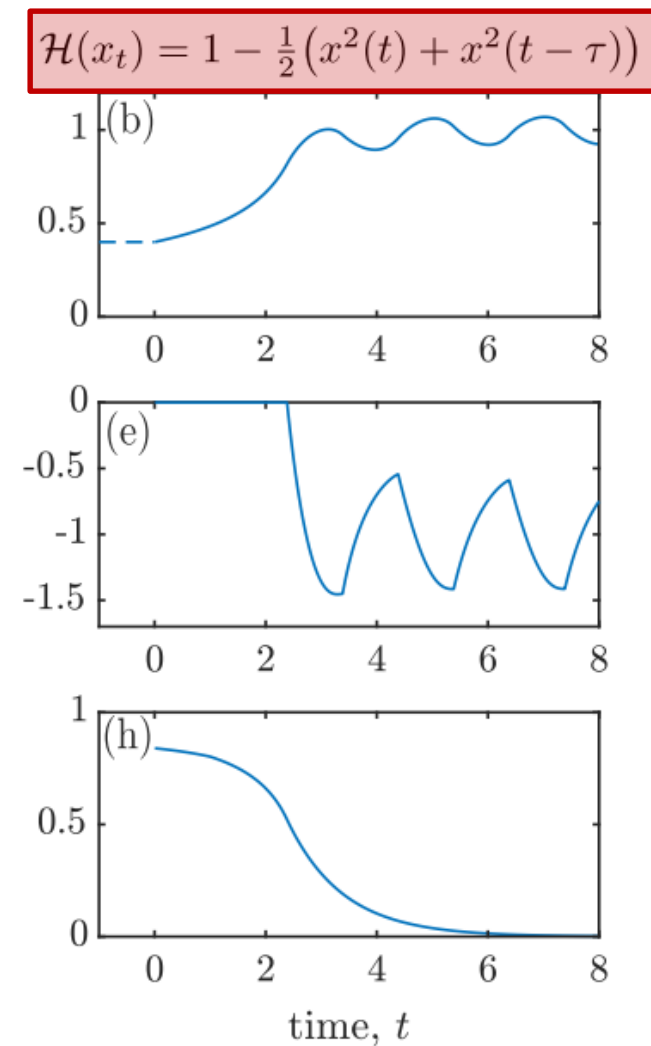
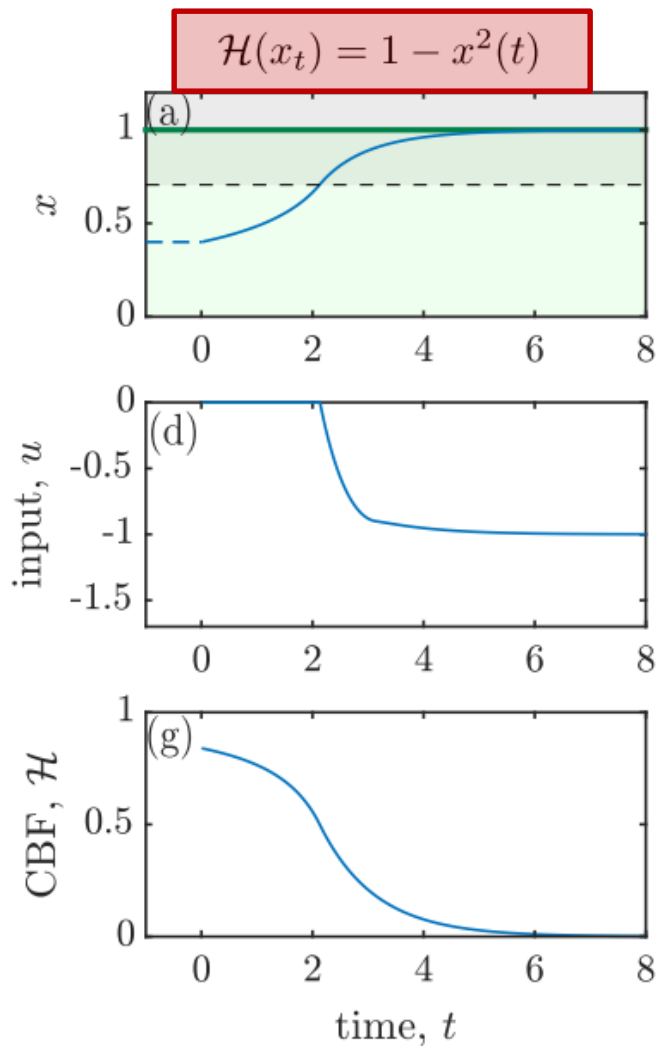
Control law:

$$u = \mathcal{K}(x_t, \dot{x}_t) = k(x(t), x(t - \tau), \dot{x}(t - \tau))$$

Neutral Functional differential equation:

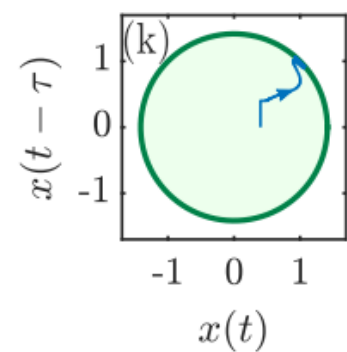
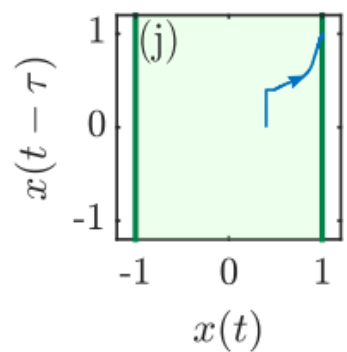
$$\dot{x}(t) = \mathcal{F}(x_t) + \mathcal{G}(x_t)\mathcal{K}(x_t, \dot{x}_t)$$

Example

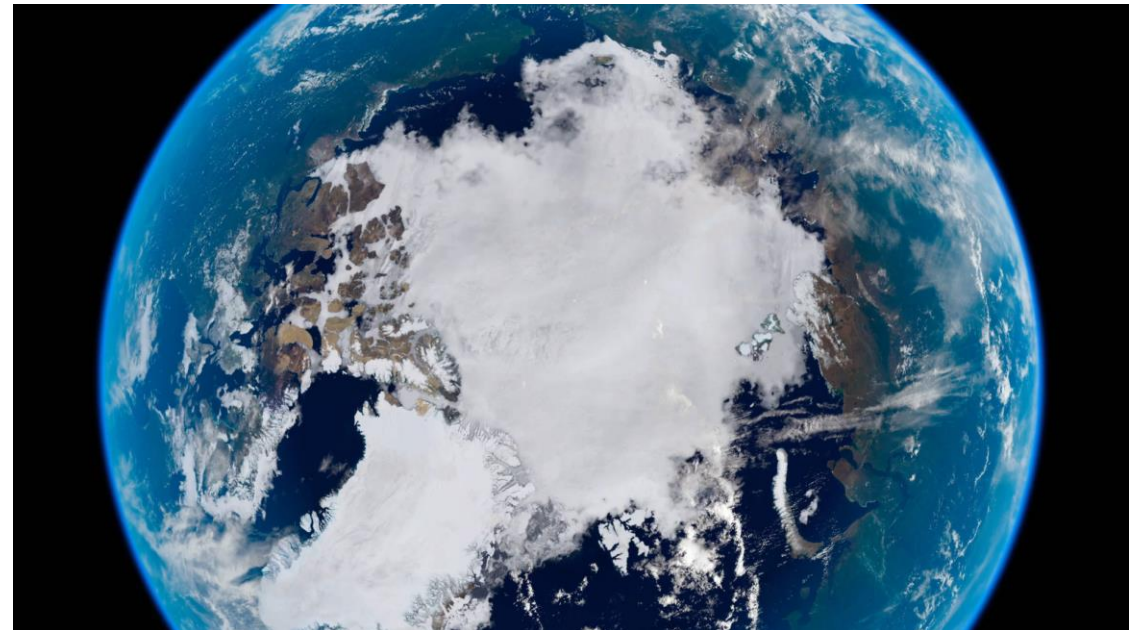
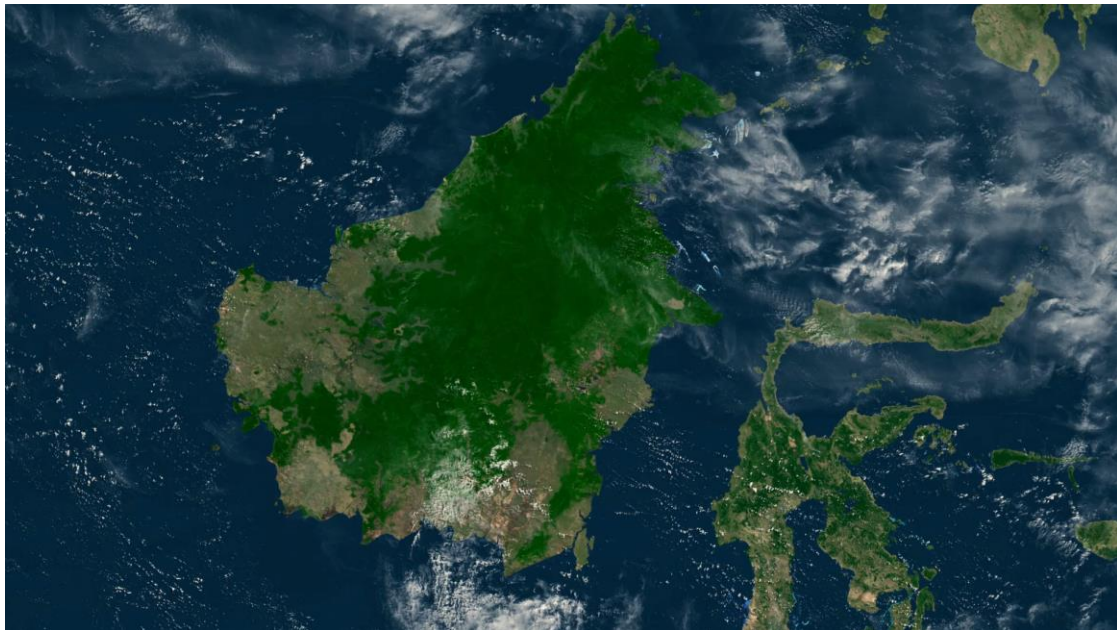


- safe set
- switching surface
- intervention
- no intervention

$\dot{x}(t) = x(t)^3 + x(t - \tau)u(t)$



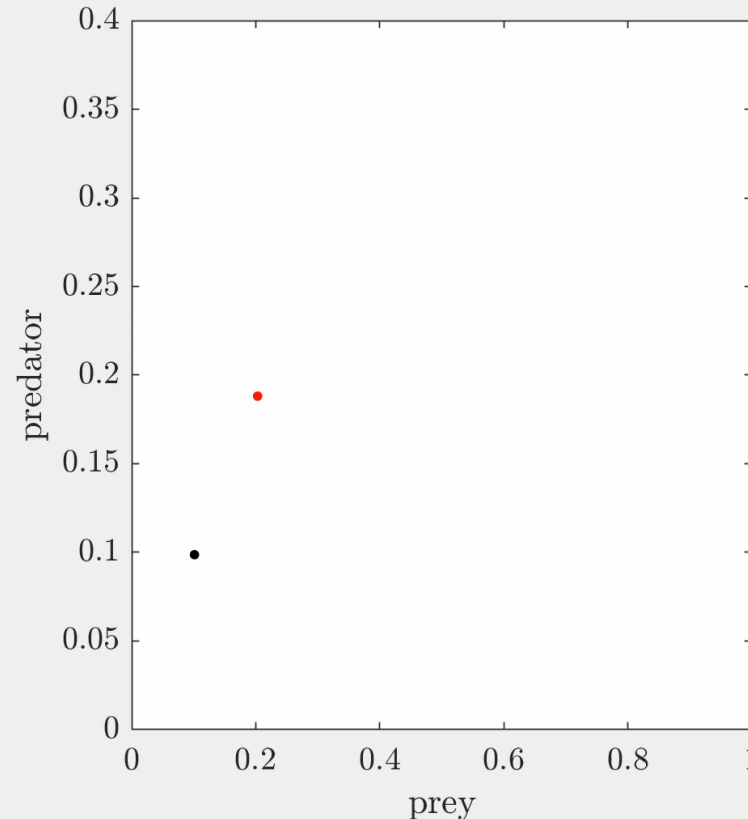
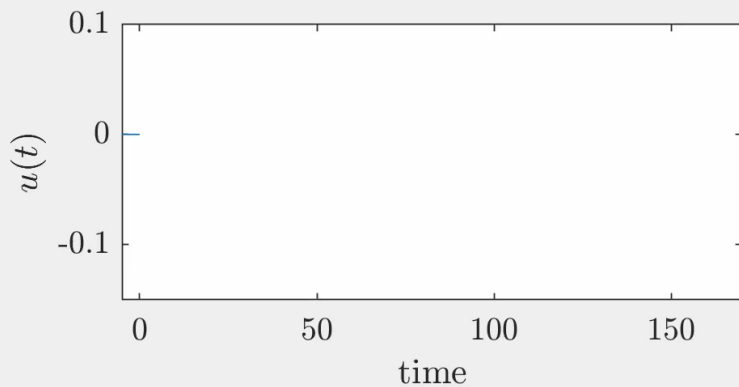
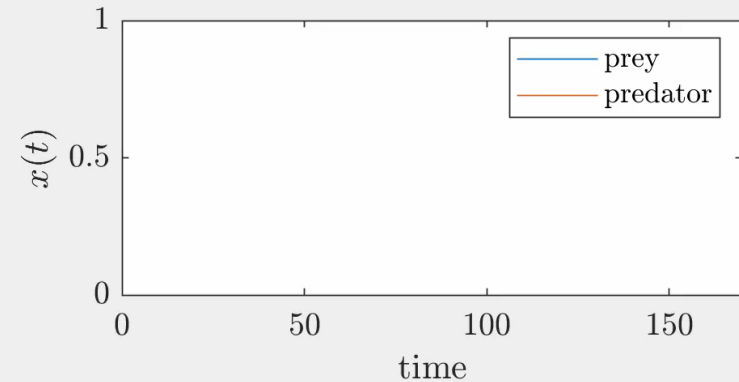
Many Ecosystems are at the Verge of Safety



Population Dynamics

delayed predator-prey model

$$\underbrace{\begin{bmatrix} \dot{y}(t) \\ \dot{z}(t) \end{bmatrix}}_{\dot{x}(t)} = \underbrace{\begin{bmatrix} ry(t) - ay^2(t) - py(t)z(t) \\ bpy(t - \tau)z(t - \tau) - dz(t) - mz^2(t) \end{bmatrix}}_{f(x(t), x(t-\tau))}$$



y prey population
 z predator population

r prey growth rate
 a self-regulation rate
 p predation rate
 b conversion rate
 d predator mortality
 m predators competition

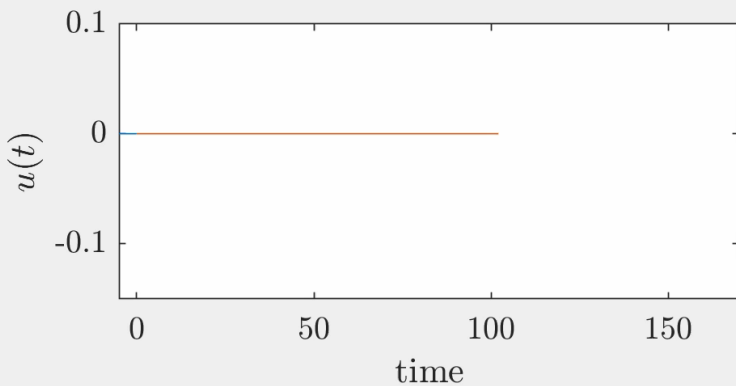
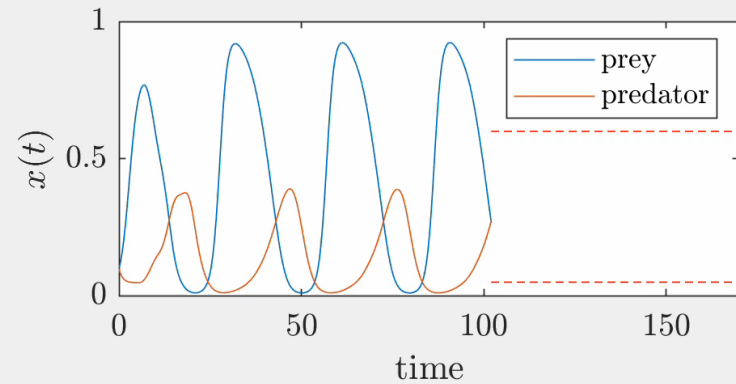
τ maturation time

Population Dynamics

delayed predator-prey model

$$\underbrace{\begin{bmatrix} \dot{y}(t) \\ \dot{z}(t) \end{bmatrix}}_{\dot{x}(t)} = \underbrace{\begin{bmatrix} ry(t) - ay^2(t) - py(t)z(t) \\ bpy(t - \tau)z(t - \tau) - dz(t) - mz^2(t) \end{bmatrix}}_{f(x(t), x(t - \tau))} + \underbrace{\begin{bmatrix} 0 \\ 1 \end{bmatrix}}_g u(t)$$

$$u(t) = K(x(t), x(t - \tau))$$



y prey population
 z predator population

r prey growth rate
 a self-regulation rate
 p predation rate
 b conversion rate
 d predator mortality
 m predators competition

τ maturation time

Collaborators



Anil
Alan



Chaozhe
He



Harvey
Bell



Gábor
Orosz



Andrew
Taylor



Tamás
Molnár



Aaron
Ames



Ádám
Kiss



Budapest University of Technology