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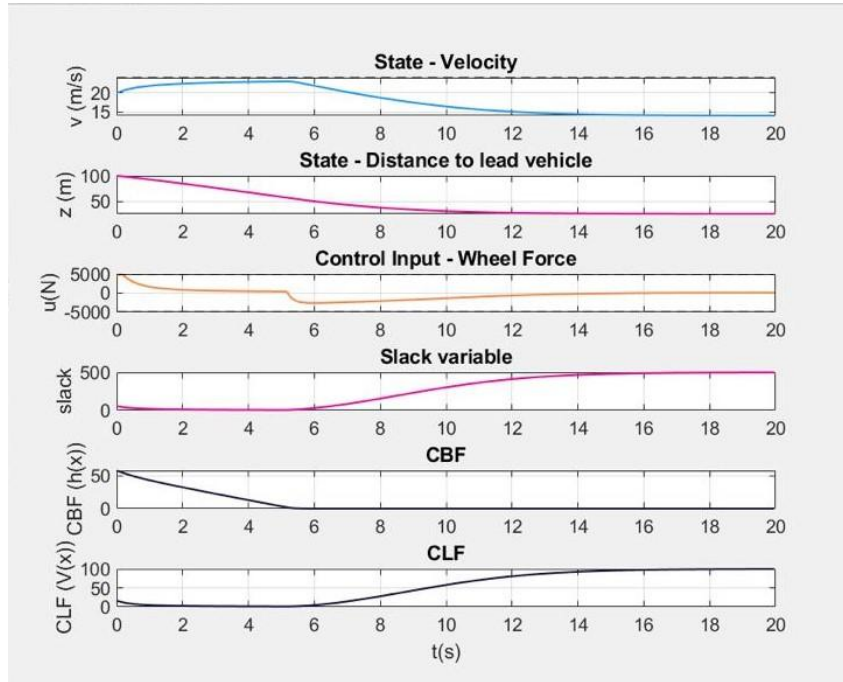
Self-triggered Control for Safety Critical Systems using
Control Barrier
Functions

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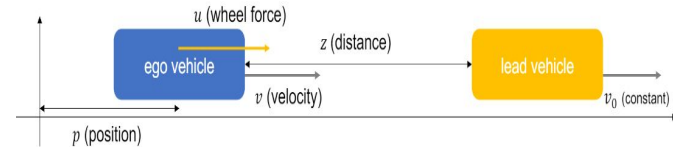
Groundwork

- Started off with understanding the concepts of Control barrier functions and control Lyapunov functions.
- After getting a good grasp on them, worked on implementing a CBF-CLF QP controller for a simple adaptive cruise control system (relative degree 1 system) and on a double integrator system (relative degree 2).
- The adaptive cruise control system example is based on the Control barrier function-based quadratic programs with application to adaptive cruise control paper.

Results from the Adaptive cruise control example.

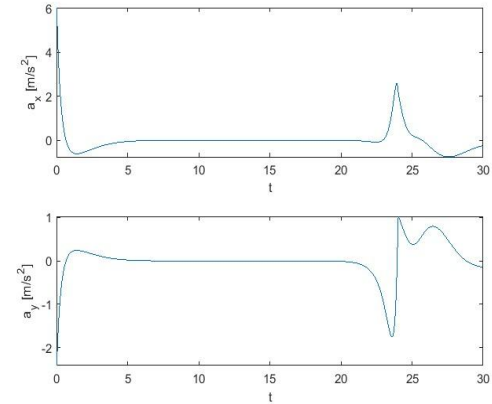
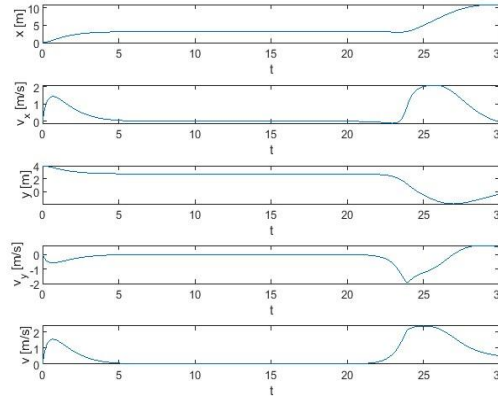
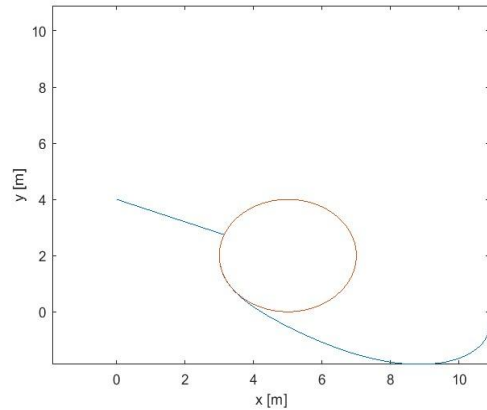


System dynamics of an adaptive cruise control system:



State	Control Input	Input constraints
$x = [p \ v \ z]^T \in \mathbb{R}^3$	$u \in \mathbb{R}^1$	$-mc_d g \leq u \leq mc_u g$
Dynamics		Stability objective
$\dot{s} = \underbrace{\begin{bmatrix} v \\ -\frac{1}{m}F_r(v) \\ v_0 - v \end{bmatrix}}_{f(s)} + \underbrace{\begin{bmatrix} 0 \\ \frac{1}{m} \\ 0 \end{bmatrix}}_{g(s)} u$		$v \rightarrow v_d$ (v_d : desired velocity)
		Safety objective
		$z \geq T_h v$ (T_h : lookahead time)

Results from 2D Double integrator example (HOCBF)



Goal of the project

- Implementation of a real-time control strategy that combines self-triggered control with Control Lyapunov Functions
- Implementation of the controller that overcomes the main limitations of traditional approaches based on periodic controllers, i.e. unnecessary controller updates and potential violations of the safety constraints.
- Central to this approach is the notion of a safe period, which enforces a safety guarantee for implementing ZOH control.

Why self-triggered control for safety critical systems?

- Previous works on CBF-CLF controllers are based on a continuous time formulation, which contradicts the reality that these controllers are implemented on digital platforms.
- Traditionally, digital controllers are implemented using discretized periodic control inputs. A popular discretization method is the Zeroth-Order Hold (ZOH)
 1. Given a fixed update period, there is no guarantee that the safety constraints will hold.
 2. There are unnecessary computations and control updates due to fixed-time sampling.

How do we overcome these issues?

1. To overcome these issues, we can implement self-triggered CBF, The core of all self-triggered controllers consists of two parts. First, a designed feedback controller computes the control input at a given time instance.
2. Second, it determines the next controller update time instance based on sensor measurements and mission requirements.

Introduction

- Implement a self-triggered controller that pre-computes the next update time instance given the current state, control objective, and safety requirements.
- The controller is applied in a ZOH manner.
- To validate the implementation of a self-triggered controller, A second-order double integrator dynamical system is used.
- Compare our self-triggered control strategy with standard periodic control.

Application to second order double integrator system

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

- A continuous time dynamical control affine system.

- The second order double integrator system is defined. And has a relative degree of 2.
The goal is to stabilize our system to a desired state.

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u.$$

Defining the CBF constraints

$h^{rb}(x) = \mathcal{L}_f^{rb} h(x) + \mathcal{L}_g \mathcal{L}_f^{rb-1} h(x) u.$ The time derivative h are related to the Lie derivatives by

$$\xi_b(x) = \begin{bmatrix} h(x) \\ \dot{h}(x) \\ \vdots \\ h^{rb}(x) \end{bmatrix},$$

A transverse variable is defined for ECBF.

$$\mathbf{h}(x) = \begin{bmatrix} h_1(x(t)) \\ h_2(x(t)) \\ h_3(x(t)) \\ h_4(x(t)) \end{bmatrix} = \begin{bmatrix} x_1(t) - x_{1,min} \\ -x_1(t) + x_{1,max} \\ x_2(t) - x_{2,min} \\ -x_2(t) + x_{2,max} \end{bmatrix},$$

Given the dynamic system, The ECBF (Exponential control barrier function) is defined as $h_1(x)$, $h_2(x)$, $h_3(x)$, $h_4(x)$, as the safety constraints.

$$\inf_{u \in U} [\mathcal{L}_f^{rb} h(x) + \mathcal{L}_g \mathcal{L}_f^{rb-1} h(x) u + K_b \xi_b(x)] \geq 0, \forall x \in \text{Int}(C).$$

indicates that the magnitude of the safety constraint is unbounded, and therefore the constraint cannot be violated.

$$\begin{aligned} \zeta_1 &= u + k_1 x_2 + k_2 (x_1 - x_{1,min}), \\ \zeta_2 &= -u + k_1 (-x_2) + k_2 (-x_1 + x_{1,max}), \\ \zeta_3 &= u + k (x_2 - x_{2,min}), \\ \zeta_4 &= -u + k (-x_2 + x_{2,max}). \end{aligned}$$

If $\zeta_i \geq 0$, $i = 1, \dots, 4$ holds, then our system is forward invariant.

Defining the CLF constraints

$V(x) = \begin{bmatrix} x_1 - x_{1,d} \\ x_2 \end{bmatrix}^T \begin{bmatrix} 1 & 0.5 \\ 0.5 & 1 \end{bmatrix} \begin{bmatrix} x_1 - x_{1,d} \\ x_2 \end{bmatrix}.$ The Lyapunov Function candidate for this example is defined as $V(X)$.

$$c_1 \|x\|^2 \leq V(x) \leq c_2 \|x\|^2,$$

$$\inf_{u \in U} [\mathcal{L}_f V(x) + \mathcal{L}_g V(x)u + \epsilon V(x)] \leq 0, \quad \forall x \in \mathbb{R}^n.$$

Definition: Exponentially stabilizing control Lyapunov function.

- Based on the definition and the Lyapunov candidate our constraint can be defined as:

$$\eta(x) = [2x_2 + (x_1 - x_{1,d})]u + x_2(2(x_1 - x_{1,d}) + x_2) + \epsilon V.$$

CBF-CLF QP formulation.

The QP formulation for system (25) is

$$\begin{aligned} \min_{u \in U} \quad & u^T u \\ \text{s.t.} \quad & \zeta_i \geq 0, i = 1, \dots, 4 \\ & \eta \leq 0 \\ & x(t_k) \in \text{Int}(C) \\ & u_l \leq u \leq u_u. \end{aligned}$$

Computation of CBF safe periods

- We find a bound on the system trajectory that exclusively depends on the general properties of the system dynamics.
- Given the dynamical system defined in starting at $x(t_k)$ the distance between the trajectory $x(t+t_k)$ and $x(t_k)$ is bounded by: $\bar{r}_{t_k}(t) = r_0 e^{L(t-t_k)} - \frac{1}{L} \|f(x(t_k)) + g(x(t_k))u_k\|$.
- With lower bound $\zeta(t)$, we determine safe period τ_{CBF} , such that lower bound $\zeta(t_k + \tau_{\text{CBF}}) = 0$.
- The CBF safe period τ_{CBF} is then calculated by finding the minimum of the roots of these equations. This was achieved by implementing the Bisection method.

$$\begin{aligned}\underline{\zeta}_1 &= (k_1(x_2(t_k) - r_{t_k}(t)) - k_2\|u_k\|)t + \zeta_1(t_k), \\ \underline{\zeta}_2 &= (-k_1(x_2(t_k) + r_{t_k}(t)) - k_2\|u_k\|)t + \zeta_2(t_k), \\ \underline{\zeta}_3 &= -k\|u_k\|t + \zeta_3(t_k), \\ \underline{\zeta}_4 &= -k\|u_k\|t + \zeta_4(t_k).\end{aligned}$$

Derivation of the distance bound on system trajectory

$$r_{t_k}(t) = \|x(t_k + t) - x(t_k)\|$$

$$\begin{aligned}\dot{r}(x(t + t_k)) &= \frac{(x(t + t_k) - x(t_k))^T}{\|x(t + t_k) - x(t_k)\|} \dot{x}(t + t_k) \\ &= \frac{(x(t + t_k) - x(t_k))^T}{\|x(t + t_k) - x(t_k)\|} f(x(t + t_k), u).\end{aligned}$$

$$\begin{aligned}\dot{r}_{t_k} &\leq \|f(x(t + t_k), u)\| \\ &\leq \|f(x(t + t_k), u) - f(x(t_k), u) + f(x(t_k), u)\| \\ &\leq \|f(x(t + t_k), u) - f(x(t_k), u)\| + \|f(x(t_k), u)\|,\end{aligned}$$

$$\begin{aligned}\bar{r}_{t_k}(t) &= L\|x(t + t_k) - x(t_k)\| + \|f(x(t_k), u)\| \\ &\leq L\bar{r}(t + t_k) + \|f(x(t_k), u)\|.\end{aligned}$$

$$\bar{r}_{t_k}(t) = r_0 e^{L(t-t_k)} - \frac{1}{L} \|f(x(t_k)) + g(x(t_k))u_k\|.$$

$$r_0 = \frac{1}{L} \|f(x(t_k)) + g(x(t_k))u_k\|.$$

Computation of CLF safe period

- To achieve at least asymptotic stability, we must define a CLF update period that guarantees that the Lyapunov function decreases at every step.

- Upper bound of $V(t)$: $\bar{V}(t) \geq V(x(t)), \forall t_{k+1} \geq t \geq t_k,$

$$V(x(t)) \leq V(t_k) + (t - t_k)V'(t_k) + (t - t_k)^2 \frac{D}{2} = \bar{V}(t).$$

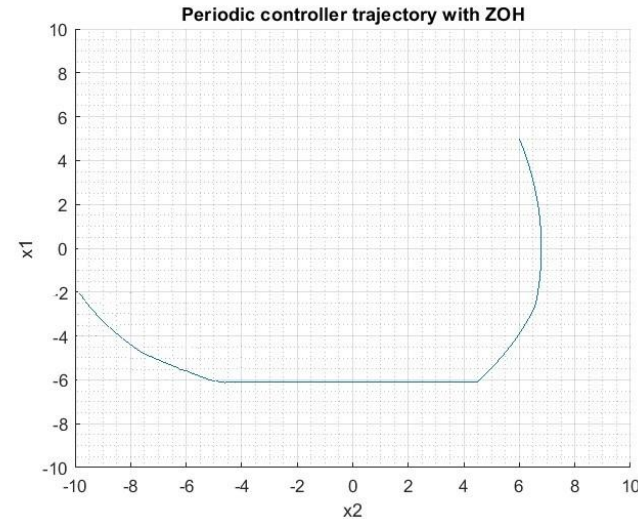
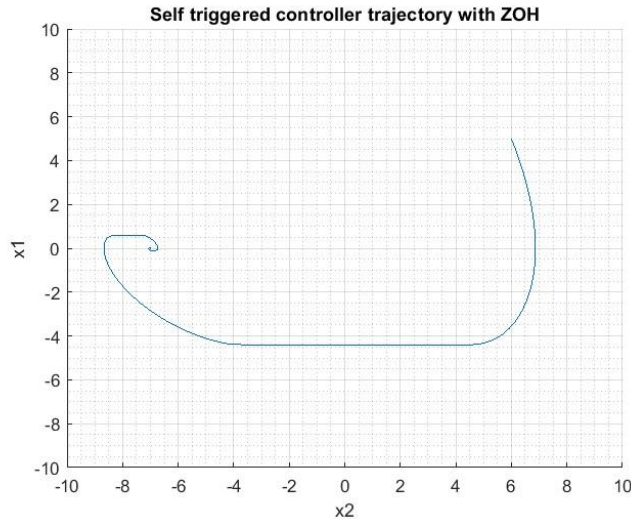
- Since the upper bound of $V(t)$ is a quadratic function in terms of t , there exists a closed-form solution for the roots. The non-zero root is given by: $\tau_{CLF} = \frac{-2V'(x(t_k))}{D}.$

$$V'(x(t_k)) = 2x_2(x_1 - x_{1,d}) + x_2^2 + ((x_1 - x_{1,d}) + 2x_2)u_k$$

- Here the first derivative of $V(x(t_k))$ is $\dot{D} = \max_t V''(x(t))$

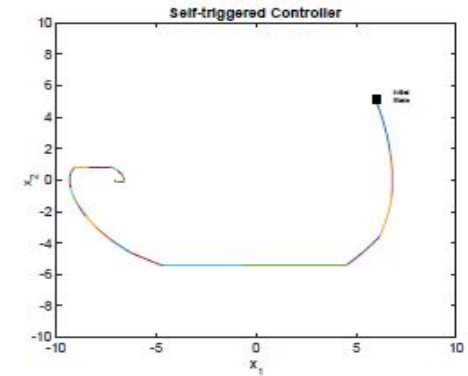
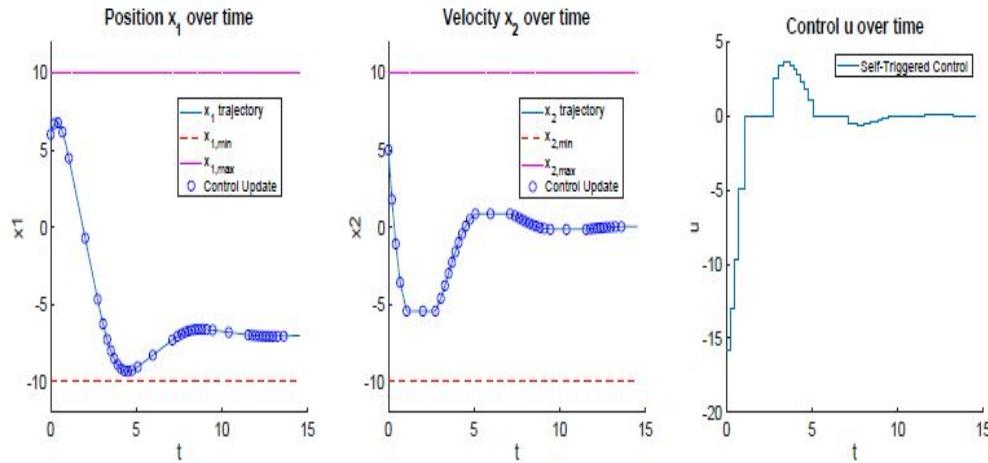
- And the denominator D is given by: $= 2V(x_{t_k}) + 2|u_k|\sqrt{V(x_{t_k})} + 3|\sqrt{V(x_{t_k})}||u_k| + 2|u_k|^2$

Results: comparing the trajectory generated by both the controllers

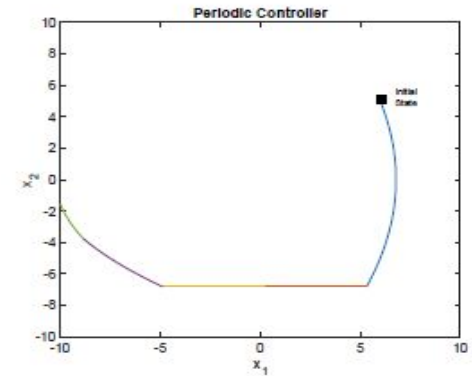


- In the self triggered controller case CBF constraint does not conflict with the input constraints whereas in the periodic controller case it does, the position x_1 violates $x_{1,\min}$.

Comparing the results from the paper.

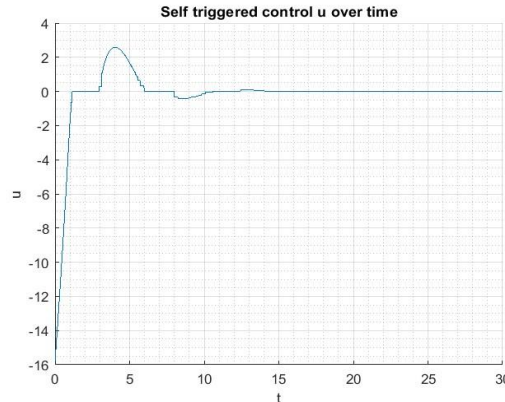


(a)

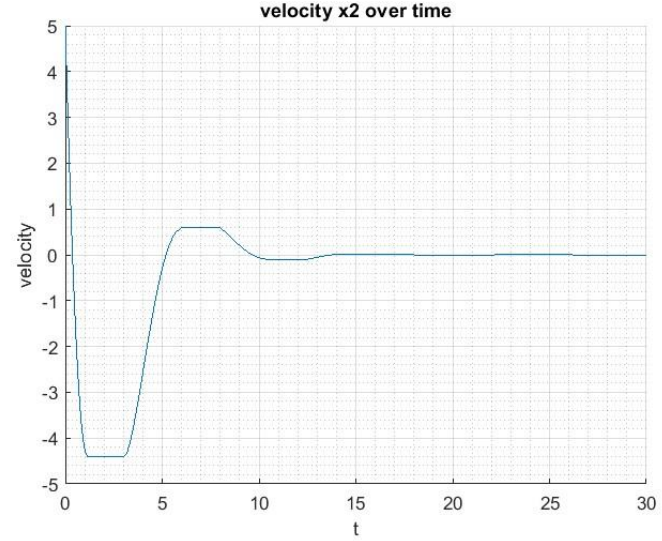
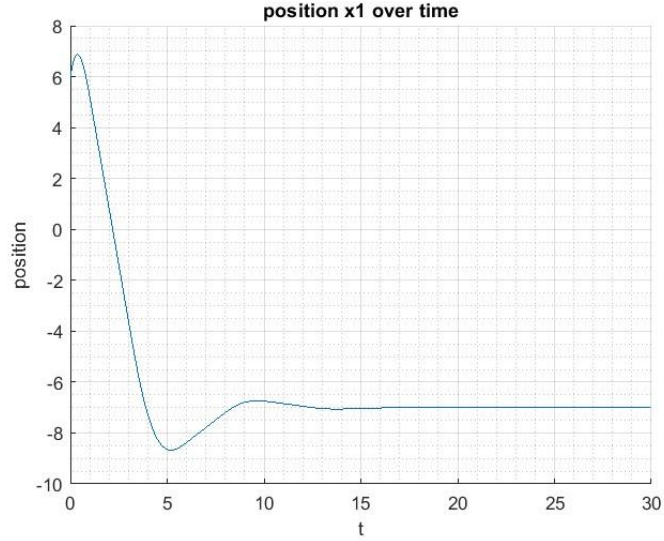


Results: Optimal control generated by self-triggered controller

- For self-triggered control, the controller only updates when the system is about to violate CBF constraints, or the system is deviating away from the desired states.
- the update interval for self-triggered controller becomes a lot faster as the system approaches to the unsafe region ($x_1 < x_{1,\min}$) in order to prevent violation on safety constraint.



Result: Position and velocity of the system for self-triggered controller



Conclusion and remarks

- The self-triggered controller is designed to obtain the safety property from CBFs while stabilizing a system asymptotically.
- In this example, since we cannot directly control the state x_1 , the use of ECBF becomes necessary. ECBF framework is used to ensure the controller works for system with high relative degrees.
- The CLF safe period follows an increasing trend as the system reaches the equilibrium state which contradicts the results of the paper. (working on resolving this issue)