

Obstacle Avoidance for Autonomous Racing using Reinforcement Learning



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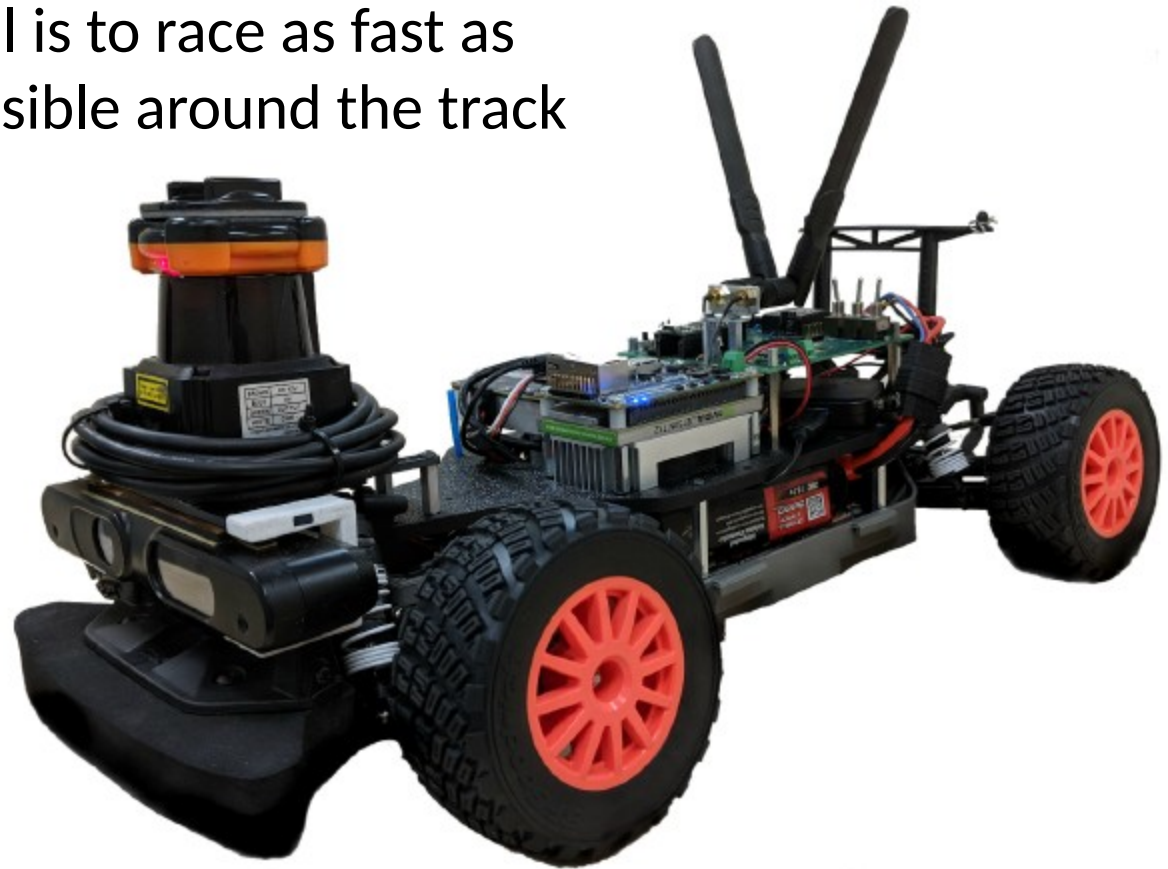
Introduction & Background



What is F1tenth?

F1tenth Racing

You have a car and the goal is to race as fast as possible around the track

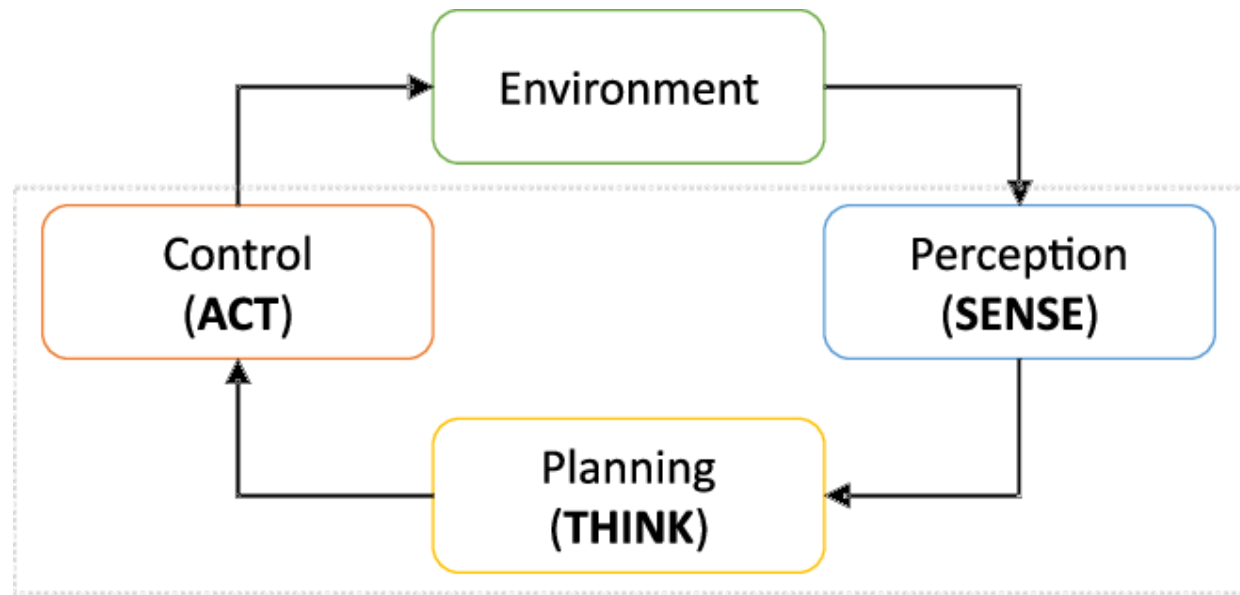


F1 Tenth Demo



F1tenth Components

- Usually split into three parts
 1. Perception: what is happening around you
 2. Planning: deciding what to do
 3. Control: executing your plan



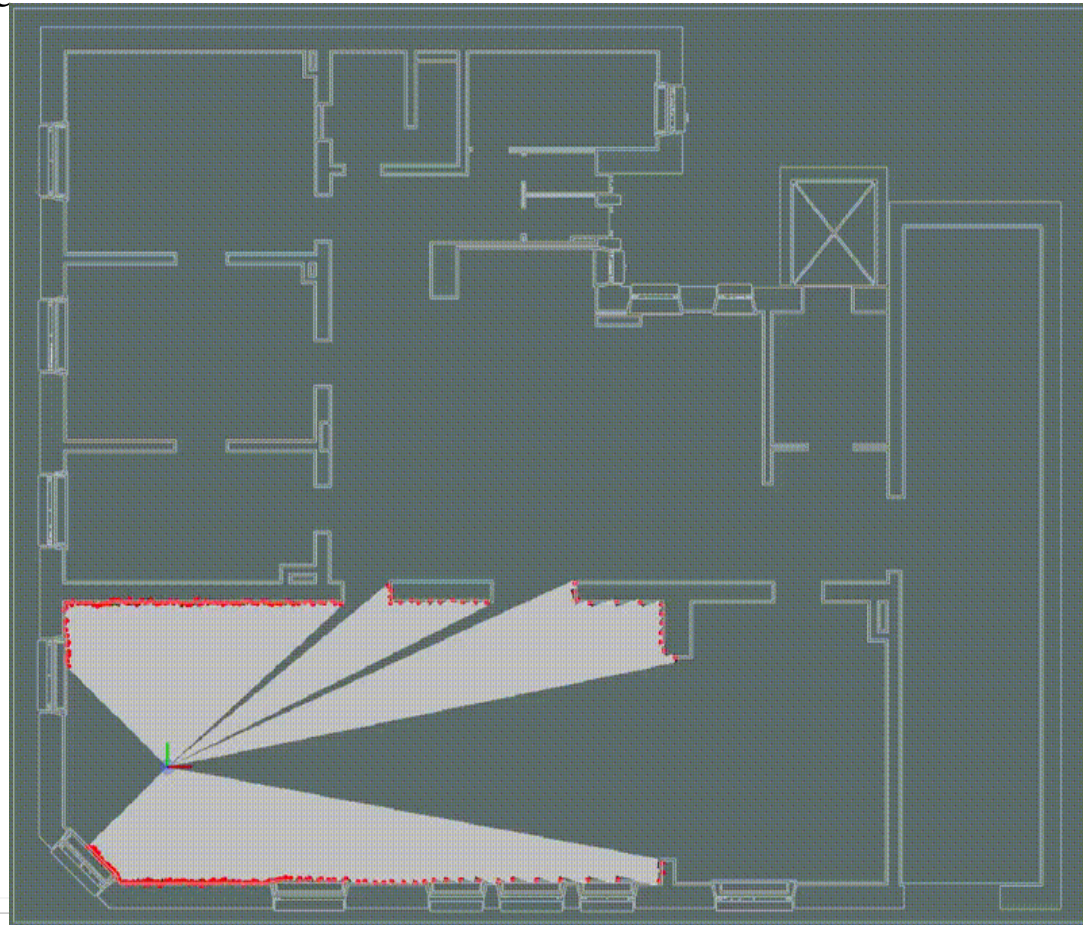
Planning to Win

- Planning is done in two stages
- Global planning
 - Done before the race
 - Takes in a map of the track
 - Produces low resolution waypoints through the course
 - Usually uses a path finder like A* with an optimization step
- Local planning
 - Done on a continuous basis through the race
 - Relies on an update obstacle map built with a SLAM algorithm
 - Uses the updated map and current state of vehicle to determine control references for the control system.
 - Output: velocity magnitude, heading direction.



SLAM

- As the vehicles race, they use SLAM algorithms to "simultaneously, localize and map" the environment.
 - Localize: determine their location.
 - Map: build a map of the environment around them.



Problems with SLAM

- Takes a long time
 - This is a problem in racing where you need your system to run fast
- Requires a lot of processing power
 - Sometimes have limited processing power
- Complex to implement
 - Requires advanced algorithms
- Struggles to map dynamic targets
 - It doesn't know which parts of the environment are moving and which are stationary.



Problem Definition

- Create a method of racing that can plan without explicitly maintaining an updated obstacle map using SLAM.
- Specifically, the system must be able to avoid unknown obstacles in the environment.
- NB: must be practically possible.
- Two step problem
 - Avoid static obstacles
 - Avoid dynamic obstacles





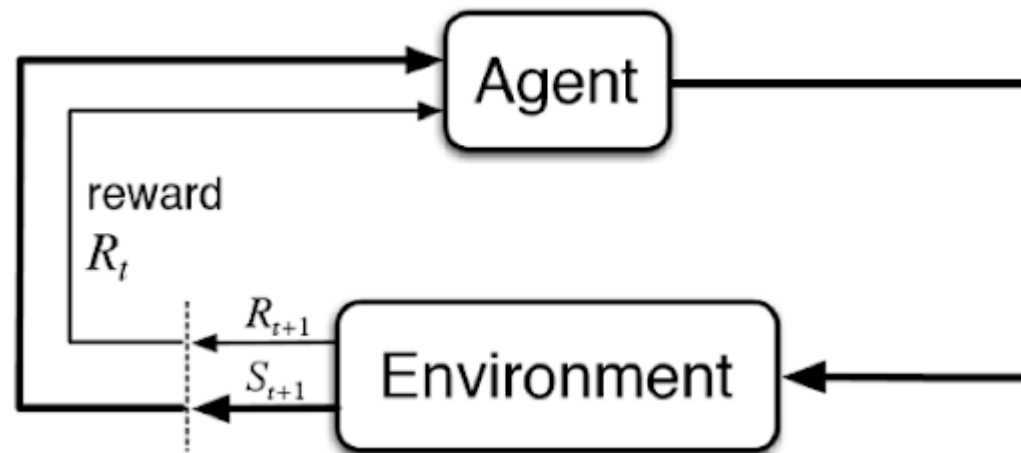
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Solutions



Solution Framework

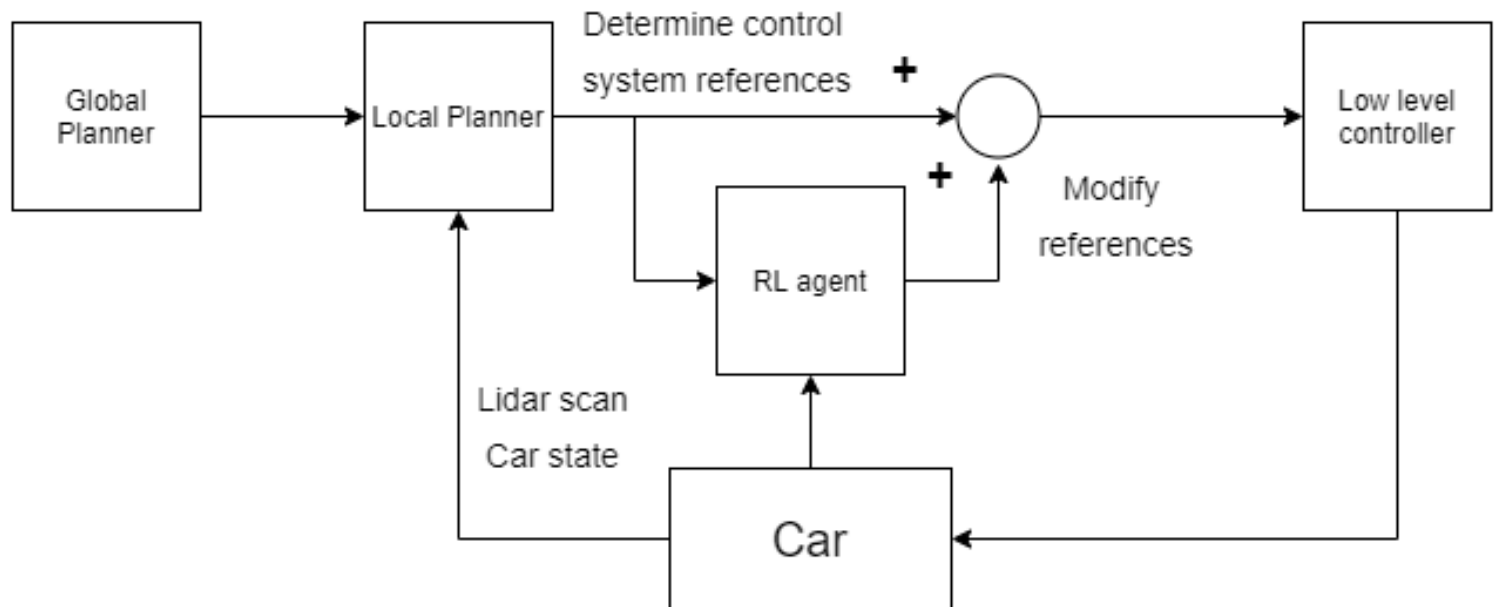
- Reinforcement learning is theoretically able to learn from experience to map sensor readings to actions.
- Generally RL uses neural networks to approximate the state space and determine the relevant actions.



Solution 1: RL Agent in Parallel

Solution 1:
Parallel agent

- We use the local planner with the known track map that assumes that there are no obstacles
- We then send all that as the state to the agent which can modify the local planner references.
 - Either the magnitude or heading



Training of Parallel Agent

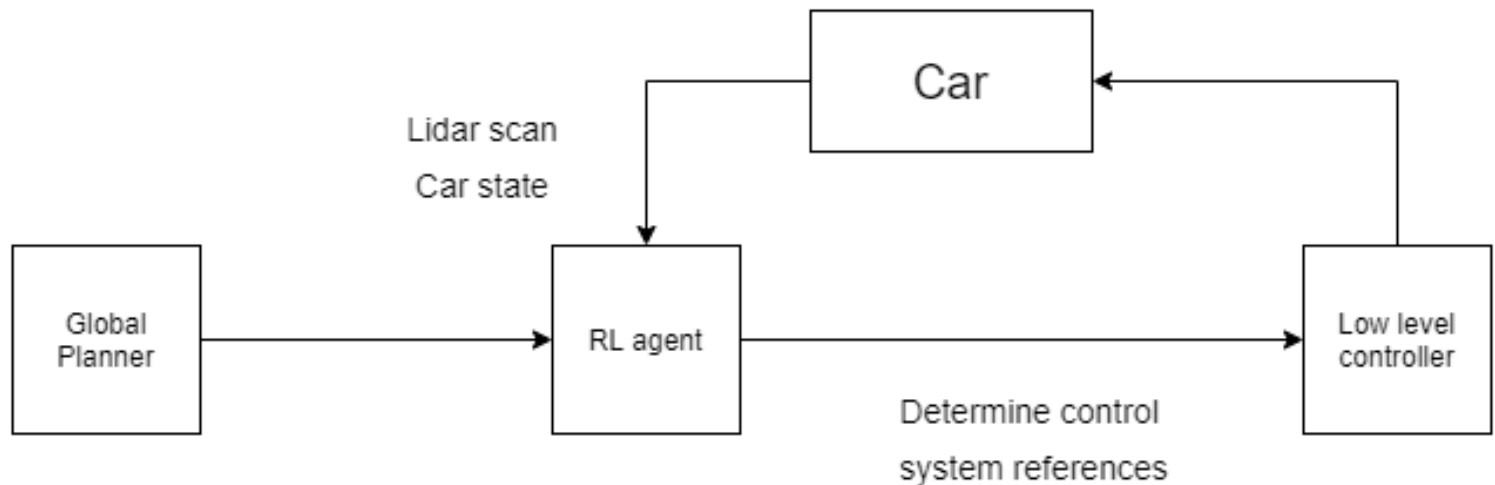
Solution 1:
Parallel agent

- Aim is to encourage it to swerve just enough from the local planner to not crash.
- Training environment
 - Custom built env of a room with obstacles
 - Train on an actual racetrack
- Reward (scaled to $[-1, 1]$):
 - -1 for crashing
 - $-\beta * \text{deviation from local planner}$
- Train in simulation because we can't afford to crash real cars.



Solution 2: Replacement Agent

- We replace the local planner with a RL agent
- The RL agent gets the lidar readings in and determines magnitude and velocity references.



Training of Replacement Agent

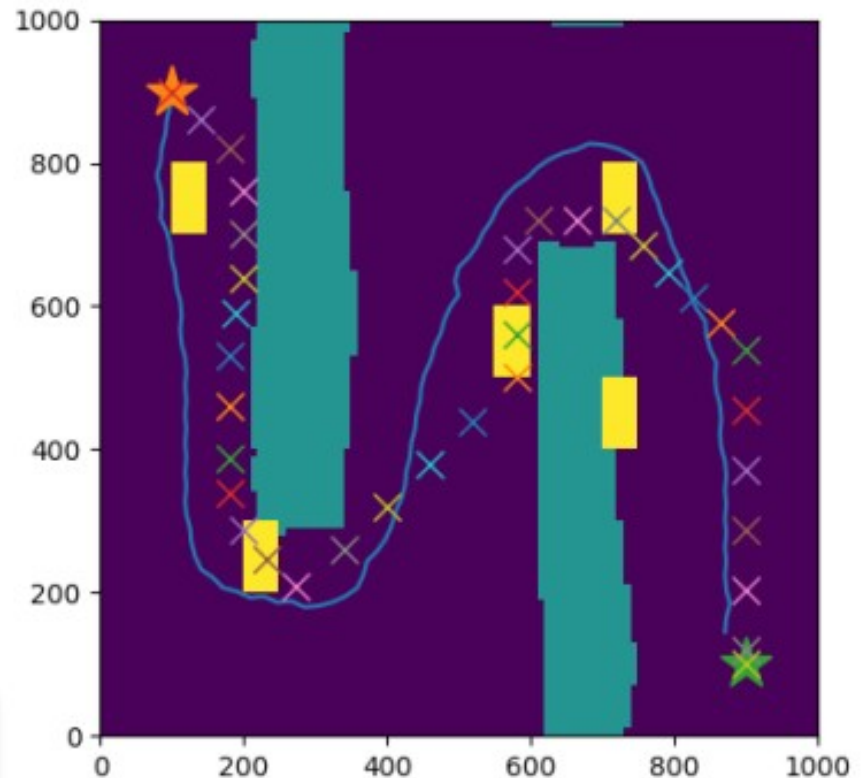
- Generate trajectories using map with known obstacles and optimal control.
- Use supervised (imitation) learning to learn to imitate the controller.
- State:
 - Relative global plan target
 - Current velocity
 - Lidar Scan
- Action space:
 - Continuous speed and heading references
- Train in simulation or in practice because we are following an optimal control strategy in training
 - Use additional external sensors that are able to supply a perfect map of the environment



Current Progress

Current progress

- Plan
 - Build solution architectures in a simple simulator
 - Upgrade simulation to use ROS and Gazebo
 - Run physical tests
- Current result
 - Replacement Architecture
 - Learning to avoid obstacles
- Next step
 - Improve simulation quality



The End

Questions?



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