







# Autonomous racecar control in head-to-head competition using Mixed-Integer QP

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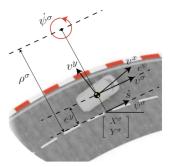
# Introduction: the problem definition

- Ego Vehicle, EV and Leading Vehicle, LV perform the head-to-head competition for achieving fastest lap-time
- LV's trajectory is known by EV during its execution
- EV is responsible for the collision-avoidance during overtaking and its control strategy is calculated online



# Introduction: the base-solution

In [Verschueren et al., 2016], a spatial Nonlinear Model Predictive Control (NMPC) method is proposed for time-optimal racing in a curvilinear coordinate system.



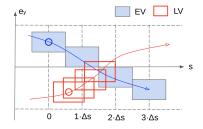


Figure: Curvilinear coordinate system. [Frasch\_et al., 2013] Figure: The prediction horizon in MPC.



### Introduction

#### The NMPC problem formulation



#### Method

#### Over-approximate vehicle's shape as a set

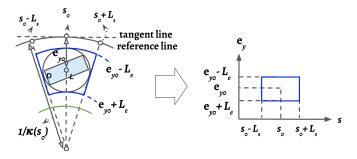


Figure: Vehicles' shape is firstly approximated as a circle and then projected as a set (blue sector) in the curvilinear coordinates system.



# Method

#### Set up collision-avoidance constraint

There should be no overlap between two shape approximation of EV and LV:

$$(A) \ s_{i}^{LV} + (L_{s})_{i}^{LV} \leq s_{i}^{EV} - (L_{s})_{i}^{EV} \\ \vee (B) \ s_{i}^{EV} + (L_{s})_{i}^{EV} \leq s_{i}^{LV} - (L_{s})_{i}^{LV} \\ \vee (C) \ e_{y_{i}}^{LV} + (L_{e})_{i}^{LV} \leq e_{y_{i}}^{EV} - (L_{e})_{i}^{EV} \\ \vee (D) \ e_{y_{i}}^{EV} + (L_{e})_{i}^{EV} \leq e_{y_{i}}^{LV} - (L_{e})_{i}^{LV}$$
(2)

Formulate it as a mixed-integer problem.

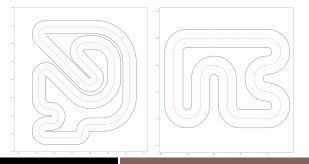


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# Simulation results

#### Experimental configuration

- use a 1:43 miniature racecar model (car size: 6.2cm × 3cm) and two different tracks (track width: 34cm) [Liniger et al., 2015][Verschueren et al., 2016]
- suppose that EV and LV have the same dynamics and LV's optimal trajectory is pre-calculated





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#### Simulation results

Track	Predeiction horizon length	# of collisions (in H2H*)	Mean lap time [s]		Mean calc. time [ms/step]	
			H2H*	Single*	H2H*	Single*
1	15	3/24	4.942	4.852	247	137
	30	0/24	4.899	4.773	905	245
2	15	0/45	10.278	10.189	244	118
	30	0/45	10.148	10.064	832	205

\*H2H = Head-to-head mode, Single = Single car racing mode

- ▶ There might be collisions when horizon length is short.
- By increasing horizon length, we obtain better lap time while the calculation time increases too.



#### [animation]



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### Conclusions and future works

- the effectiveness of the collision-avoidance time-optimal control algorithm.
- potential possibility to implement the algorithm with a short prediction horizon on a real-world vehicle or wheel-robot.
- potential possibility to reduce computation time for a long prediction horizon by simplifying the decision combinatorics etc.



Thanks for your listening!

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Liniger, A., Domahidi, A., and Morari, M. (2015).
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