

# Constrained Path planning with RRT\*

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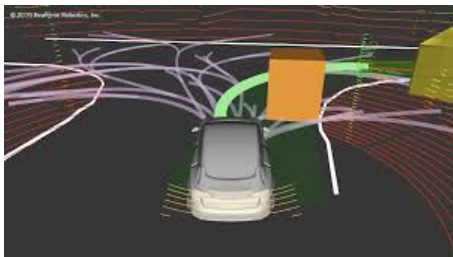
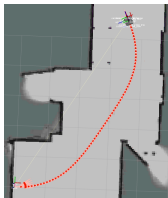
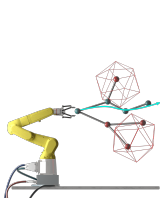
# Introduction - Path Planning

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Planning a path within a 2D or 3D space is fundamental in robotics.

A few major algorithms :

- Dijkstra
- A\*
- RRT



# Introduction - Path planning

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## Limitations

Issues generally encountered :

- No pre-existing graph
- Convergence towards an optimal solution
- Account for complex path constraints

## Proposition

A variant of the RRT algorithm,  $RRT^*$

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- 2 Kinematic constraints
- 3 Results
  - Without kinematic constraints
  - With kinematic constraints
- 4 Conclusion and perspectives

# RRT\* algorithm

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**Result:** RRT\* tree  $T$

Initialization;

**while** *Not converged* **do**

    Sample random configuration  $q_{\text{rand}}$ ;

    Find nearest node  $q_{\text{near}}$  to  $q_{\text{rand}}$  in  $T$ ;

    Steer towards  $q_{\text{rand}}$  from  $q_{\text{near}}$  to get  $q_{\text{new}}$ ;

**if**  $q_{\text{new}}$  *is collision-free* **then**

        Find nearby nodes  $N_{q_{\text{new}}}$  in  $T$ ;

        Choose parent  $q_{\text{min}}$  from  $N_{q_{\text{new}}}$  minimizing cost;

        Rewire  $T$  to  $q_{\text{new}}$  if it improves cost-to-come;

        Add  $q_{\text{new}}$  to  $T$  with  $q_{\text{min}}$  as parent;

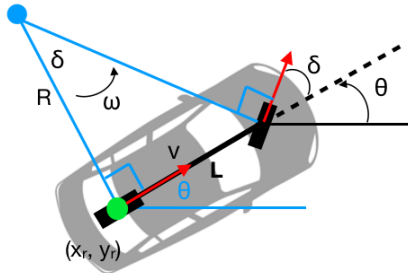
**end**

**end**

**Algorithm 2:** RRT\* Algorithm

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# Kinematic model

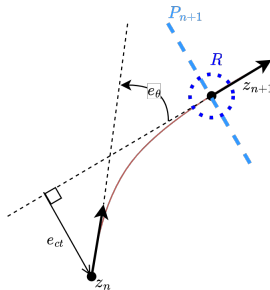


$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{v} \\ \dot{\theta} \end{bmatrix} = \begin{bmatrix} v \cos(\theta) \\ v \sin(\theta) \\ a \\ \frac{v \tan(\delta)}{L} \end{bmatrix} = f(X, u) \quad \text{with } u = [a, \delta]^T \quad (1)$$

# Control strategy

$$a_c = \frac{v^2}{R} = \frac{v^2 \sin(\delta)}{L}$$

$$v_R = \min(v_{n+1}, \sqrt{\frac{a_{c_{max}} L}{\sin(\delta)}})$$



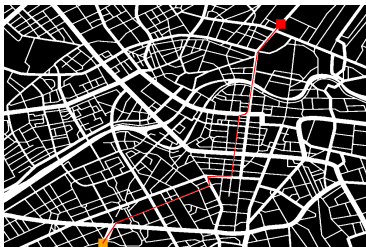
## Commands

$$\begin{cases} a = \text{clamp}_{[-a_m, a_m]}(K_p(v_R - v)) \\ \delta = \text{clamp}_{[-\delta_m, \delta_m]}(K_{str} \arctan(K_{ct} e_{ct}) + K_{str} e_{\theta}) \end{cases} \quad (2)$$

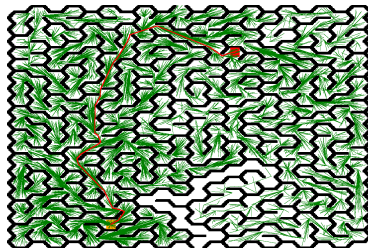


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# First results



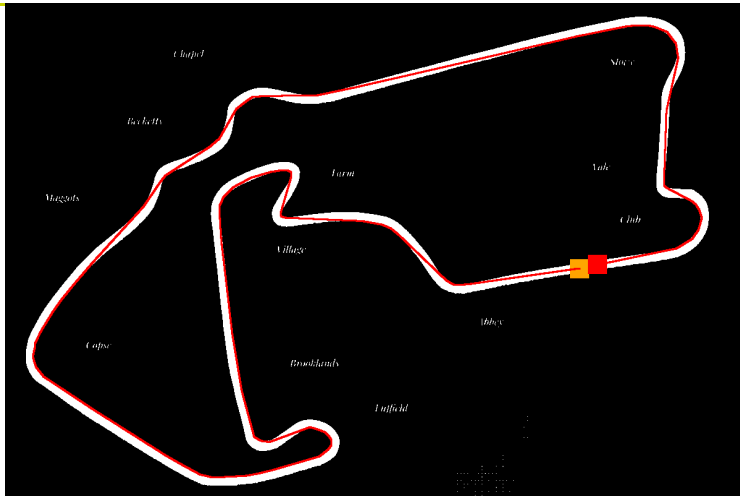
(a) Path found on an urban map



(b) Tree graph (green) and path (red) in an hexagonal maze

**FIGURE** – Shortest paths using euclidean distance and no kinematic constraints

# First results



**FIGURE** – Shortest (but obviously not fastest) trajectory on the Silverstone racetrack

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# Choice of the $C$ Cost function

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## Different choices for $C(x, u)$

- 1 Path length :  $C(x, u) = \int_0^{t_{end}} |v(t)| dt$
- 2 Total time :  $C(x, u) = t_{end}$
- 3 Energy consumption, tire fatigue, ...

# Finding the fastest path



(a) Simulation of the fastest path with 2114 nodes



(b) Reference race line on the Silverstone racetrack

**FIGURE** – Attempting to find the fastest path on the Silverstone racetrack

# Conclusion

## Limitations

- Implementation issues
- Optimization (ex : better node sampling)

## Results

- Rather satisfactory results
- Many applications (ex : parking manoeuvres)

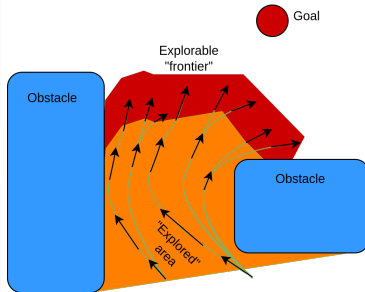


FIGURE – Proposition of a dynamic generation of nodes

# References I

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## Conclusion

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# End of the presentation