Layered controller synthesis for dynamic multi-agent systems

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Introduction

Dynamic multi-agent system's verification

https://perso.eleves.ens-rennes.fr/people/Emily.Clement/Videos/example_episodes/ex_0.mp4



Issues of different methods

- ▶ Timed Automata: issues to · · ·
 - 1) represent speed variation
 - 2) scales to be executed in real-time

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 - 2) continuous aspects

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Solve a (simplified) model with an efficient Timed Automata reachability algorithm

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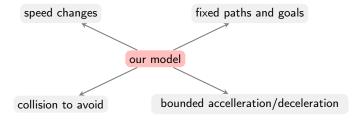
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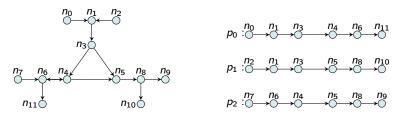
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- Solve a (simplified) model with an efficient Timed Automata reachability algorithm
- Relax the simplification assumption for the speed changes using an SMT solver
- ▶ Generate an SWA-SMT solver to help RL solving this problem.

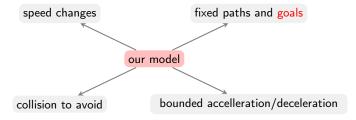
• Our model



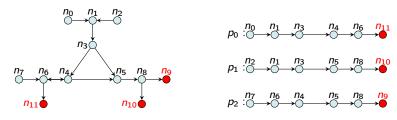
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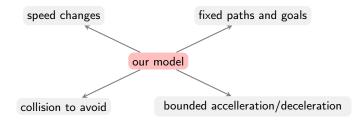


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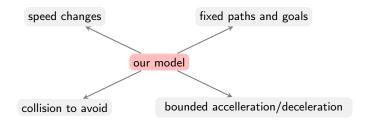


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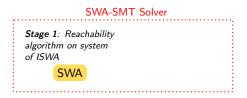


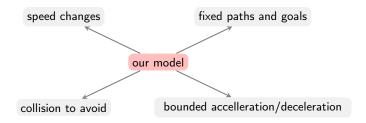


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 - ▶ Goal: reach goals while avoiding collisions between agents

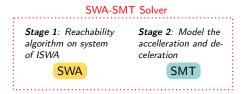


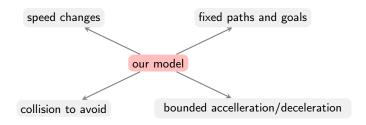
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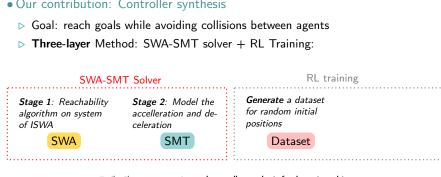


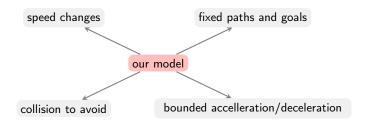
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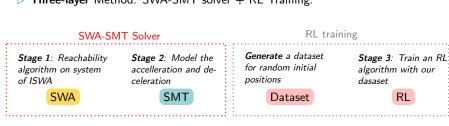


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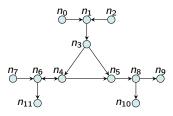


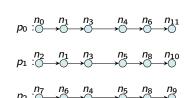
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How to model a Car Traffic ?

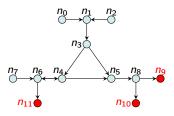
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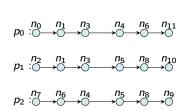




How to model a Car Traffic ?

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Rules to model collision avoidance

- •#1: security distance when driving in the same direction and between neighbouring sections
- •#2: cars cannot share a section if driving in **opposite** direction
- •#3: No Overtaking between cars

SWA-SMT solver

SWA solver

Stage 1: Reachability algorithm on system of ISWA

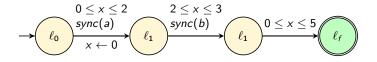
SWA

Stage 2: Model the accelleration and deceleration

SMT

What is a Timed Automaton (and its variants)?

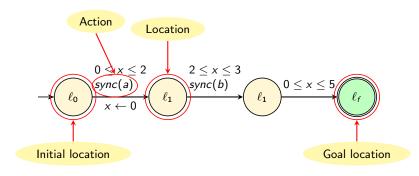
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Variants

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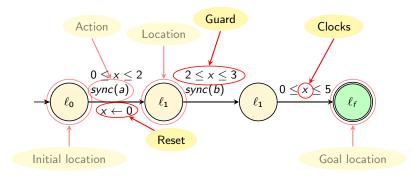
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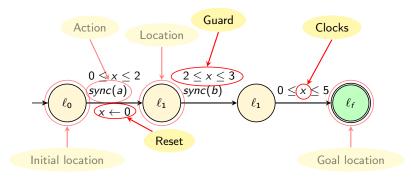
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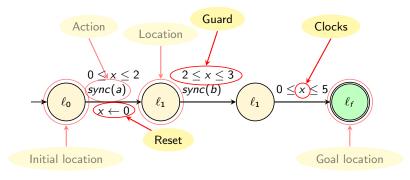
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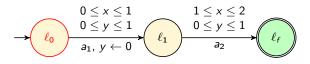
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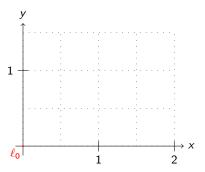
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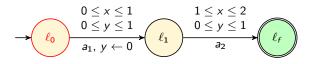
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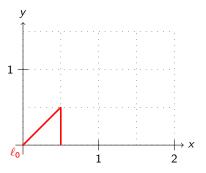


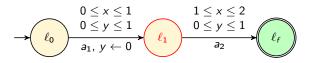
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 - Channels: FiFo queue of symbols (actions) to be pushed/read

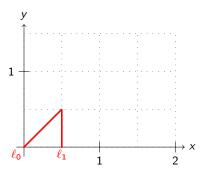


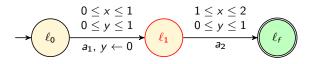


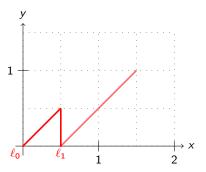


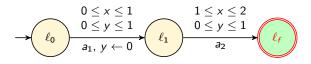


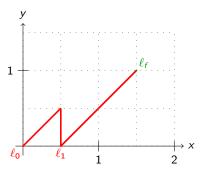


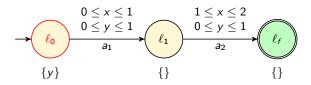


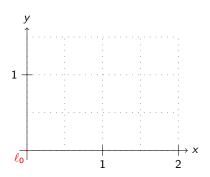


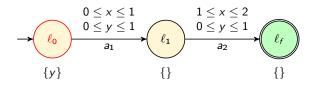


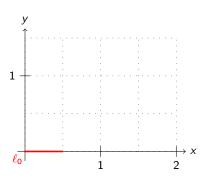


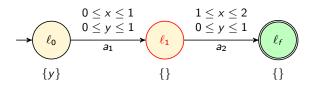


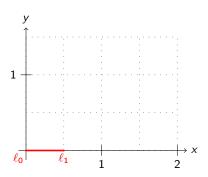


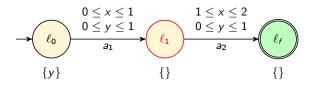


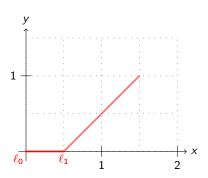


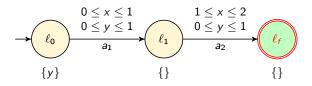


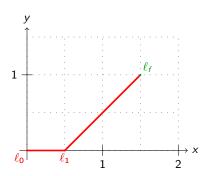






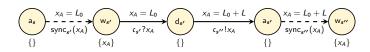






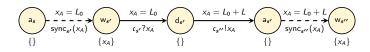
Automata-based Modelisation: car progress

- Car A progress along its paths
 - $\triangleright x_A$: distance travelled along its paths



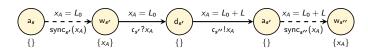
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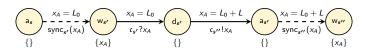
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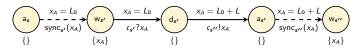


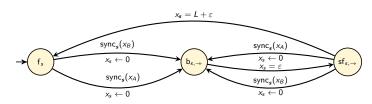
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 - Intersection: use classical synchronized action to active intersection automata

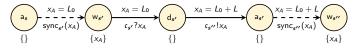


• Reminder: car automaton

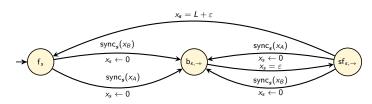




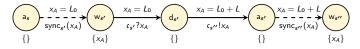
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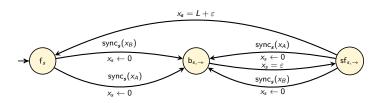
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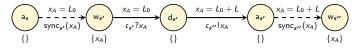
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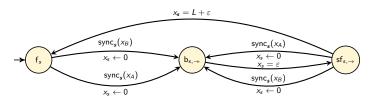


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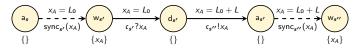


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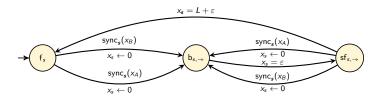


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- ▶ Forbid cars to drive in both direction at the same time.



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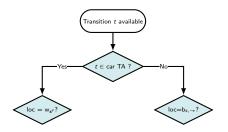
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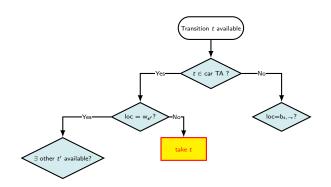
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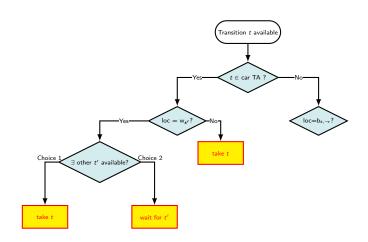
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 - ▶ They have a specific form (few cycles, stopwatches are always in the same locations...)

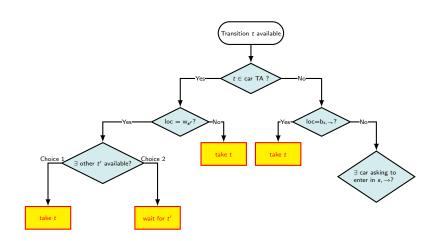
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 - Our contribution: a specified reachability algorithm: a DFS with an optimised successor function.

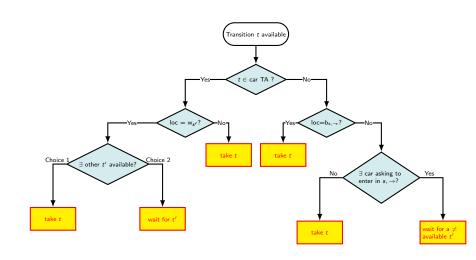






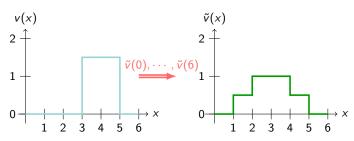






SWA-SMT solver

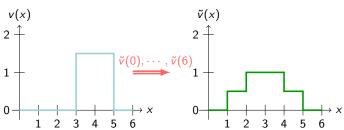
SMT solver



New model for speed graph

- Model speed as a constant piecewise affine function
 - ▶ A more realistic model that takes into account the **dynamic of the system**
 - Different car speed
 - **Bounds** on deceleration and acceleration

$$v_i \Rightarrow \tilde{v}_i(0), \cdots, \tilde{v}_i(k-1)$$



New positions : $\tilde{x}_i(k) = \sum_{l=0}^{k-1} \tilde{v}_i(l)$

How to still ensure security distance?

$$\tilde{x}_i(k) = \sum_{l=0}^{k-1} \tilde{v}_i(l)$$

- SMT solver
 - ▶ Variables: $(\tilde{v}_i(I))_I$
 - \triangleright Constraints: linear inequalities w.t.r \tilde{v}_i and some constants (max acceleration, deceleration...)

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- Example of constraints given to the SMT solver for each step k:
- $(1) \ \tilde{v}_i(k) d_{\mathsf{max}} \leq \tilde{v}_i(k+1) \leq \tilde{v}_i(k) + a_{\mathsf{max}}$
- (2) $0 \leq \tilde{v}_i(k) \leq v_{\text{max}}$

RL training

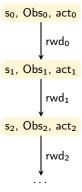
Generate a dataset for random initial positions

Dataset

Stage 3: Train an RL algorithm with our dasaset

RL

Trajectories



• Discounted returns:

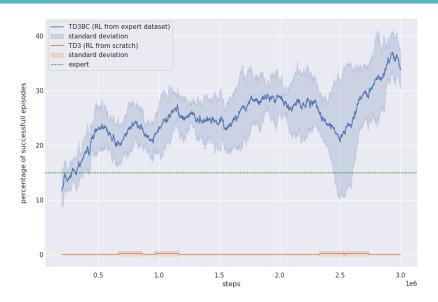
 $\mathsf{rwd}_0 + \mathsf{discount} \cdot \mathsf{rwd}_1 + \mathsf{discount}^2 \cdot \mathsf{rwd}_2 + \cdots$

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 - > Sparce reward signal is a hard exploration problem
- Our problem is not easy to solve with single RL training:
 - Combinatorial aspects
 - Discrete and continuous decision together



Conclusion

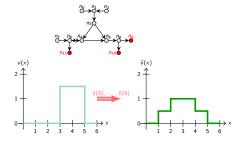
SWA-SMT Solver

Automata-based model

Efficient algorithm
Abstract model with unrealistic speed model

Piecewise-affine speed graph

Bounded accelleration and deceleration Different speed SMT solver to model and solve the distance constraintes



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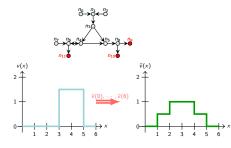
• RL training

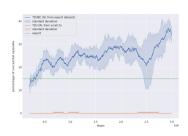
Dataset

Trace generated with SWA-SMT solver Random positions

Performance of RL (helped with SWA-SMT solver)

Better than single RL Better than SWA-SMT solver Runtime: ~ 2 days





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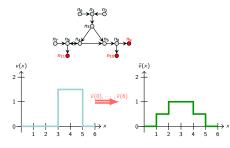
• RL training

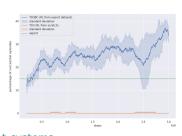
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Appendix

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Formal translation of these rules

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▶ Same directed section: for all cars $c_i, c_j \in C$, $c_i \neq c_j$, for all $t \geq 0$, if $sect_d(c, t) = sect_d(c', t) = s$ then:

$$|p_s(c_i,t)-p_s(c_j,t)|\geq \varepsilon$$

▶ **Neighbouring sections**: for all cars $c_i = (i, [\cdots, (s', \mathbf{d}'_i), \cdots]) \in \mathcal{C}$, if there exists a car $c_j = (j, [\cdots, (s, \mathbf{d}_j), (s', \mathbf{d}'_j), (s'', \mathbf{d}''_j), \cdots]) \in \mathcal{C}$ we have two cases:

If $m{d}_i' = m{d}_j'$: then for all t s.t. ${\sf sect}_d(c_i,t) = (s',m{d}_i')$, ${\sf sect}_d(c_j,t) = (s'',m{d}_j'')$ we have

$$L'-p_{(s',d'_i)}(c_i,t)+p_{(s'',d''_i)}(c_j,t)\geq \varepsilon.$$

If $m{d}_i' \neq m{d}_j'$: then for all t s.t. ${\sf sect}_d(c_i,t) = (s',m{d}_i')$, ${\sf sect}_d(c_j,t) = (s,m{d}_j)$ we have

$$L' - p_{(s',d'_i)}(c_i,t) + L - p_{(s,d_i)}(c_j,t) \ge \varepsilon$$

▷ Same section, opposite direction: for all section $s \in S$, for all $t \ge 0$ and for each pair of cars $c_i, c_j \in C$:

$$\neg(\operatorname{sect}_d(c_i,t)=(s,\rightarrow)\wedge\operatorname{sect}_d(c_j,t)=(s,\leftarrow))$$

No overtaking: we use channels to respect the order of cars.