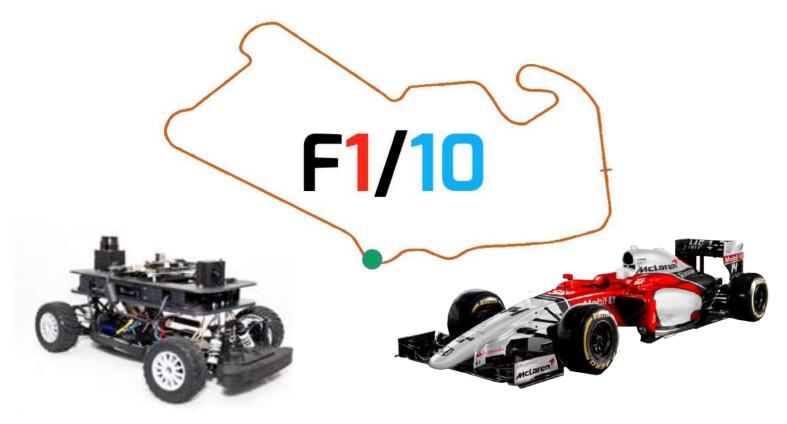
Autonomous Racing



Rahul Mangharam

So what is F100

?



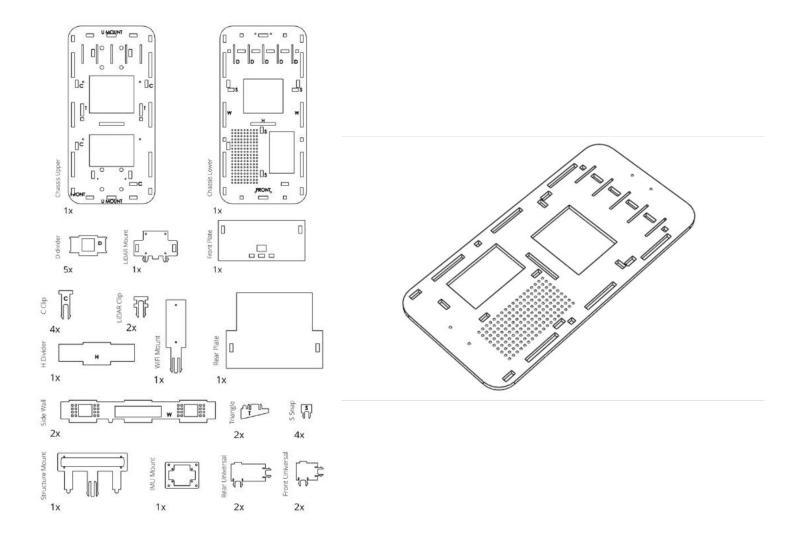


Easy to build in 3 hours.

F1/10 racing platform



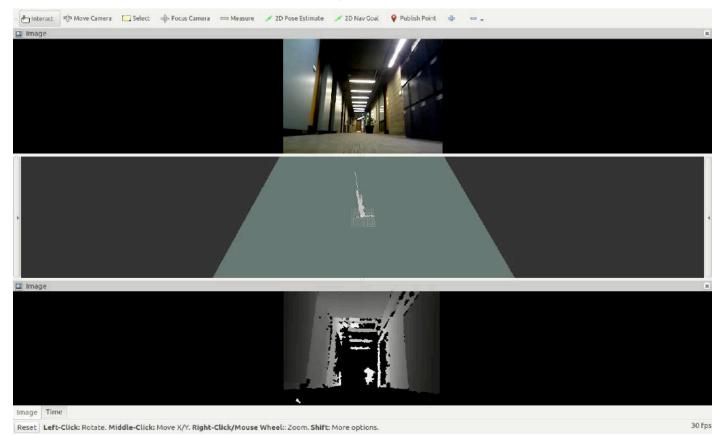
With simple IKEA style instructions at http://f1tenth.org



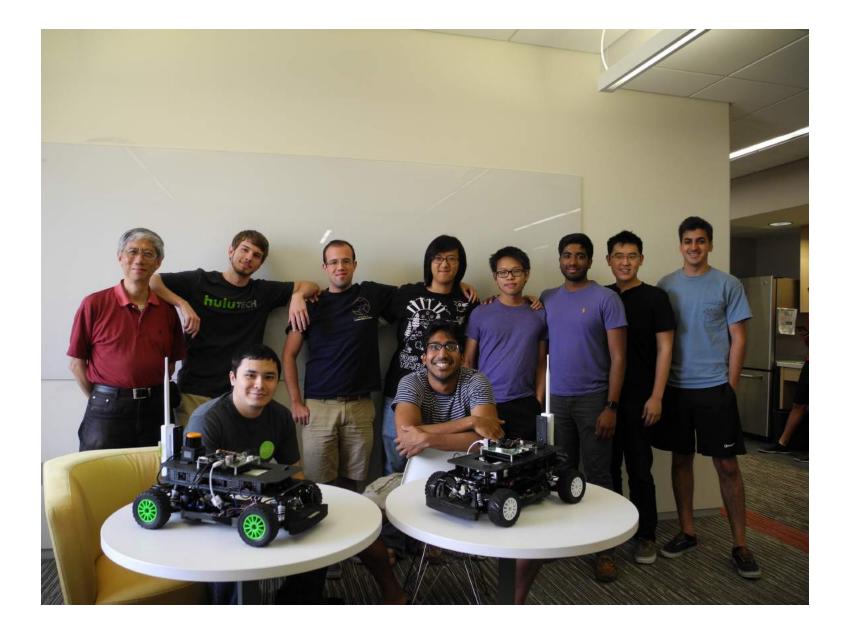
Learn

Perception. Planning. Control All lectures at http://f1tenth.org

Learn about Autonomy SLAM, AMCL, Scan matching, planning algorithms, optimal control strategies for the fastest autonomous driver, safe decision making..



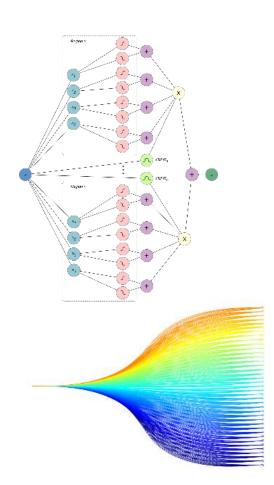
Note: Map based approach where loop closure fails...

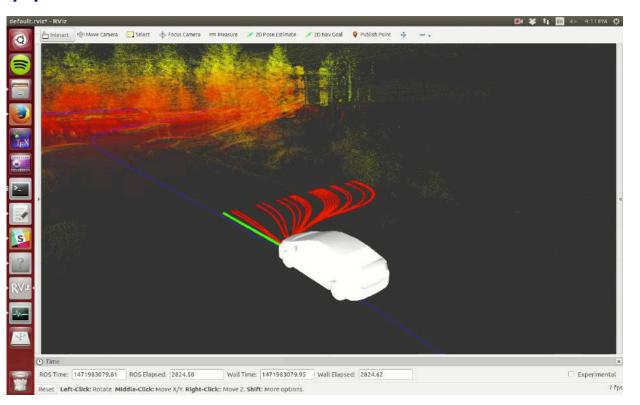


Research

Closing the loop fast enough.

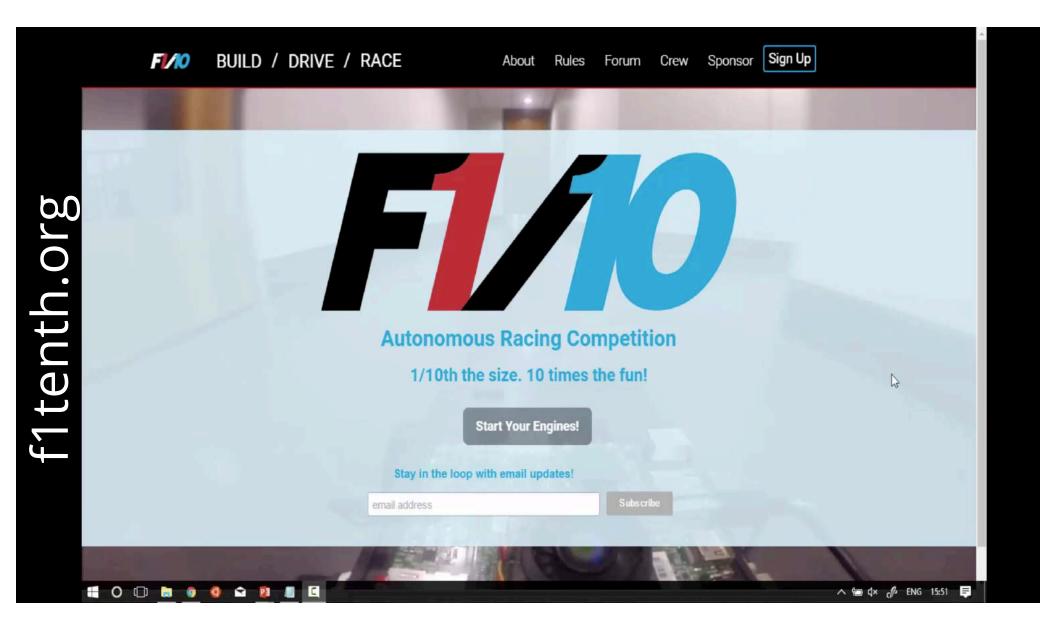
Map Free Approach



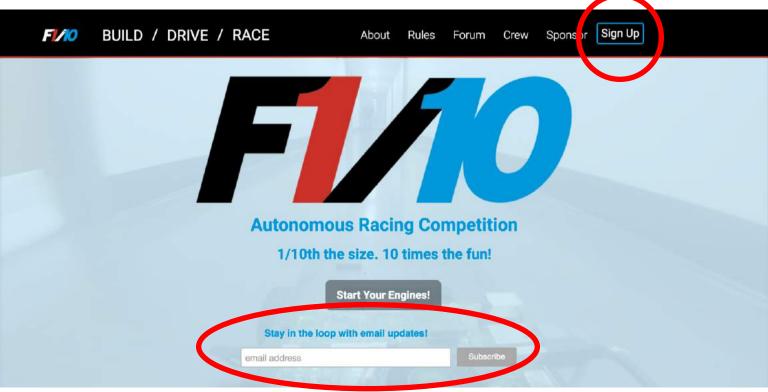


Efficient Adaptive State Lattice Planning Use linear optimization to learn weights for a network of radial basis functions. Quickly compute a variety of trajectories in the configuration space of the robot in order to create local plans...

Open Source



f1tenth.org

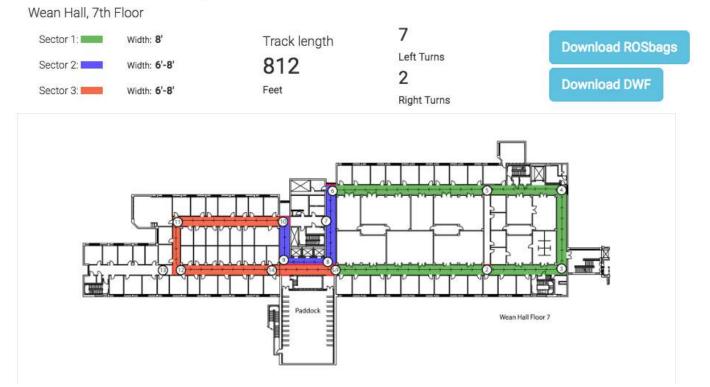


Get Involved !

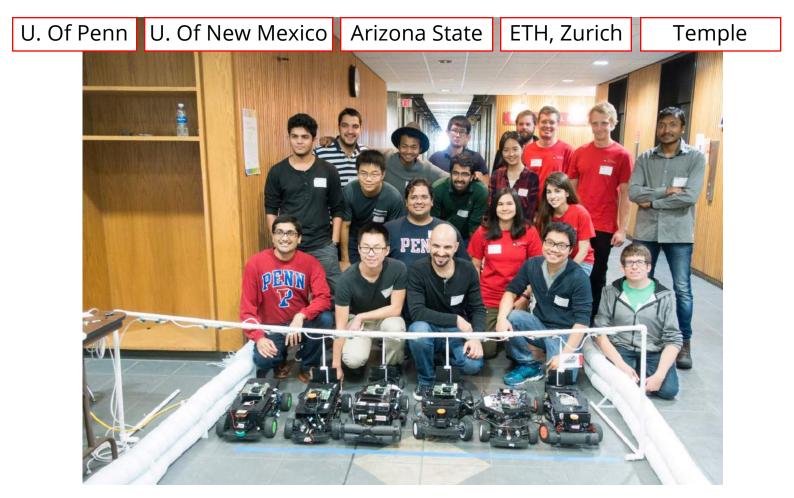
Compete



ES Week Track Layout



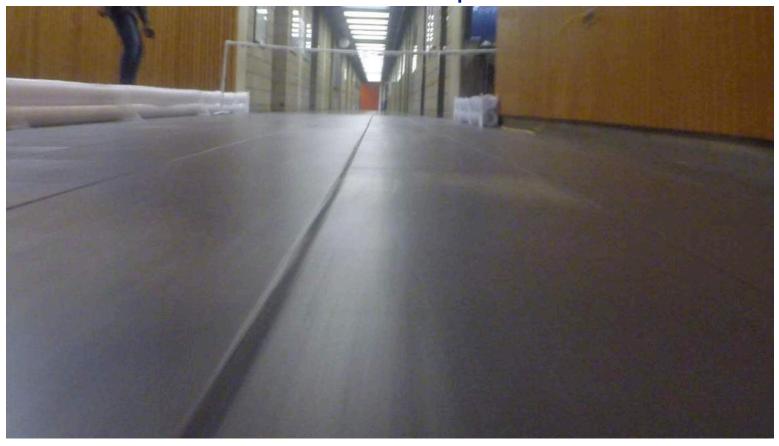
1st F1/10 Autonomous Racing Competition [Oct 1-2, 2016, CMU, Pittsburgh]



1st F1/10 Autonomous Racing Competition [Oct 1-2, 2016, CMU, Pittsburgh]



Fastest Lap



812 ft 63.6 secs Avg 8.7mph Top ~16mph



Cyber Physical Systems (CPS) Week 2017 April 18-21, Pittsburgh, PA

1st F1/10 Autonomous Racing Competition [Oct 1-2, 2016, CMU, Pittsburgh]



Fun!

1/10th the scale. 10 times the fun!









2016 Formula One United States GP, October, Austin, TX

Accessible Autonomy

Expensive, power hungry, large...



Cheap, low power, small...

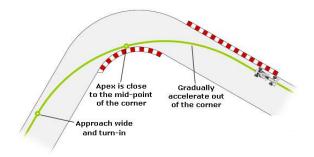


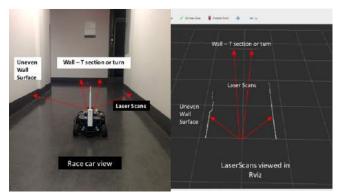
Same dynamics (different parameters), similar algorithmic challenges...

Build. Drive. Race.

Perception. Planning. Control

Making sense of the surroundings



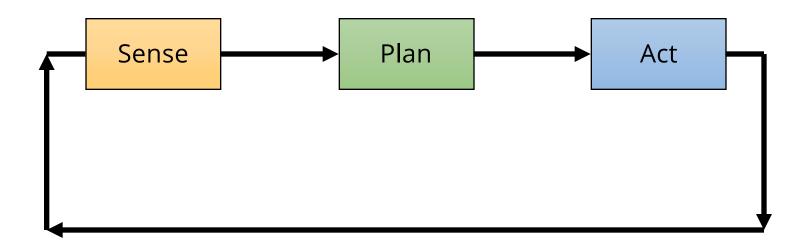


Planning the fastest racing line

Controlling the car to follow the planned path

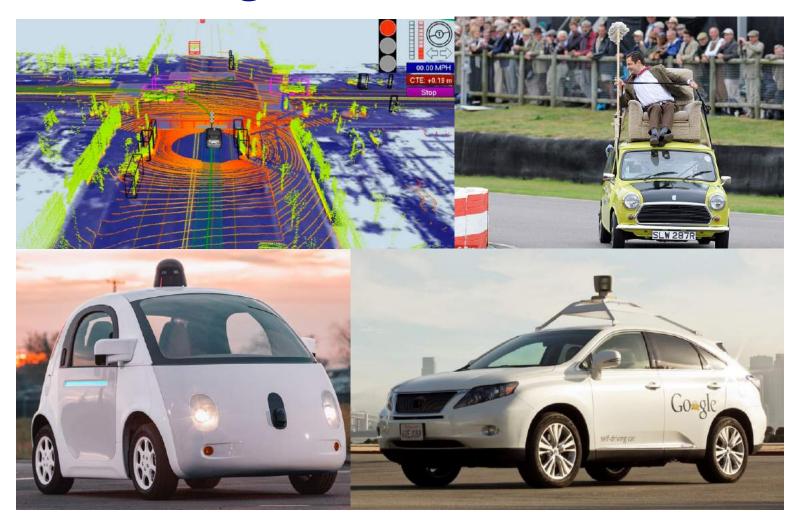


Closing the loop



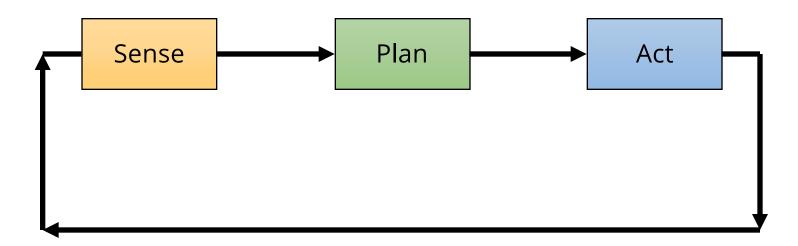


Self driving cars: Safety and Comfort

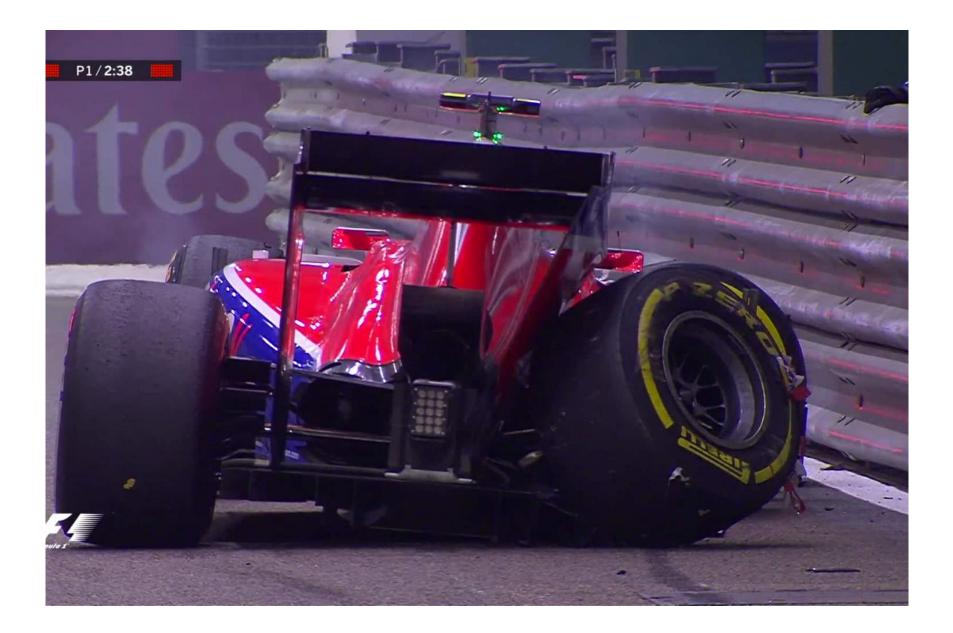




Closing the loop *fast*

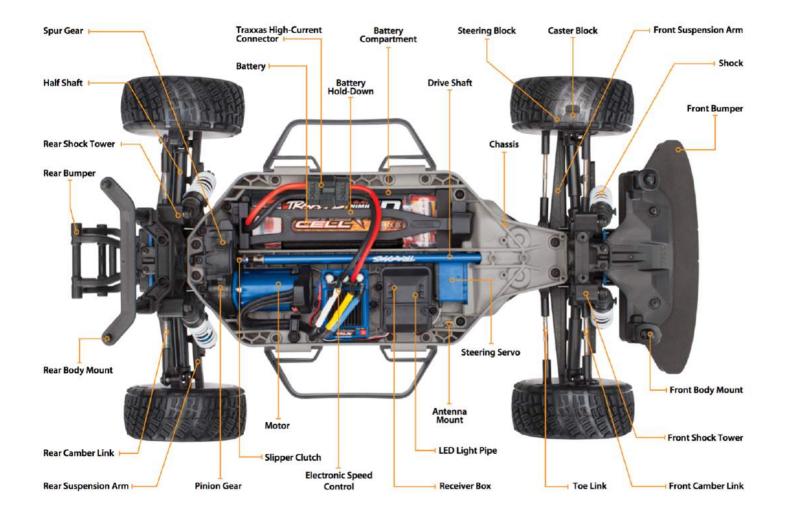




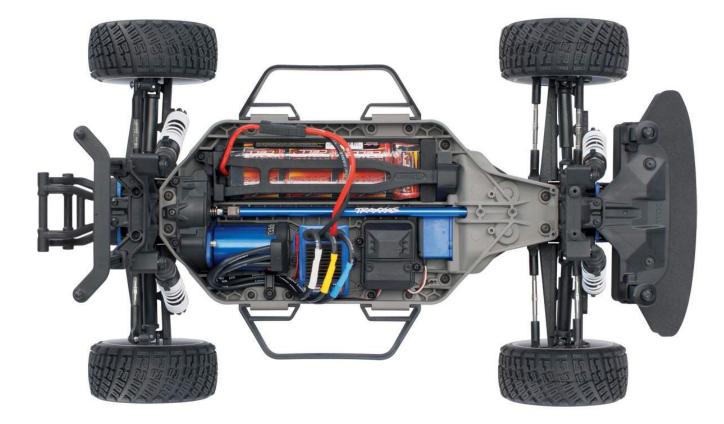


Build.

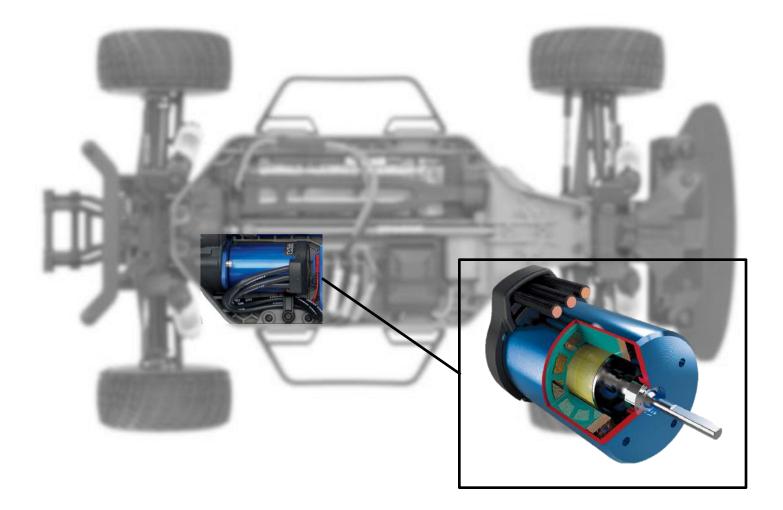
Traxxas 1/10 scale RC race car



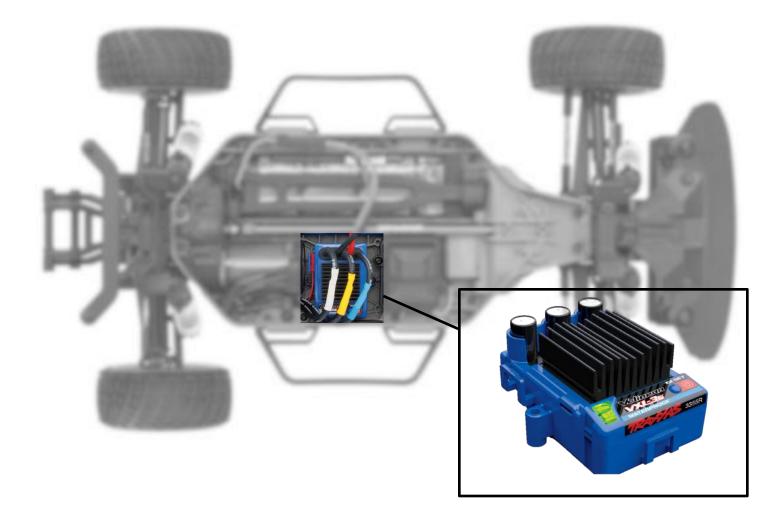
Traxxas 1/10 scale RC race car



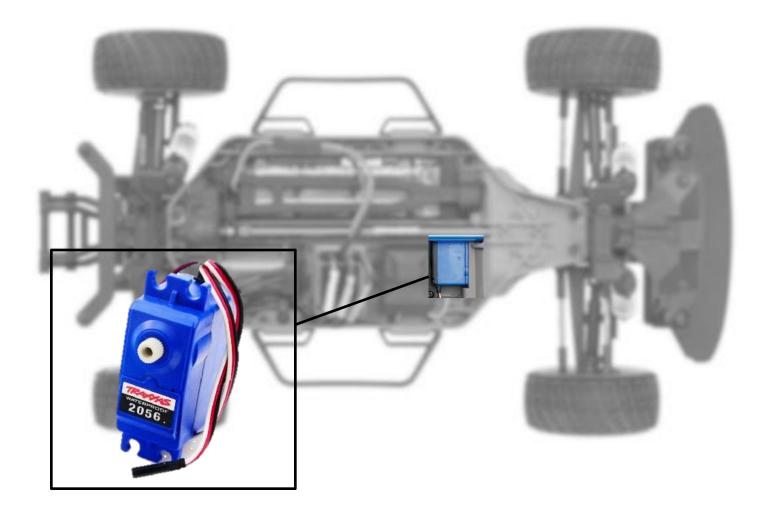
Brushless DC motor



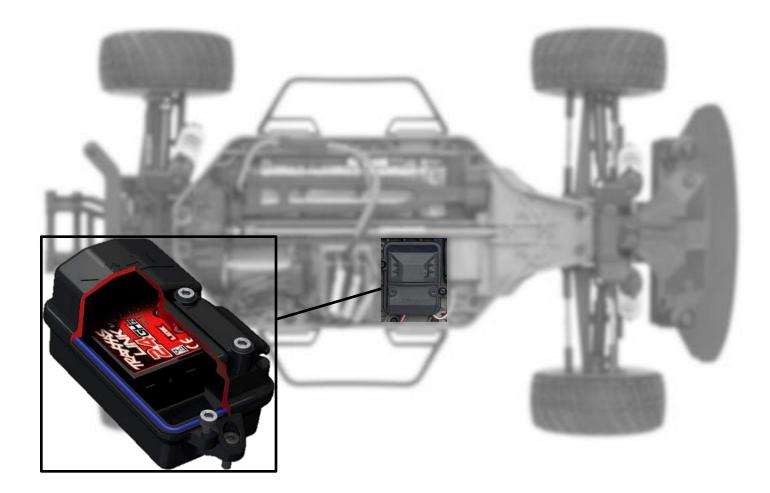
Electronic Speed Control (ESC)



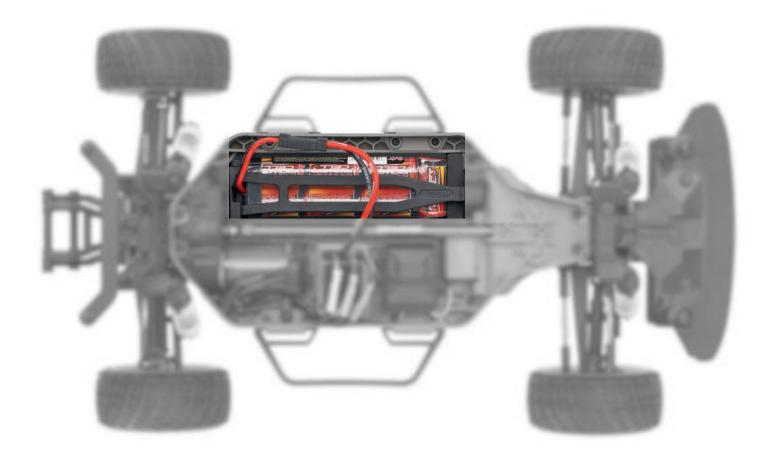
Servo motor for steering



Radio receiver



Battery pack



Sensor Integration









Camera

IMU





IR Depth Cameras



Sensor Integration











Lidar

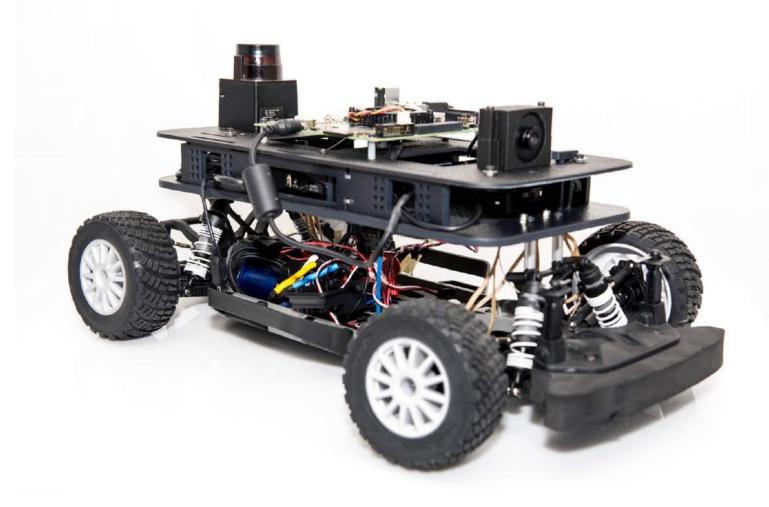
Camera

IMU

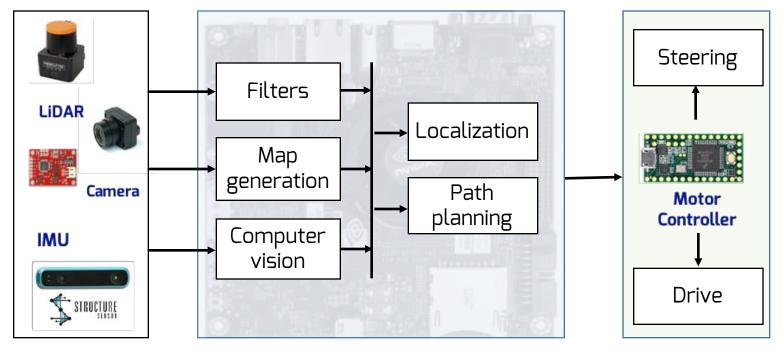




Ready to race ! Not quite..



System Architecture

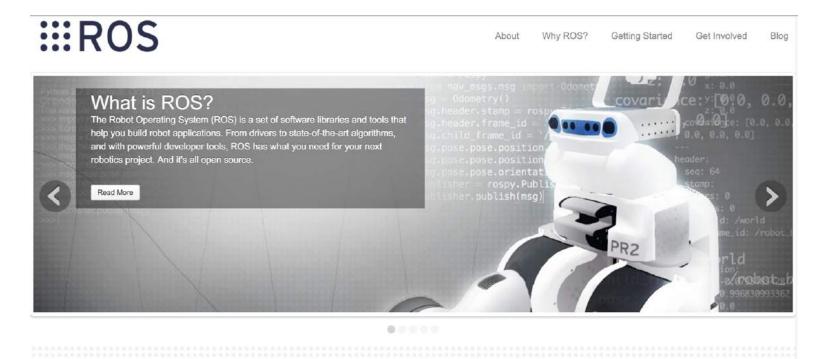


Perception

Planning

Control

ROS: Robot Operating System

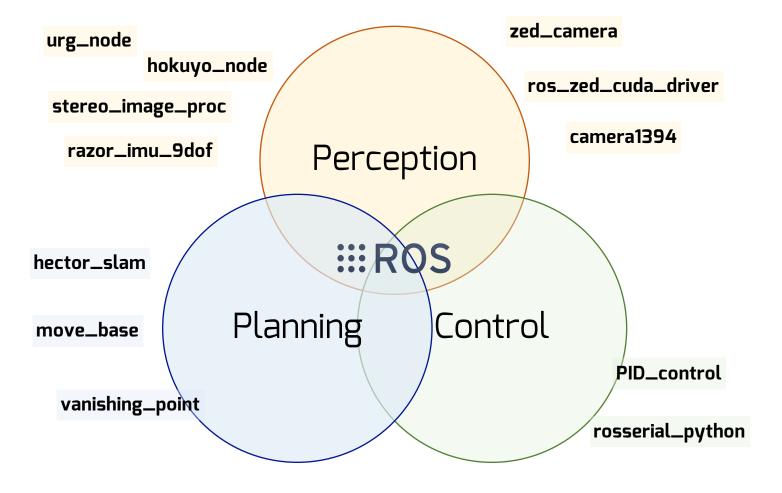


III ROS.org



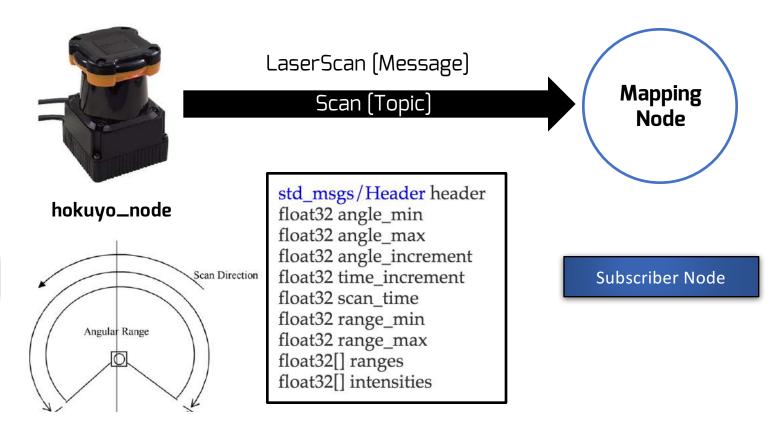
Open Source Robotics Foundation

ROS Capabilities



ROS: Messages

Messages are the strongly-typed **data structure** for a topic.



Basic ROS commands: roscore

roscore is the first thing that you should run when starting ROS.

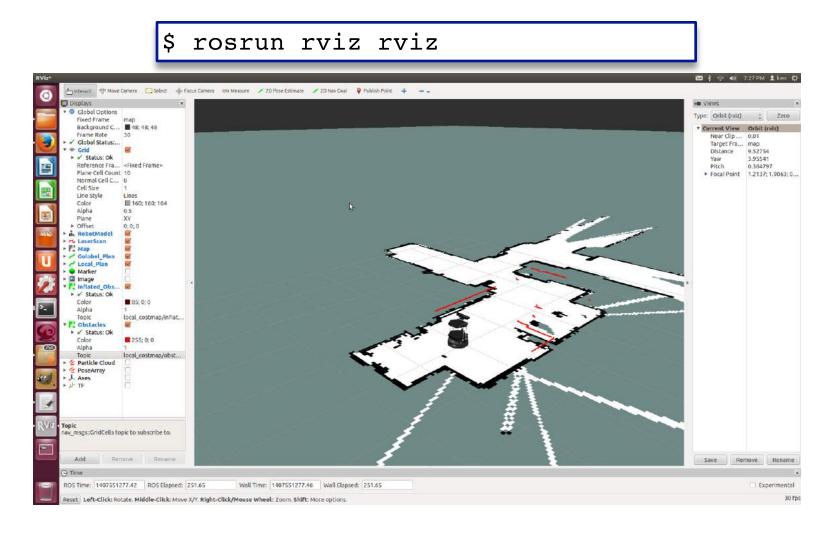
\$ roscore

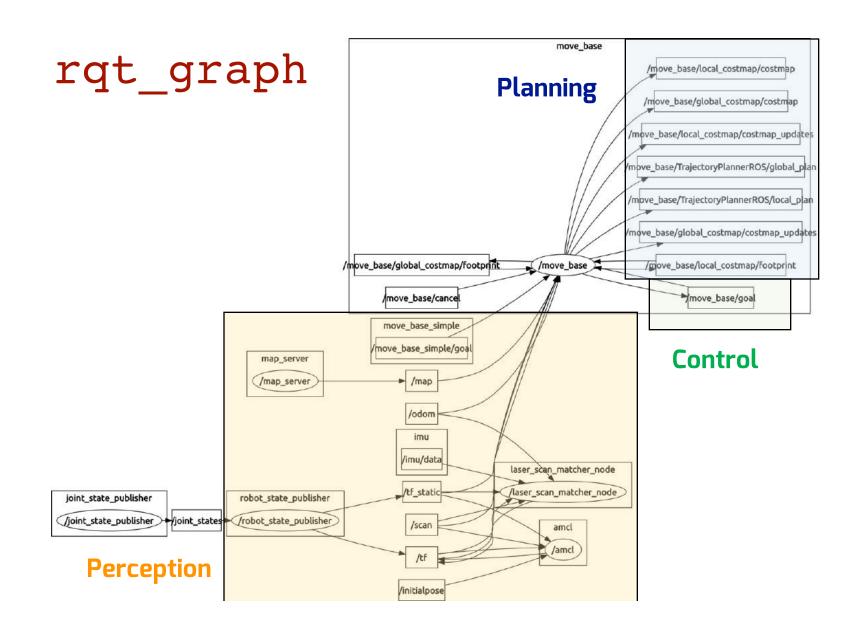
Collection of nodes and programs that are pre-requisites of a ROS-based system.

It starts up:

- The ROS Master
- A rosout logging node.

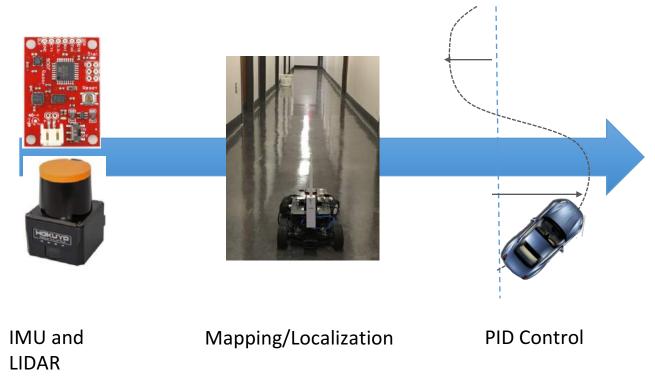
3D visualization tool: rviz





Drive

Perception. Planning. Control



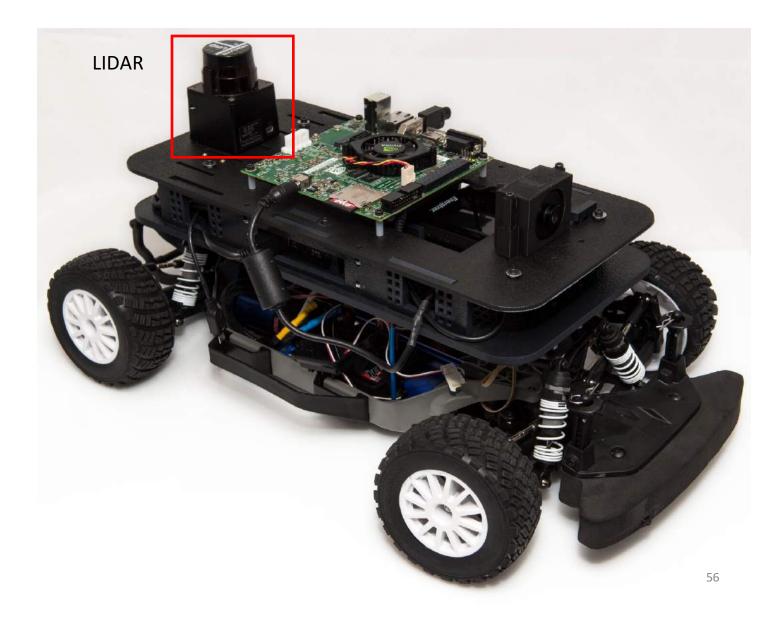


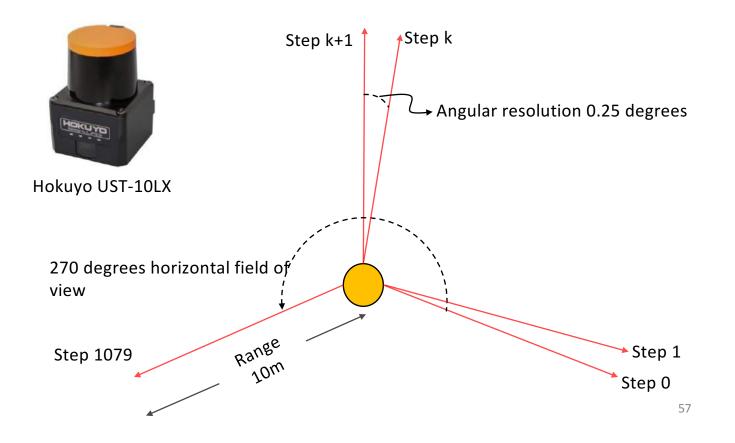
Sensors: specifications, placement and data

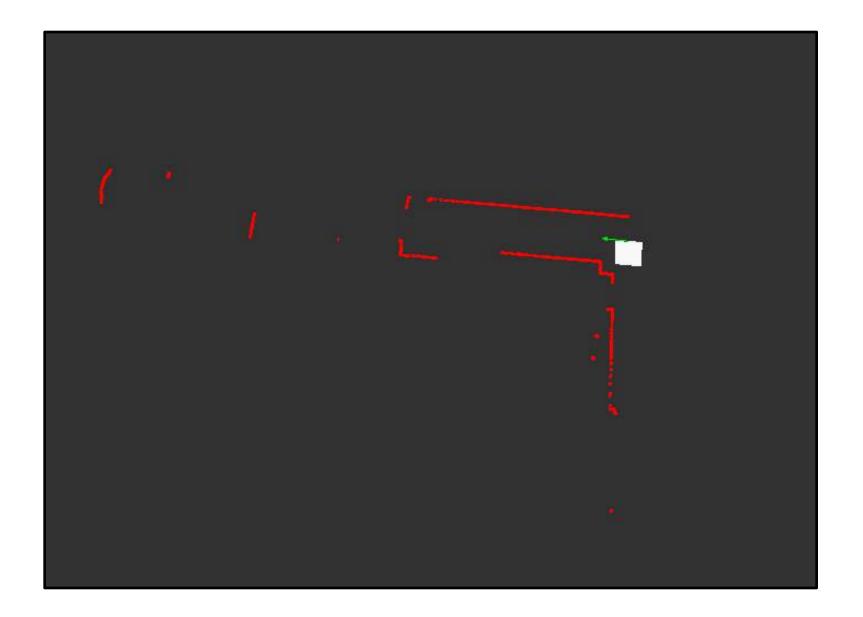
LIDAR



Depth information (how far are objects and obstacles around me)



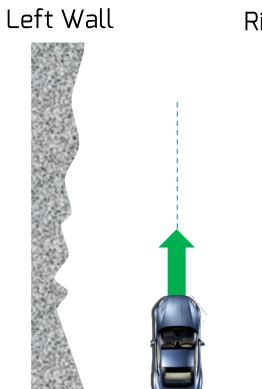




Localization

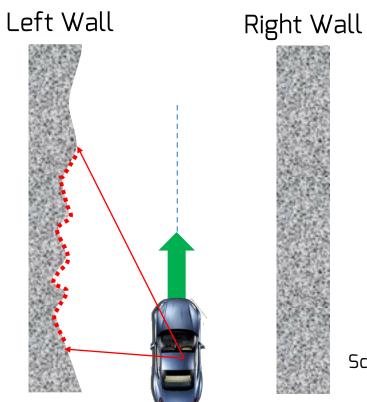
Know (where is) thyself:

Scan matching



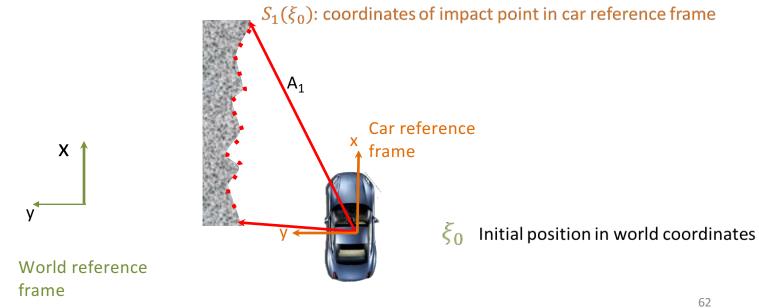


Scan matching

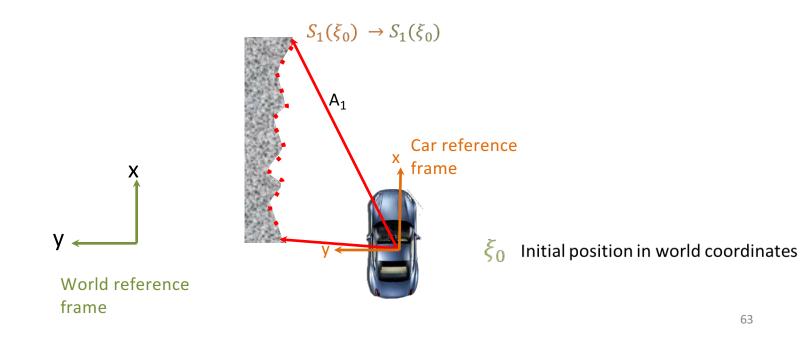


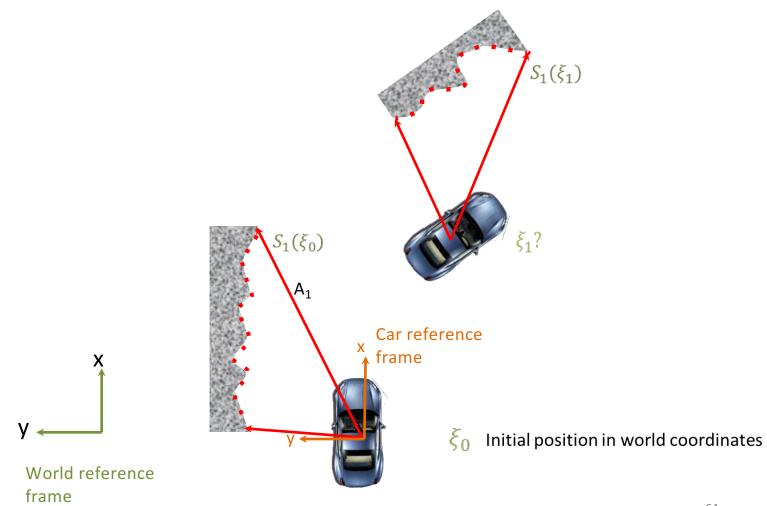


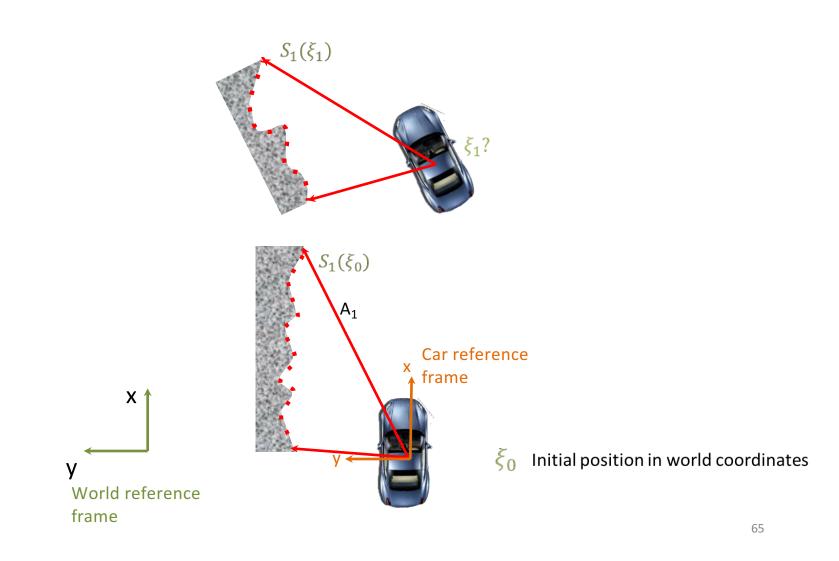
Scan 1

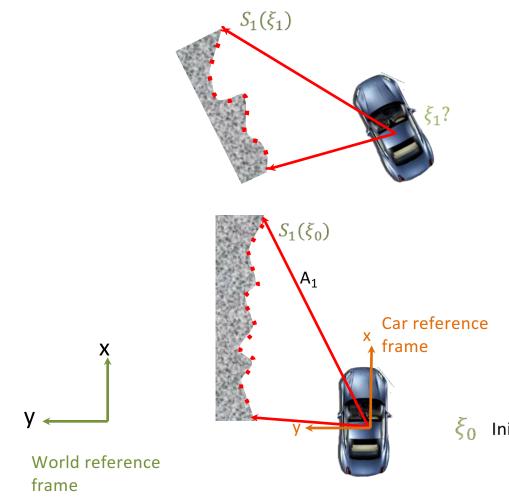


Scan 1



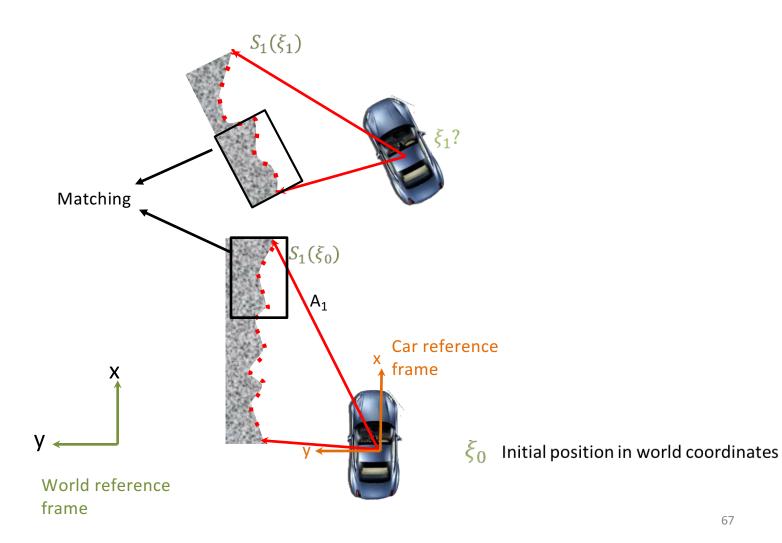


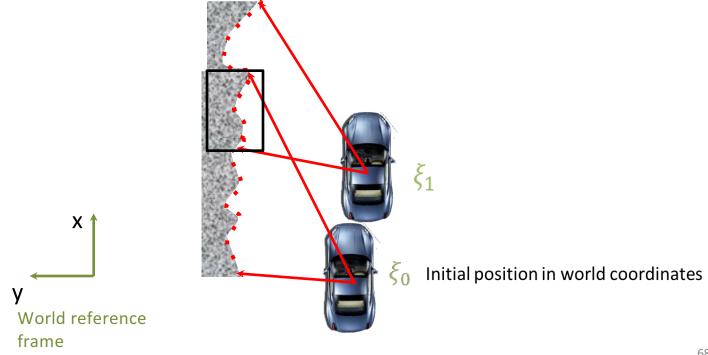




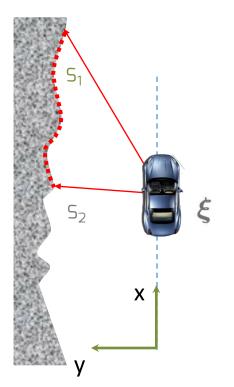
Premise: most likely car position at Scan 2 is the position that gives best overlap between the two scenes

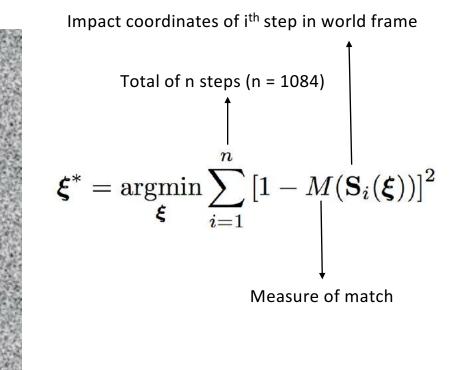
 ξ_0 Initial position in world coordinates



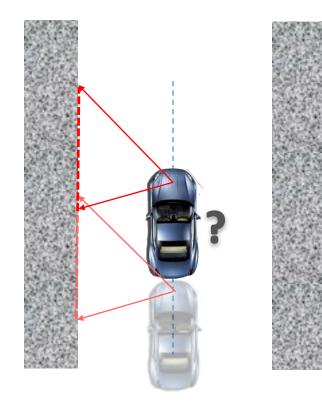


Scan matching: optimization





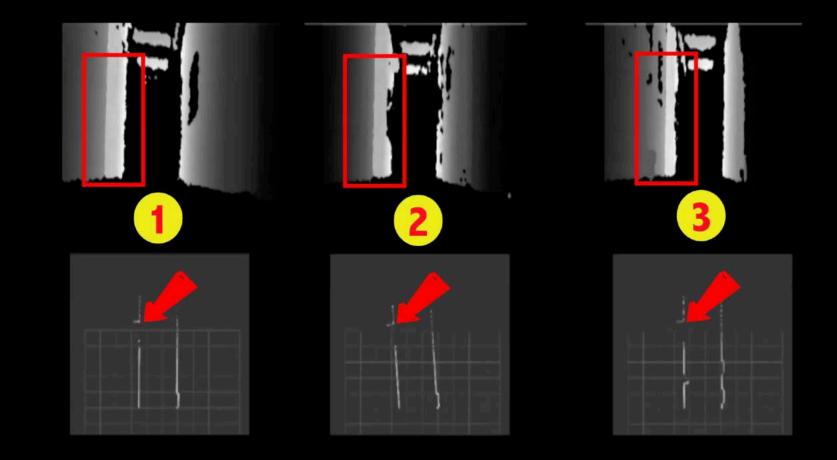
Scan matching : requirements



- Non-smooth, or heterogeneous, surfaces.
- Smooth surfaces all look the same to the matching algorithm.

Door or Turn ?

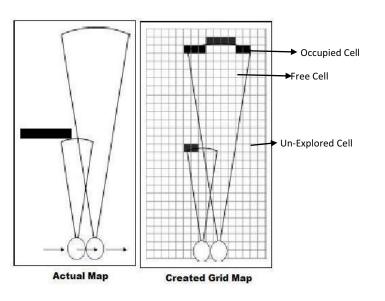




Mapping

Occupancy Grid Mapping

Measurement Model



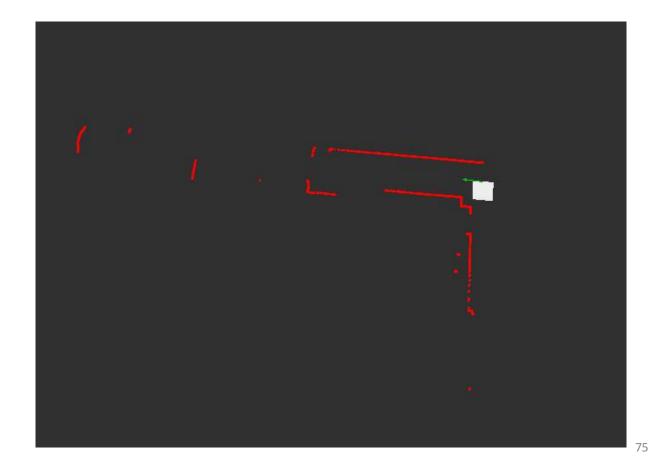
Measurement :	
m _{x,y} = 1	LiDAR hit
m _{x,y} = 0	No occlusion

Map Cell:
 Z = 1 Occupied
 Z = 0
 UnExplored
 Z = -1 Free

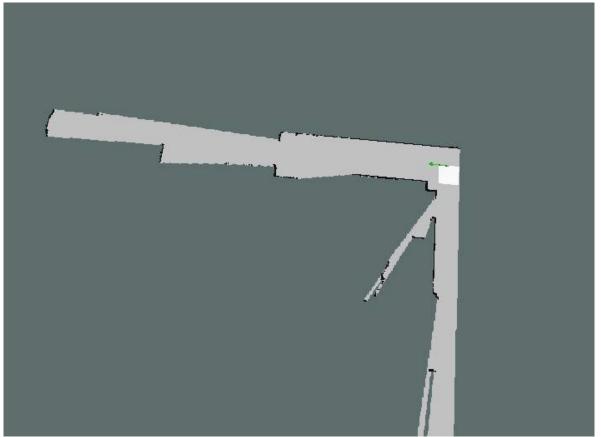
$$p(z|m_{x,y})$$

• Measurement Model :

Registering the first Scan



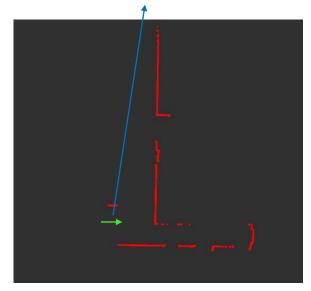
Registering the first Scan



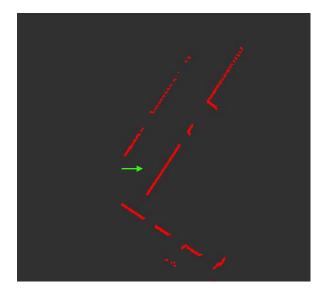
76

Scan matching

Pose of the Car at t = t1

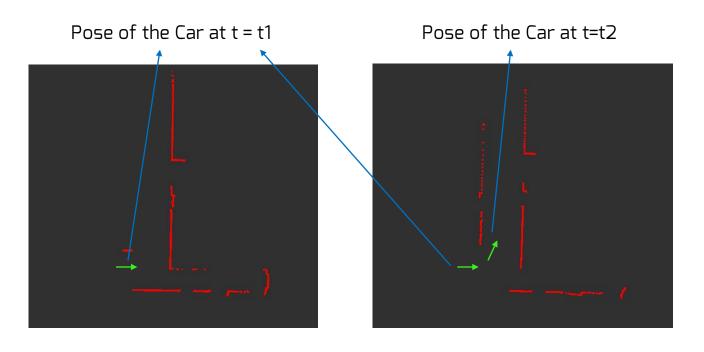


Laser Scans w.r.t car at Time t = t_1



Laser Scans w.r.t car at Time t = t_2

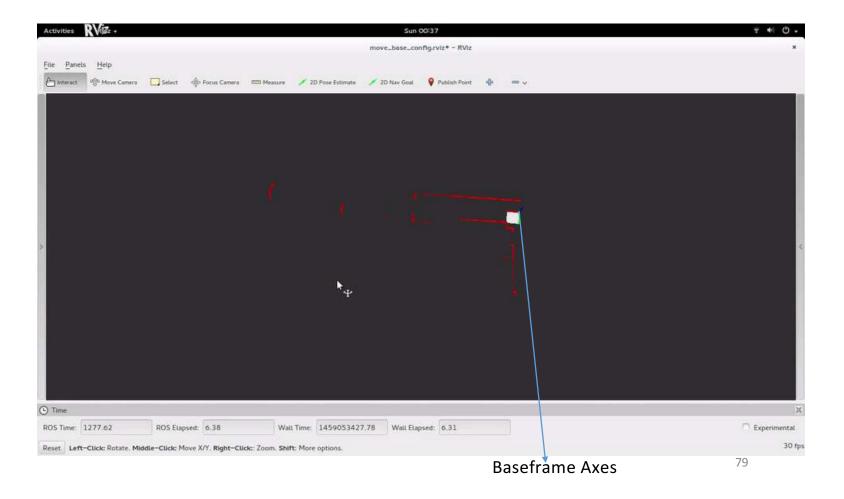
Scan matching



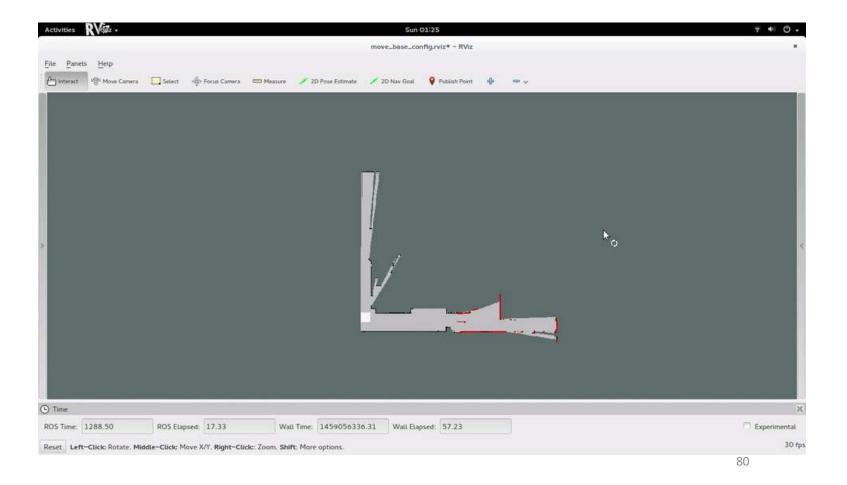
Laser Scans w.r.t car at Time t = t_1

Laser Scans w.r.t car at Time t = t_2

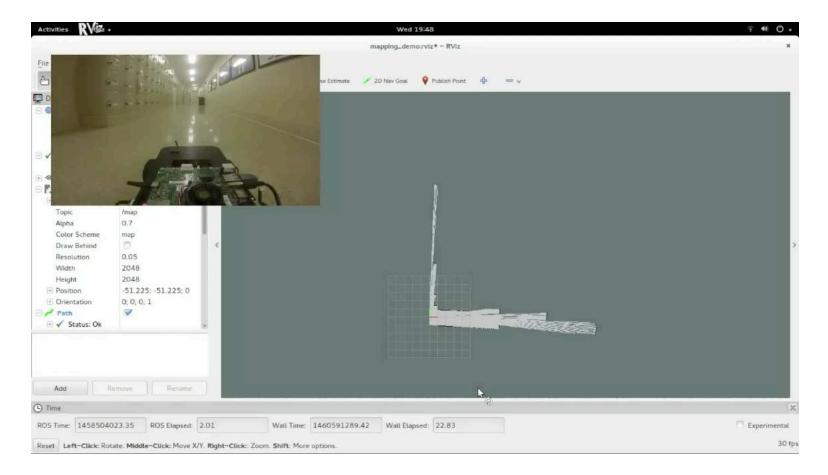
Raw LiDAR Scans



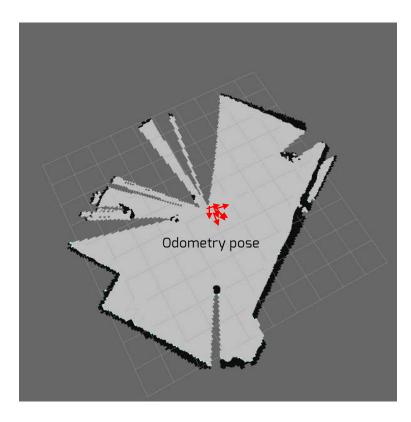
Map Update



Hector SLAM

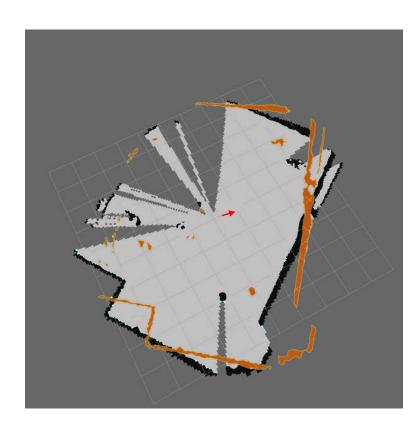


Particle Filter in 2D: Localization



$$W = \frac{\sum_{m} \sum_{n} (Amn - \overline{A})(Bmn - \overline{B})}{\sqrt{\left(\sum_{m} \sum_{n} (A_{mn} - \overline{A})^{2}\right)\left(\sum_{m} \sum_{n} (B_{mn} - \overline{B})^{2}\right)}}$$

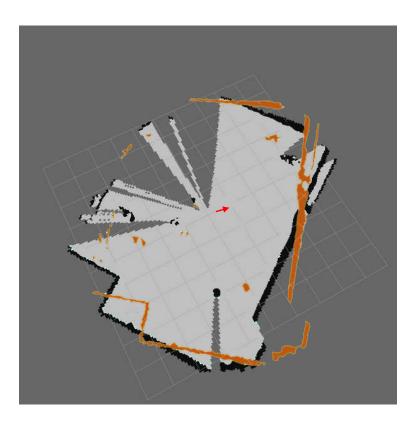
Particle Weight



$$W = \frac{\sum_{m} \sum_{n} (Amn - \overline{A})(Bmn - \overline{B})}{\sqrt{\left(\sum_{m} \sum_{n} (A_{mn} - \overline{A})^{2}\right)\left(\sum_{m} \sum_{n} (B_{mn} - \overline{B})^{2}\right)}}$$

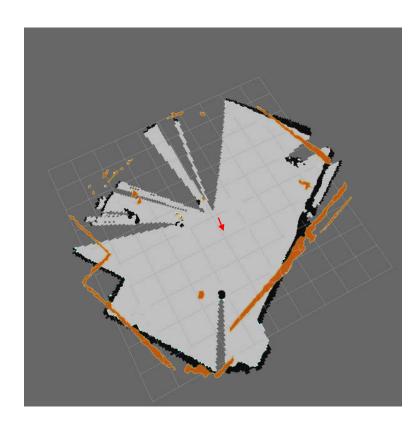
 Particle
 Weight

 Particle 1
 W1



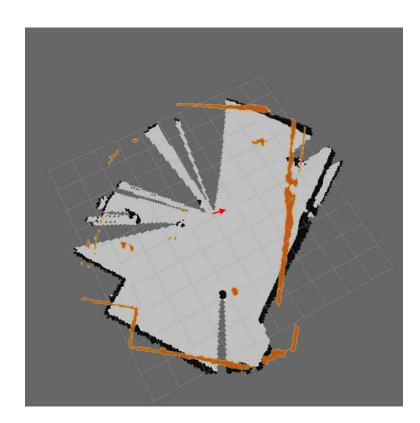
$$W = \frac{\sum_{m} \sum_{n} (Amn - \overline{A})(Bmn - \overline{B})}{\sqrt{\left(\sum_{m} \sum_{n} (A_{mn} - \overline{A})^{2}\right)\left(\sum_{m} \sum_{n} (B_{mn} - \overline{B})^{2}\right)}}$$

Particle	Weight
Particle 1	W_1
Particle 2	W ₂



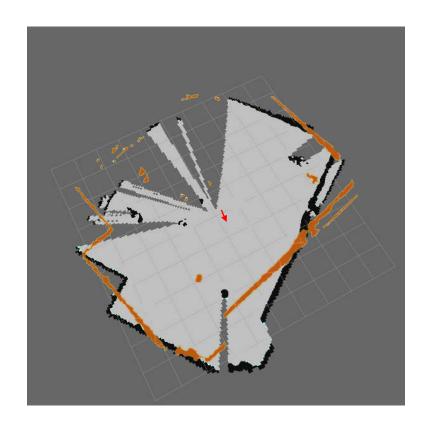
$$W = \frac{\sum_{m} \sum_{n} (Amn - \overline{A})(Bmn - \overline{B})}{\sqrt{\left(\sum_{m} \sum_{n} (A_{mn} - \overline{A})^{2}\right)\left(\sum_{m} \sum_{n} (B_{mn} - \overline{B})^{2}\right)}}$$

Particle	Weight
Particle 1	W_1
Particle 2	W_2
Particle 3	W ₃



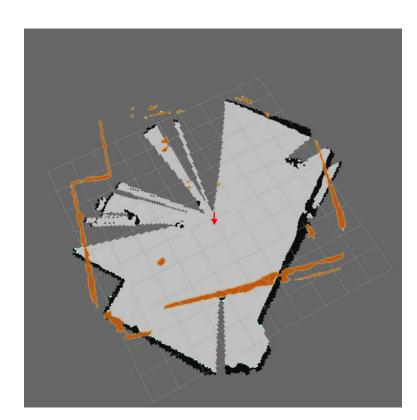
$$W = \frac{\sum_{m} \sum_{n} (Amn - \overline{A})(Bmn - \overline{B})}{\sqrt{\left(\sum_{m} \sum_{n} (A_{mn} - \overline{A})^{2}\right)\left(\sum_{m} \sum_{n} (B_{mn} - \overline{B})^{2}\right)}}$$

Particle	Weight
Particle 1	W_1
Particle 2	W_2
Particle 3	W ₃
Particle 4	W_4



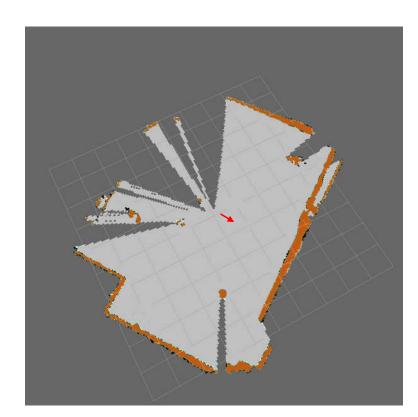
$$W = \frac{\sum_{m} \sum_{n} (Amn - \overline{A})(Bmn - \overline{B})}{\sqrt{\left(\sum_{m} \sum_{n} (A_{mn} - \overline{A})^{2}\right)\left(\sum_{m} \sum_{n} (B_{mn} - \overline{B})^{2}\right)}}$$

Particle	Weight
Particle 1	W_1
Particle 2	W_2
Particle 3	W ₃
Particle 4	W_4
Particle 5	W_5



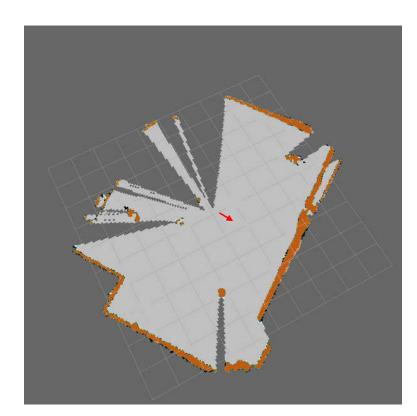
$$W = \frac{\sum_{m} \sum_{n} (Amn - \overline{A})(Bmn - \overline{B})}{\sqrt{\left(\sum_{m} \sum_{n} (A_{mn} - \overline{A})^{2}\right)\left(\sum_{m} \sum_{n} (B_{mn} - \overline{B})^{2}\right)}}$$

Particle	Weight
Particle 1	W_1
Particle 2	W_2
Particle 3	W ₃
Particle 4	W_4
Particle 5	W_5
Particle 6	W ₆



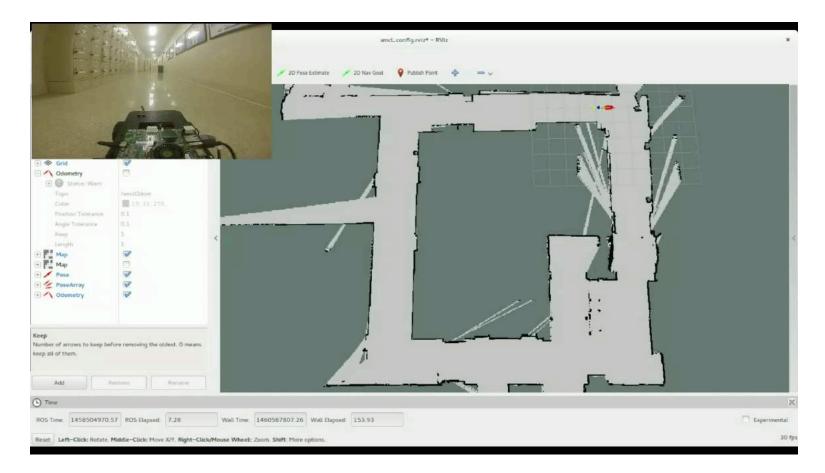
$$W = \frac{\sum_{m} \sum_{n} (Amn - \overline{A})(Bmn - \overline{B})}{\sqrt{\left(\sum_{m} \sum_{n} (A_{mn} - \overline{A})^{2}\right)\left(\sum_{m} \sum_{n} (B_{mn} - \overline{B})^{2}\right)}}$$

Weight
W_1
W_2
W ₃
W_4
W_5
W ₆



 $W_t \leftarrow W_{t-1_{\times}} W t$

AMCL- Adaptive Monte Carlo Localization



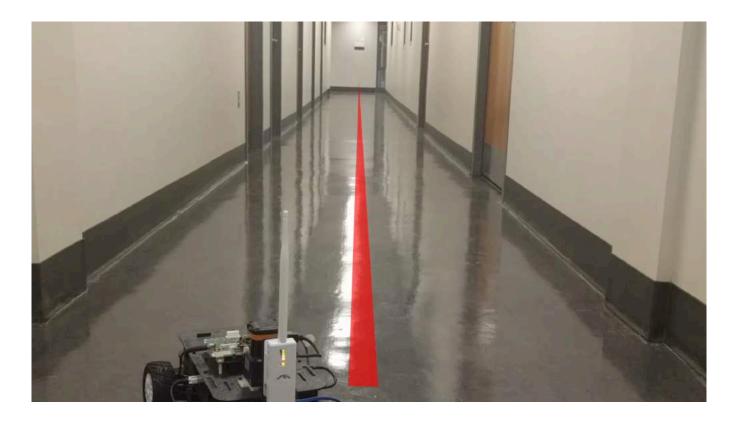
Control

Proportional, Integral, Derivative control

PID control: tuning the gains

- Default set of gains, determined empirically to work well for this car.
 - K_p = 14
 - K_i = 0
 - K_d = 0.09

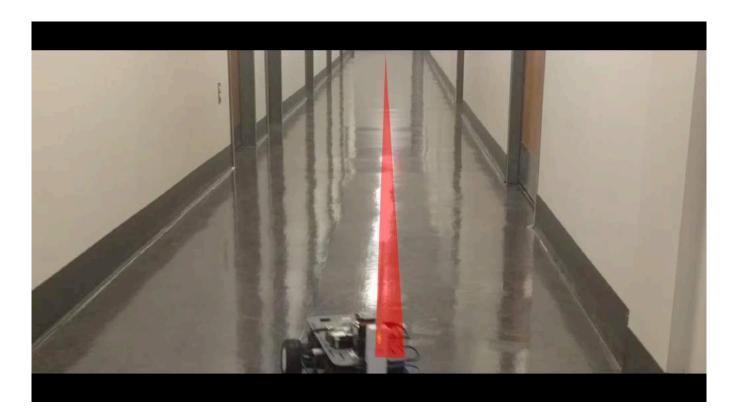
Nominal gains



PID control: tuning the gains

- Reduce $K_p \rightarrow$ less responsive to error magnitude
 - K_p = 5
 - K_i = 0
 - K_d = 0.09

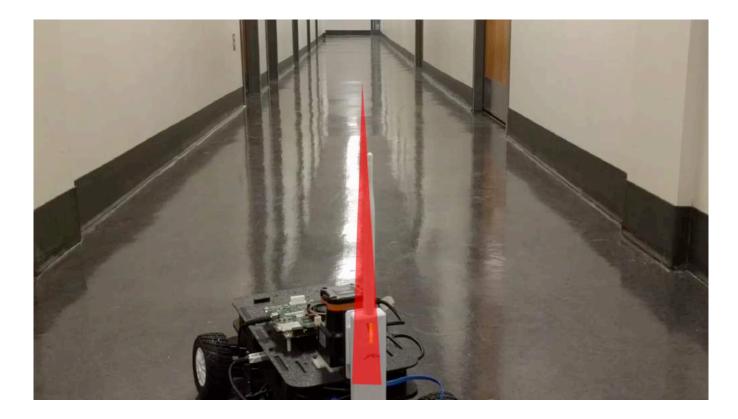
Smaller Kp



PID control: tuning the gains

- Include $K_i \rightarrow$ overly sensitive to accumulating error \rightarrow overcorrection
 - K_p = 14
 - K_i = 2
 - K_d = 0.09

Add integral control

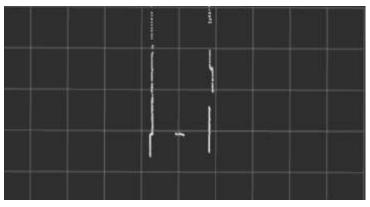


Race

Putting it all together..



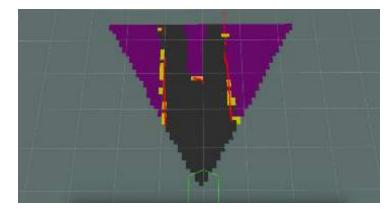
RGB Image used to ID dynamic obstacles



Laser scan 3D point cloud to 2D plane



Depth image provides 3D pointcloud

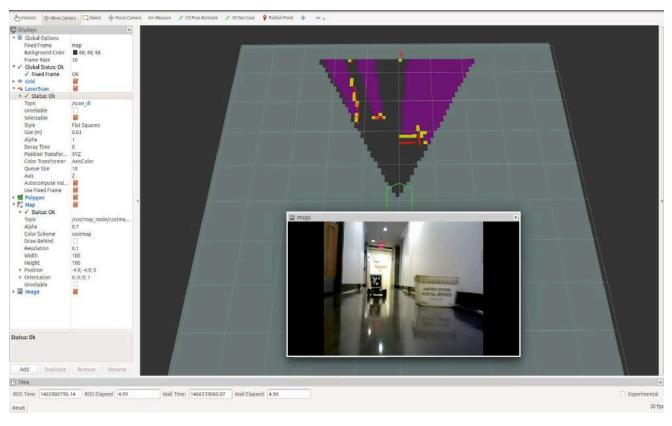


Virtual scan free space, obstacles, and occlusions

Putting it all together..

Estimate the position, orientation, and velocity of other racers?

Local trajectory Planner ? Overtaking?



F1/10 Challenges

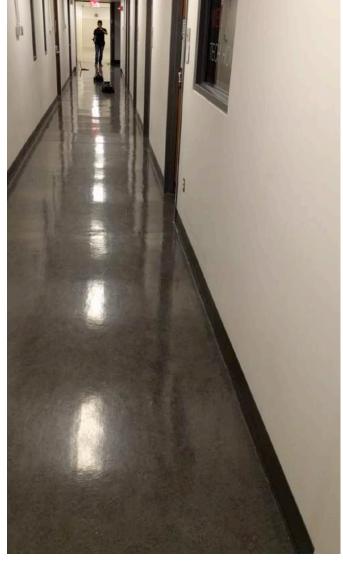
- Project Insanity: Testing the limits of racing control.
- Optimal racing line determination.
- Overtaking strategies and algorithms.

Overtaking Algorithms ?



Overtaking Algorithms ?





The F1/10 Racing Crew



Paril Jain



Matthew Kelly



Phil Hu



Nischal KN



Jake Scherlis



Houssam Abbas



Paritosh Kelkar



You !!



Madhur Behl



Matt Brady

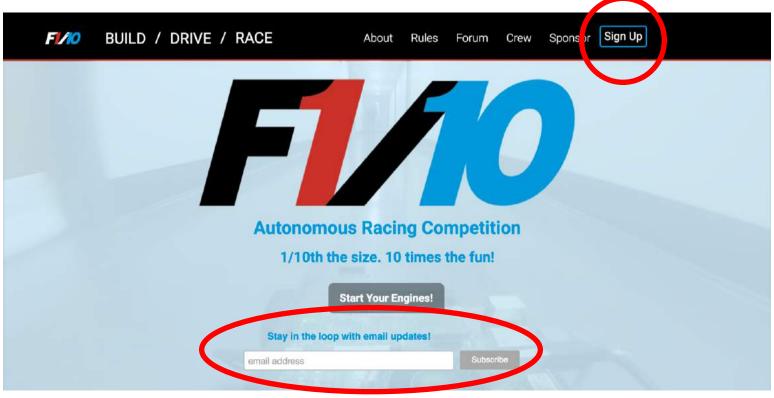


Carter Sharer



Rahul Mangharam

f1tenth.org



Get Involved !

No drivers were harmed during the making of this video.

Do you have what it takes to become the fastest autonomous racing driver ?

Do you have what it takes to become the fastest autonomous racing driver ?

Live demo ! Outside Wu & Chen Auditorium to follow