



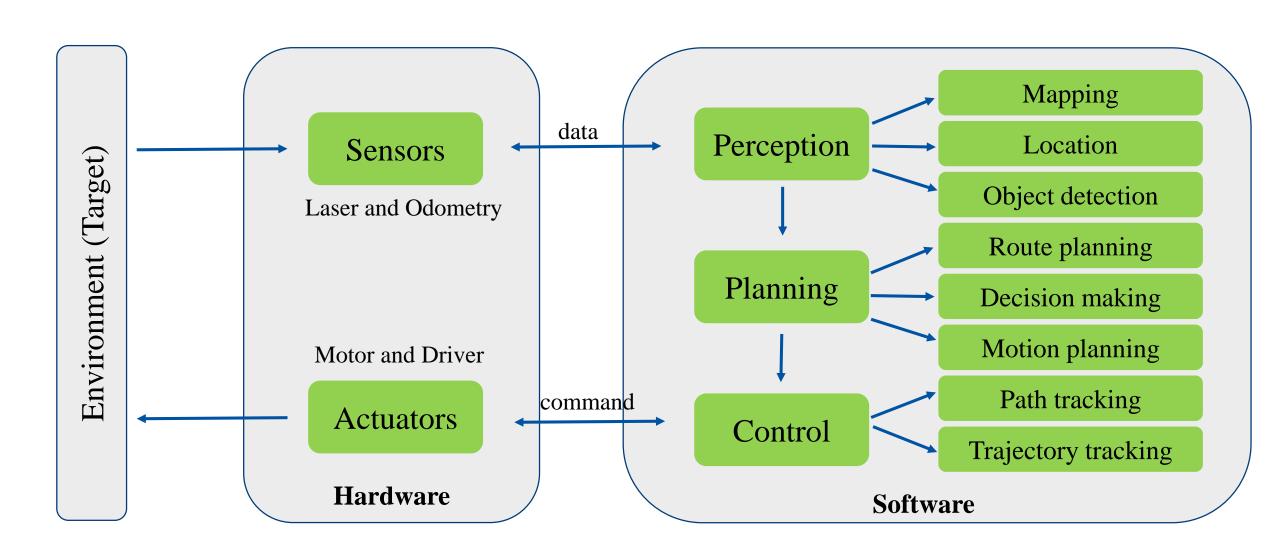
A Review of Motion Planning Techniques



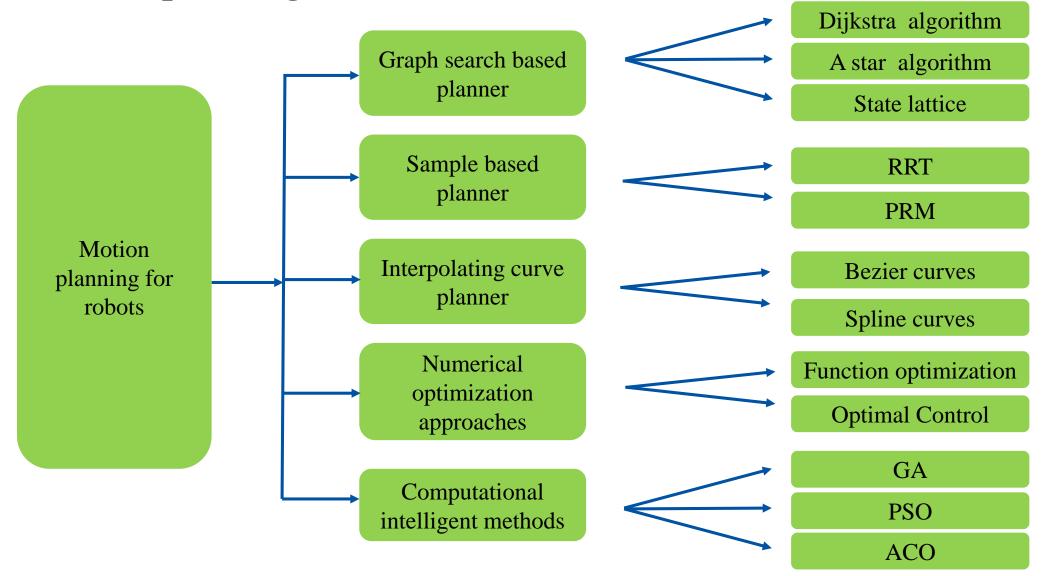
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• Framework



Motion planning state of art

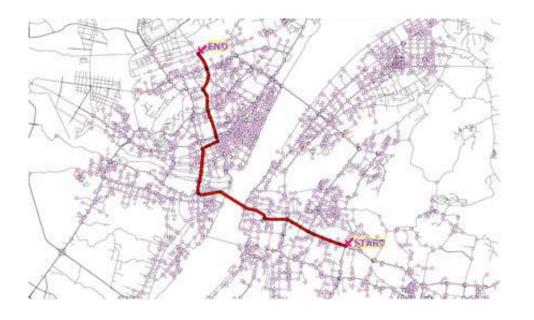


0. Framework

- **Basic conception**
- Path planning
- Motion planning
- Trajectory planning

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• Dijkstra algorithm

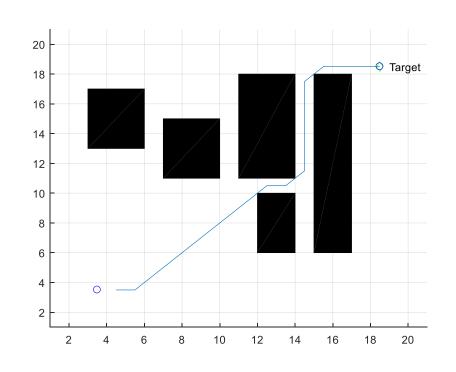


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• A* algorithm

- In A Star we maintain 2 lists valid and in_valid
- Valid List contains
- o All valid nodes as we compute costs for obstacle free cells
- Invalid List contains
- o Obstacle containing cells
- o Cells that are included in path
- Implementation
- o Compute costs of all the adjacent cells to the current cell
- o Add them to valid list
- o Find cell with minimum total cost from valid list, make it parent for next iteration and make it unavailable for next minimum total cost comparison
- o Put that in in_valid list so that we don't visit it again
- o Keep doing this until the cell with minimum cost is not target cell.

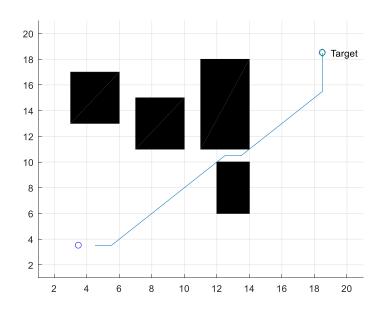


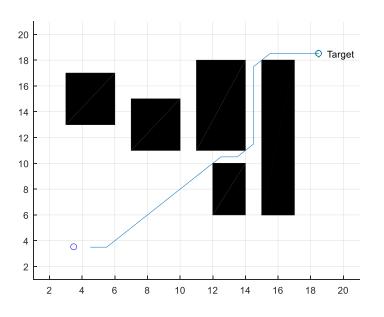
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• D* algorithm

- D star is almost similar to *A star, as the name says its robust to dynamic appearing of obstacles.
- If after computing the path an obstacle occurs while traversing the path following will be done:
- o Update in_valid list with new obstacles.
- o Remove all the obstacle cells from valid list.
- o Increase the total cost for all adjacent nodes where first obstacle was found.
- o Iteratively increase the total cost of all the cells which are children to the adjacent nodes.
- o Also make them available (by updating valid list) for reconsideration while comparing total costs.





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State lattices

• Spatiotemporal state lattices for fast trajectory planning in dynamic on-road driving scenarios

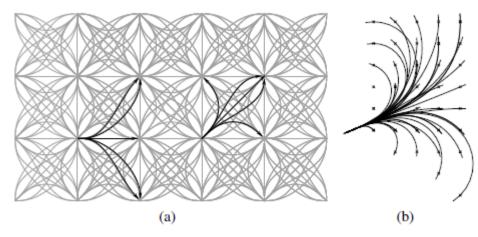


Fig. 1: (a): A nontemporal state lattice on a 2D workspace. Part of the canonical control set is displayed in black. (b): The right hand side shows part of a possible control set at finer discretisation, where the configuration space includes curvature.

2. Sample Based Planner

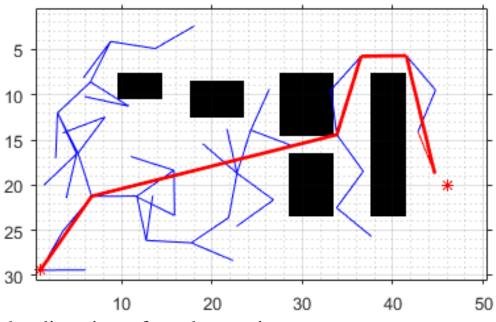
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• Rapidly-exploring random tree (RRT)

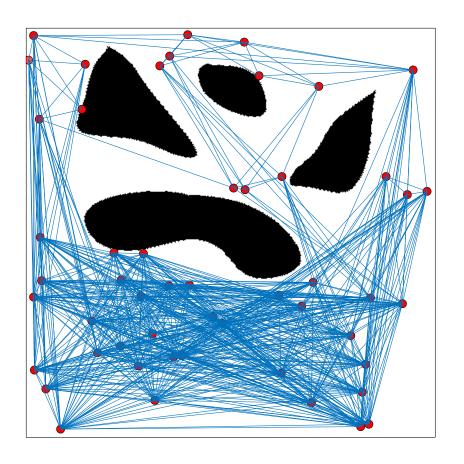
- Sampling-based Algorithms for Optimal Motion Planning
- In RRT we maintain two lists Edges and Vertices.
- Vertices contains
- o All valid nodes as we compute new vertex.
- o Vertices must lie in obstacle free space.
- Edges contains
- o Two vertices which constitutes the edge.
- o The edge must lie in obstacle free space.
- Implementation
- o Find a random point within the environment
- o Find a nearest vertex to that random point.
- o Find a new point at the delta distance away from nearest point in the direction of random point.
- o Check if the new point and edge attached to it lies in free space by breaking the edge in intermediate points and checking for each point alone.
- o Keep doing this until the cell with the target is not reached or the edge constitutes the target.

RRT (Rapidly-Exploring Random Trees)



2. Sample Based Planner

- Probabilistic roadmap method (PRM)
- Path Planning in Complex 3D Environments Using a Probabilistic Roadmap Method

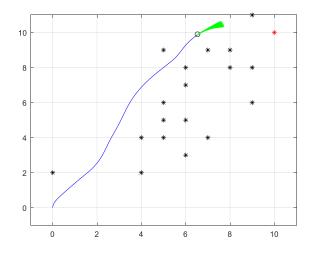


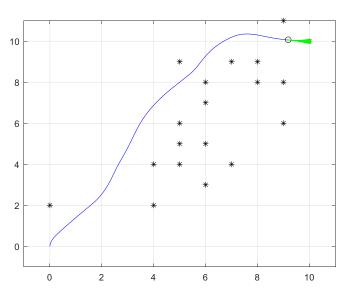
2. Sample Based Planner

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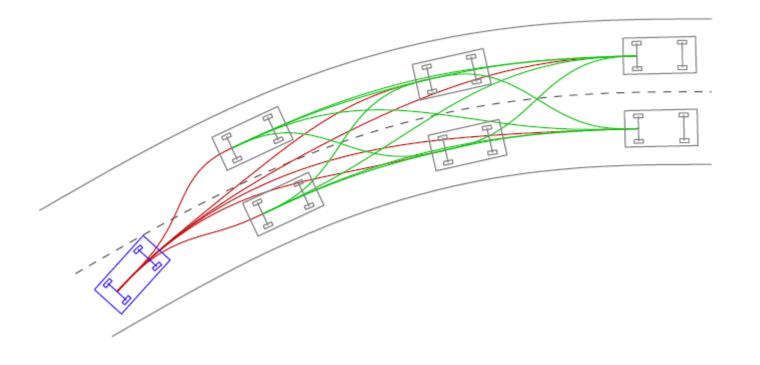
• Dynamic window approach (DWA)





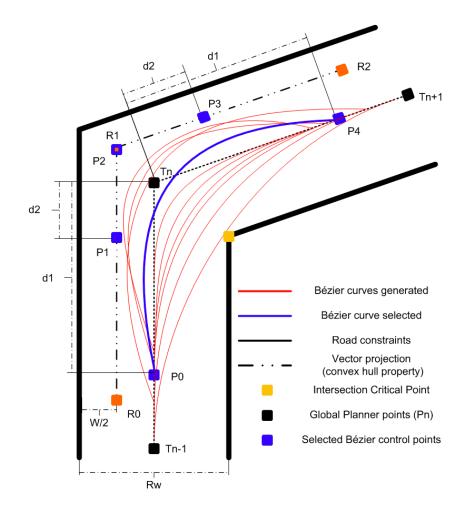
3. Interpolating Curve Planner

- Polynomial curves
- A Real-Time Motion Planner with Trajectory Optimization for Autonomous Vehicles



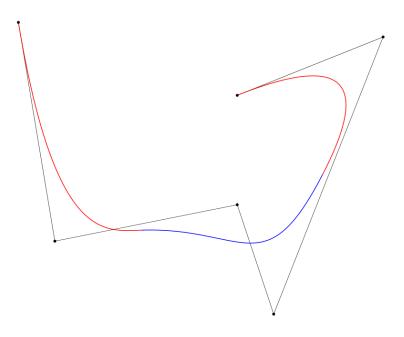
3. Interpolating Curve Planner

- Bezier curves
- Continuous curvature planning with obstacle avoidance capabilities in urban scenarios



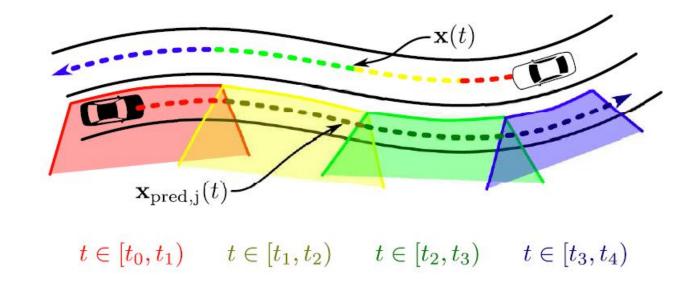
3. Interpolating Curve Planner

- Spline curves
- Kat-5: Robust systems for autonomous vehicle navigation in challenging and unknown terrain



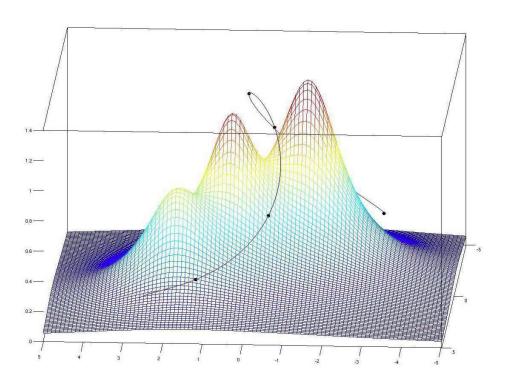
4. Numerical Optimization Approaches

- Function optimization
- Trajectory Planning for BERTHA a Local, Continuous Method



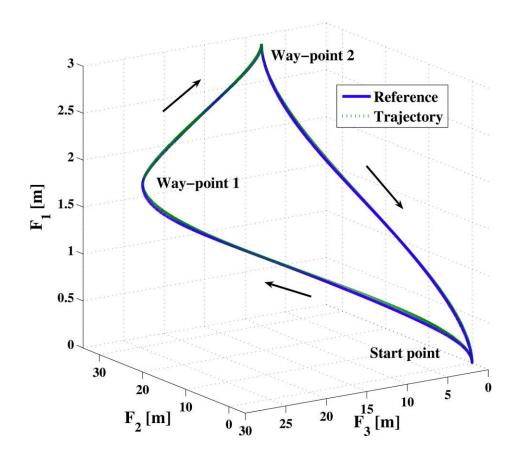
4. Numerical Optimization Approaches

- Binary linear programming
- 3D Path Planning in a Threat Environment



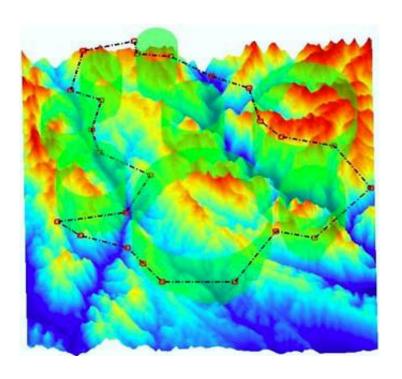
4. Numerical Optimization Approaches

- Non-linear programming
- Flatness-Based Trajectory Planning/Replanning for a Quadrotor Unmanned Aerial Vehicle



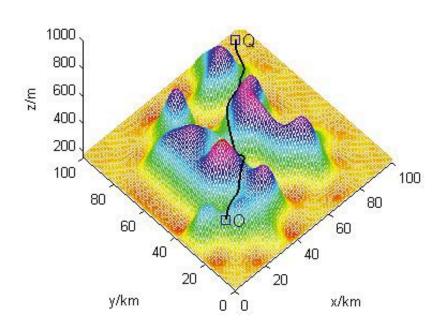
4. Computational Intelligent Methods

- Genetic algorithm(GA)
- Optimal Path Planning for UAVs Using Genetic Algorithm



4. Computational Intelligent Methods

- Particle swarm optimization (PSO)
- Three-Dimensional Path Planning for UAV Based on Improved PSO Algorithm



4. Computational Intelligent Methods

- Ant colony optimization (ACO)
- Path Planning for Indoor UAV Based on Ant Colony Optimization

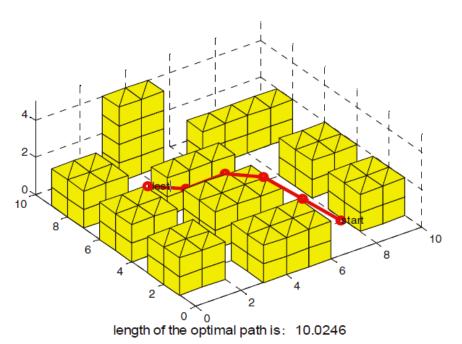
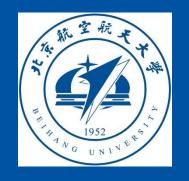


Fig 4. The optimal path in case 1



THANKS YOU!