

## Ain Shams University Faculty of Engineering Computer And Systems Engineering

#### **Authors:**

David Nader
George Mohsen
Kirollos Henry
Kirollos Wadie

#### **Supervisor:**

Dr. Hossam Hassan

# Autonomous Robot for Mapping Unknown Environments (SLAM)

## Outline

- Problem
- SLAM
- Hector SLAM Algorithm
- Robot details
- Autonomous Navigation
- Results

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#### Problem

 Build a robot that can autonomously explore an unknown environment and build a 2D map of it

 Uses include: exploring inaccessible places (e.g. caves), dangerous places (e.g. mine detection), mapping indoor locations, etc...

• Main problem: SLAM & Autonomous navigation

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#### SLAM - Overview

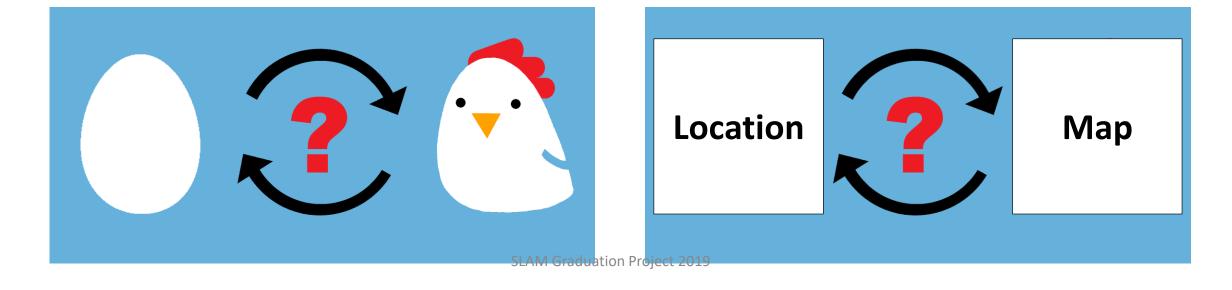
Simultaneous Localization And Mapping

 Building a map of an unknown environment while keeping track of the position of the robot in it

- Well-known problem in autonomous robots
  - e.g. self-driving cars

#### SLAM - Overview

- Why "simultaneous"?
  - Need map to localize, and
  - Need location to map
- Called *chicken-or-egg* problem



#### SLAM - Overview

- Many methods exist to solve the SLAM problem
- Examples: Kalman filters, Particle filters, Graph SLAM, etc.
- We used *Hector SLAM* algorithm

#### SLAM - LIDAR

- Range-sensor used in many SLAM algorithms
- Key advantages:
  - Accuracy and resolution (e.g. <1° resolution)</li>
  - High scan rate
     (e.g. 8000 samples/second)
  - Single sensor to measure in 360°



RPLIDAR A2

#### SLAM - LIDAR

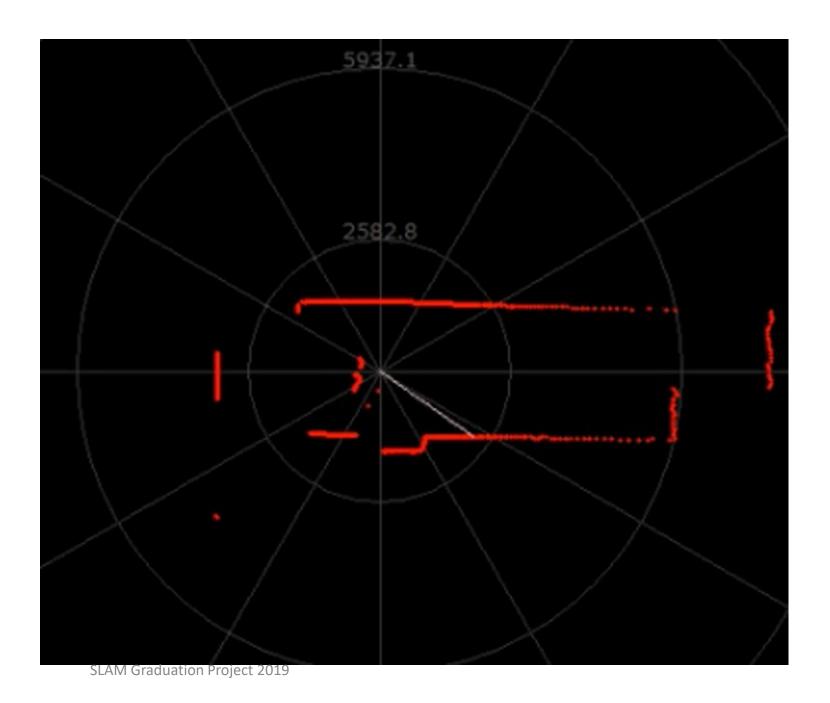
- Compare to SONAR/Visual SLAM
  - Lower accuracy and resolution
  - Require multiple sensors for 360° ranging
  - Slower scan rate
  - Visual SLAM: no ranging directly
  - Visual SLAM: higher computation



RPLIDAR A2

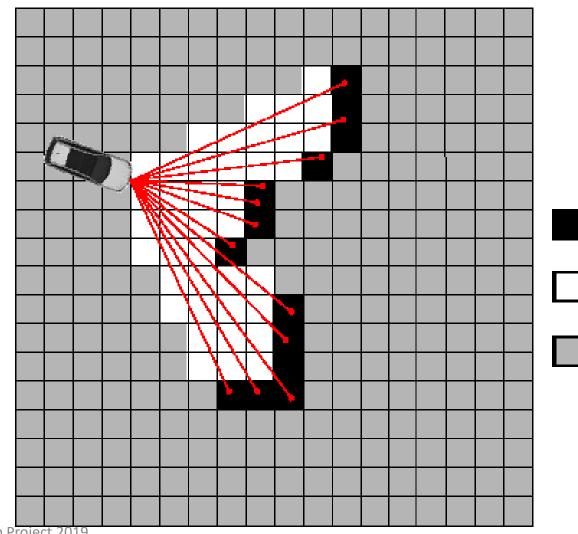
## SLAM - LIDAR

- Sample output
- Can be visualized as polar coordinates (angle + magnitude)



## SLAM - Occupancy Grid Map

- SLAM algorithms can produce different types of maps
- We're interested in Occupancy Grid maps
- Idea is: discretize the environment into grid cells (e.g. 10cm x 10cm)
- For each cell, we have one of 3 states (occupied, free, unknown)



SLAM Graduation Project 2019

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- Algorithm developed by *Team Hector* from "Technische Universität Darmstadt" (Technical University of Darmstadt) in Germany
- Presented in a published paper:

Kohlbrecher, Stefan & Von Stryk, Oskar & Meyer, Johannes & Klingauf, Uwe. (2011). A flexible and scalable SLAM system with full 3D motion estimation. 2011 IEEE International Symposium on Safety, Security, and Rescue Robotics (SSRR).

- Has an open-source implementation in ROS
- Does not require odometry like many other algorithms

- Initial position
- For motion in 2D plane, we have 3 DOF

$$\boldsymbol{\xi} = (p_x, p_y, \psi)^{\mathrm{T}}$$

Slides adapted from slides of University of Pennsylvania (<a href="http://fltenth.org">http://fltenth.org</a>)

SLAM Graduation Project 2019

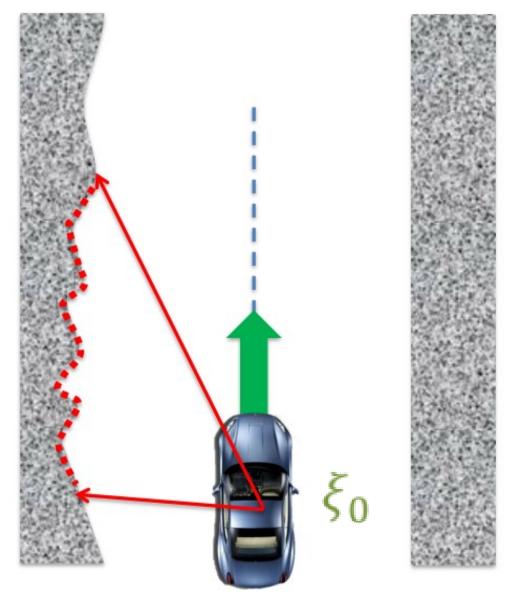
#### Left Wall

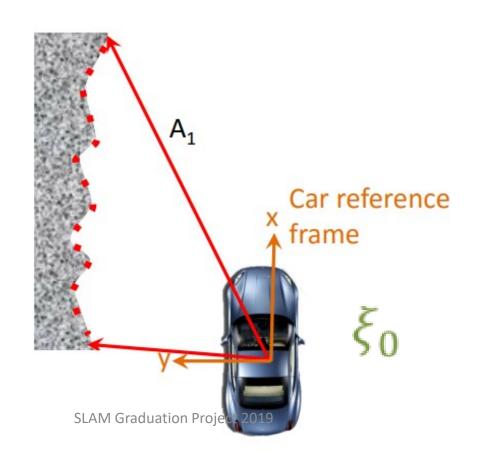


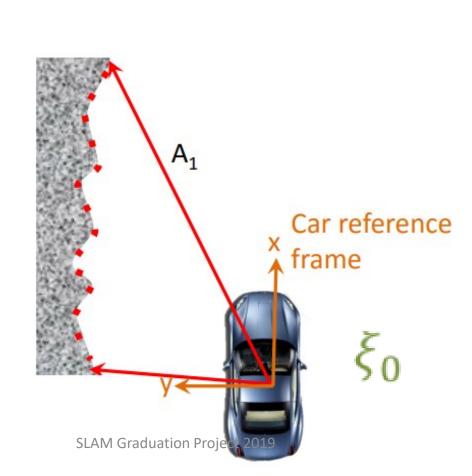
#### Right Wall

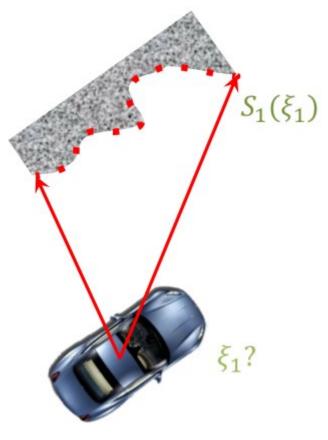


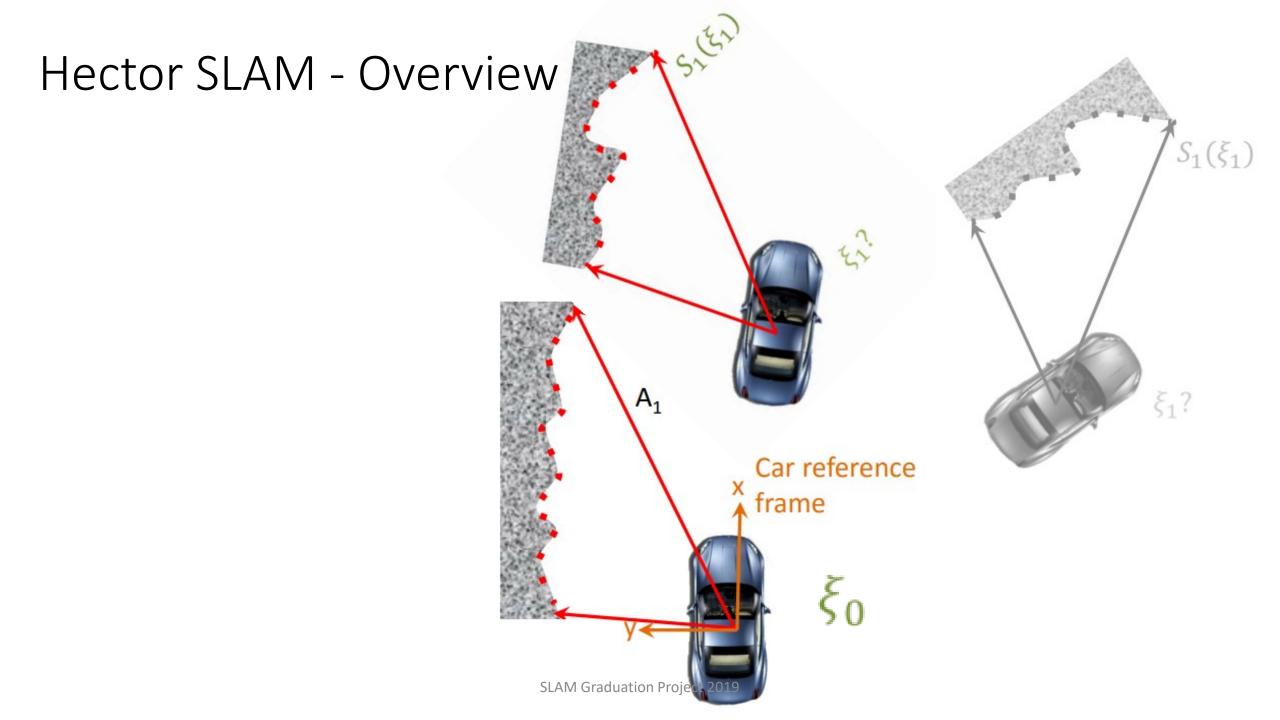
• First LIDAR scan

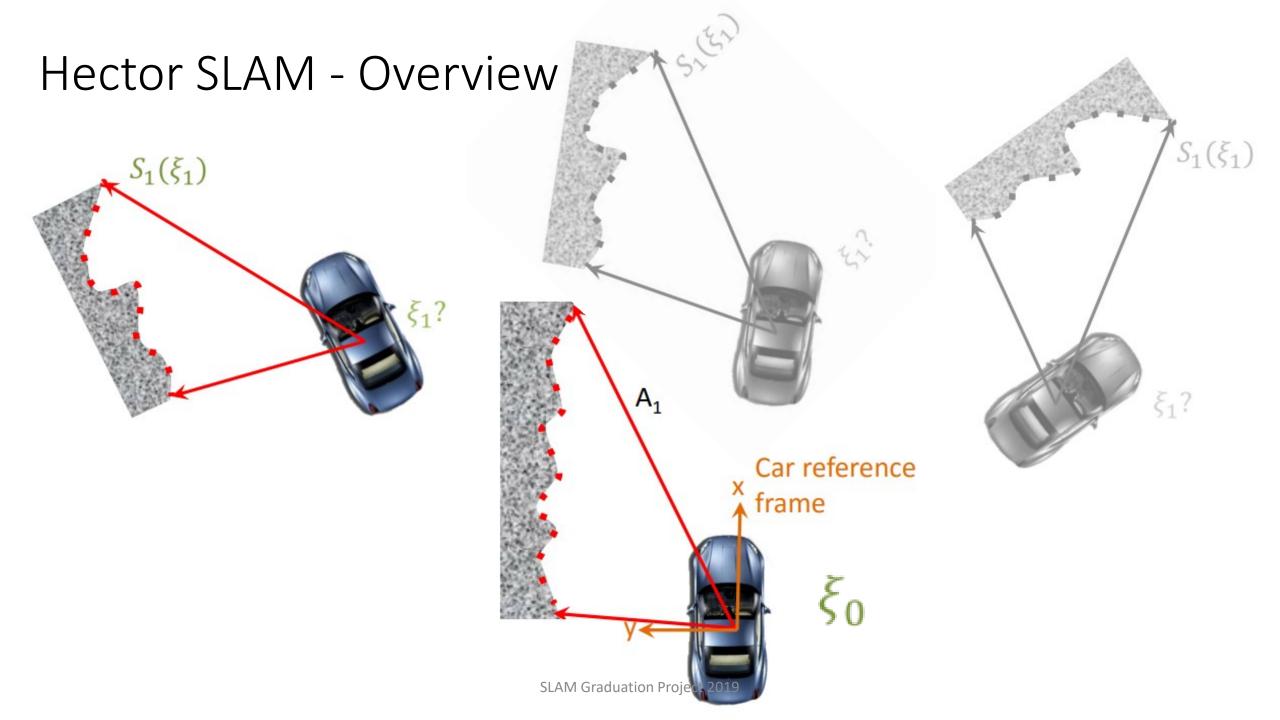


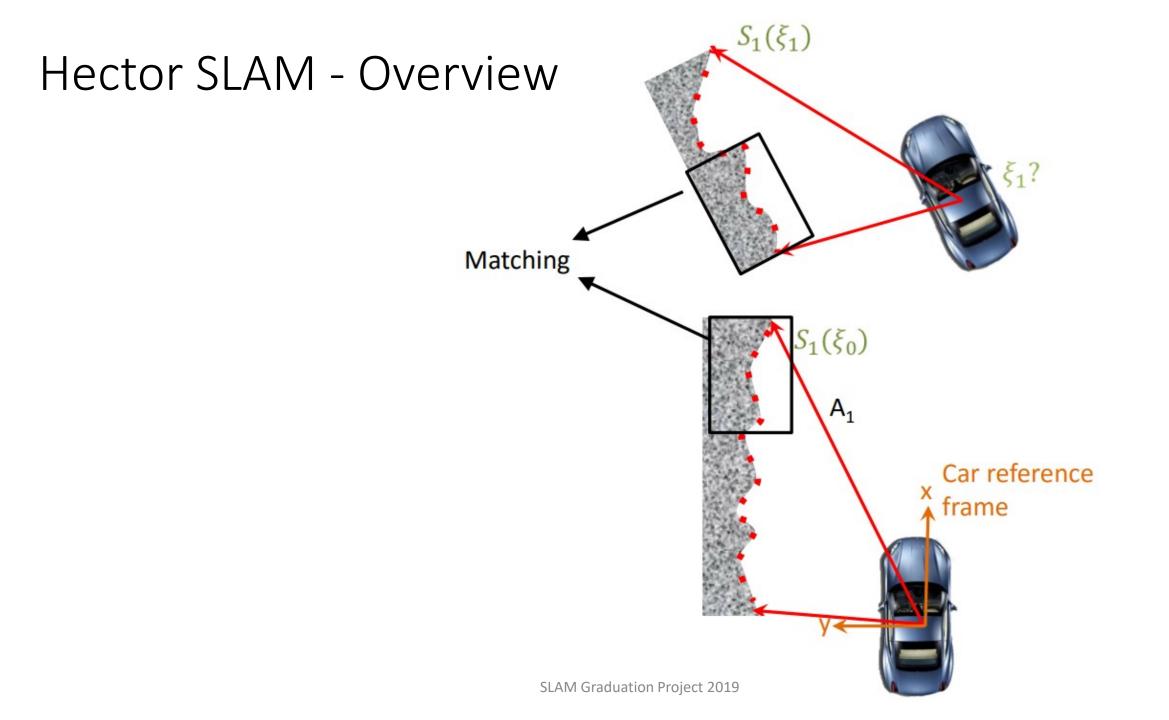


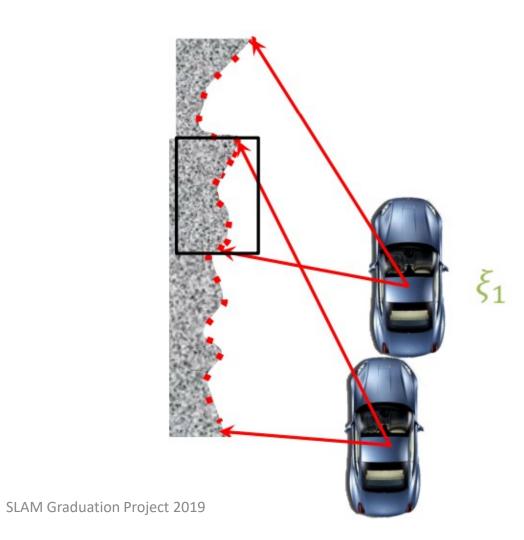












• This is an **optimization** problem

$$\boldsymbol{\xi}^* = \underset{\boldsymbol{\xi}}{\operatorname{argmin}} \sum_{i=1}^{n} \left[ 1 - M(\mathbf{S}_i(\boldsymbol{\xi})) \right]^2$$

- This is an optimization problem
- Find ξ that minimizes the square error between the existing map and the new scan

$$\boldsymbol{\xi}^* = \underset{i=1}{\operatorname{argmin}} \sum_{i=1}^{n} \left[ 1 - M(\mathbf{S}_i(\boldsymbol{\xi})) \right]^2$$

- This is an optimization problem
- Find ξ that minimizes the square error between the existing map and the new scan
- Error is the mismatch between the existing map and the new scan

$$\boldsymbol{\xi}^* = \underset{\boldsymbol{\xi}}{\operatorname{argmin}} \sum_{i=1}^n \left[ 1 - M(\mathbf{S}_i(\boldsymbol{\xi})) \right]^2$$

 After expansion and differentiation, we reach the Gauss-Newton equation, which is solved iteratively

$$\boldsymbol{\xi}^* = \underset{\boldsymbol{\xi}}{\operatorname{argmin}} \sum_{i=1}^{n} \left[ 1 - M(\mathbf{S}_i(\boldsymbol{\xi})) \right]^2$$

1

$$\Delta \boldsymbol{\xi} = \mathbf{H}^{-1} \sum_{i=1}^{n} \left[ \nabla M(\mathbf{S}_{i}(\boldsymbol{\xi})) \frac{\partial \mathbf{S}_{i}(\boldsymbol{\xi})}{\partial \boldsymbol{\xi}} \right]^{T} \left[ 1 - M(\mathbf{S}_{i}(\boldsymbol{\xi})) \right]$$

$$\mathbf{H} = \left[ \nabla M(\mathbf{S}_i(\boldsymbol{\xi})) \frac{\partial \mathbf{S}_i(\boldsymbol{\xi})}{\partial \boldsymbol{\xi}} \right]^T \left[ \nabla M(\mathbf{S}_i(\boldsymbol{\xi})) \frac{\partial \mathbf{S}_i(\boldsymbol{\xi})}{\partial \boldsymbol{\xi}} \right]$$

• After each iteration, we get a new  $\Delta \xi$  towards the minimum error

$$\boldsymbol{\xi}^* = \underset{\boldsymbol{\xi}}{\operatorname{argmin}} \sum_{i=1}^{n} \left[ 1 - M(\mathbf{S}_i(\boldsymbol{\xi})) \right]^2$$



$$\Delta \boldsymbol{\xi} = \mathbf{H}^{-1} \sum_{i=1}^{n} \left[ \nabla M(\mathbf{S}_{i}(\boldsymbol{\xi})) \frac{\partial \mathbf{S}_{i}(\boldsymbol{\xi})}{\partial \boldsymbol{\xi}} \right]^{T} \left[ 1 - M(\mathbf{S}_{i}(\boldsymbol{\xi})) \right]$$

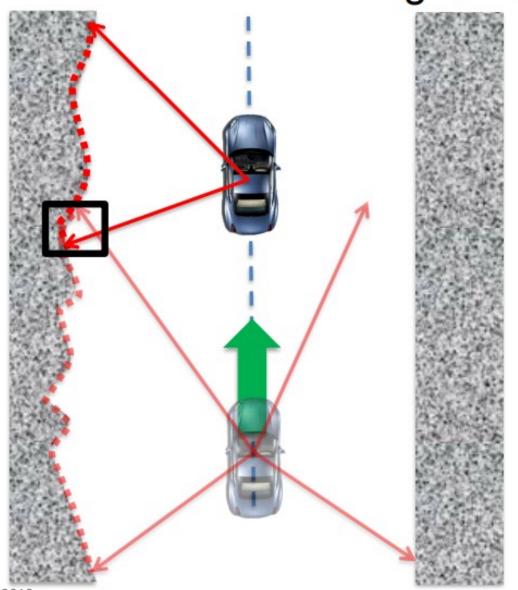
$$\mathbf{H} = \left[ \nabla M(\mathbf{S}_i(\boldsymbol{\xi})) \frac{\partial \mathbf{S}_i(\boldsymbol{\xi})}{\partial \boldsymbol{\xi}} \right]^T \left[ \nabla M(\mathbf{S}_i(\boldsymbol{\xi})) \frac{\partial \mathbf{S}_i(\boldsymbol{\xi})}{\partial \boldsymbol{\xi}} \right]$$

## Hector SLAM - Requirements

- Fast scan rate with respect to the robot's speed (and computation)
- To find an overlap between the new scan and the existing map
- Otherwise, the algorithm may fail to localize the robot correctly
- This is an advantage of the high scan rate of LIDARs

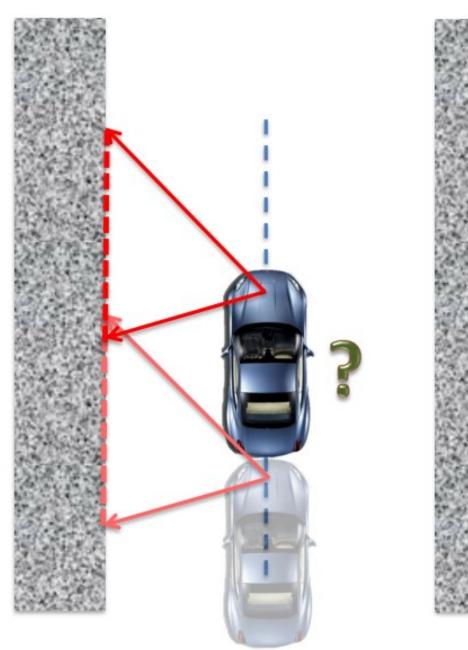
#### Left Wall

#### Right Wall



## Hector SLAM - Requirements

- Non-smooth, heterogeneous surfaces
- Otherwise, new scan looks exactly like existing map
- With no other source of odometry, algorithm thinks the robot didn't move
- Will be shown in the coming demo videos



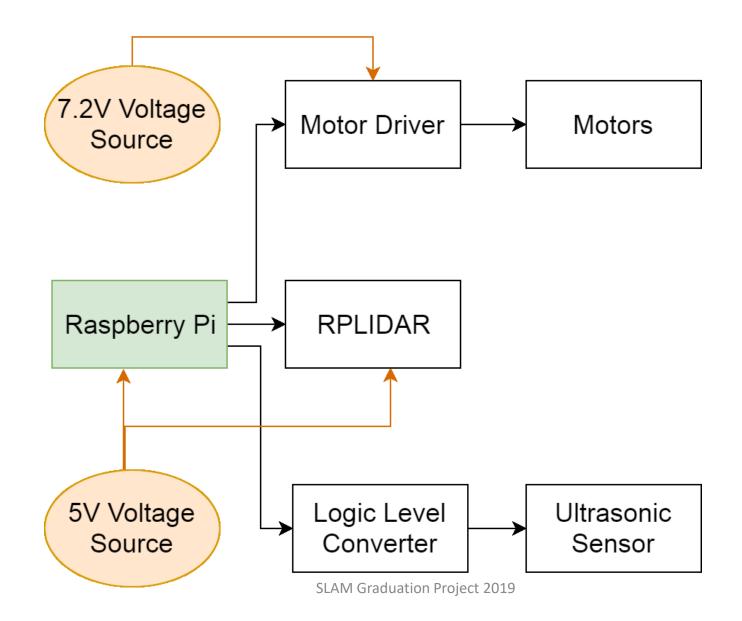
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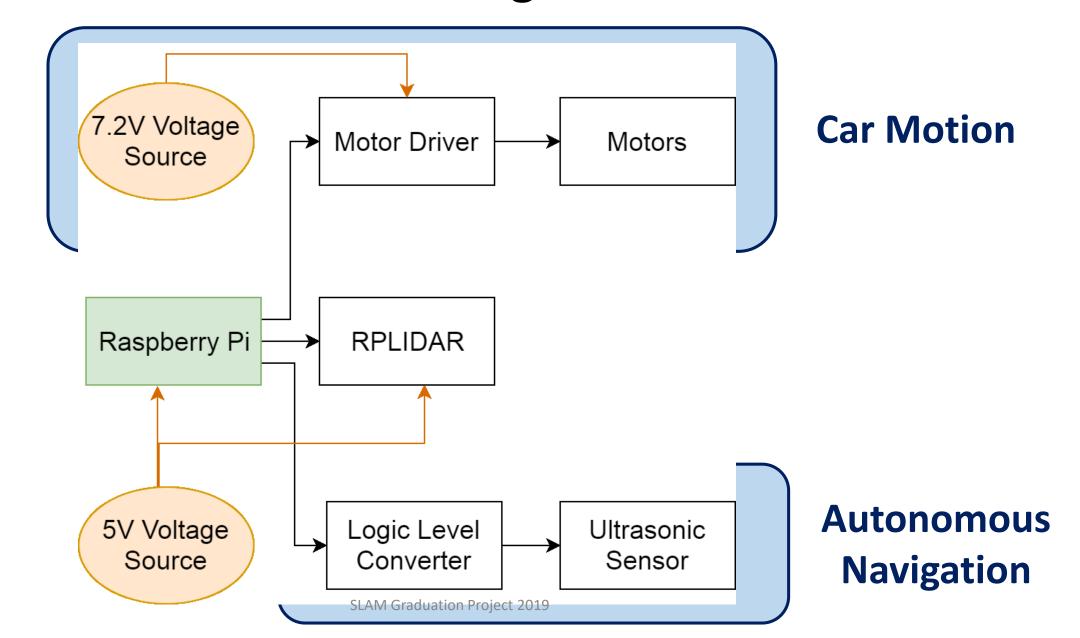
## Robot



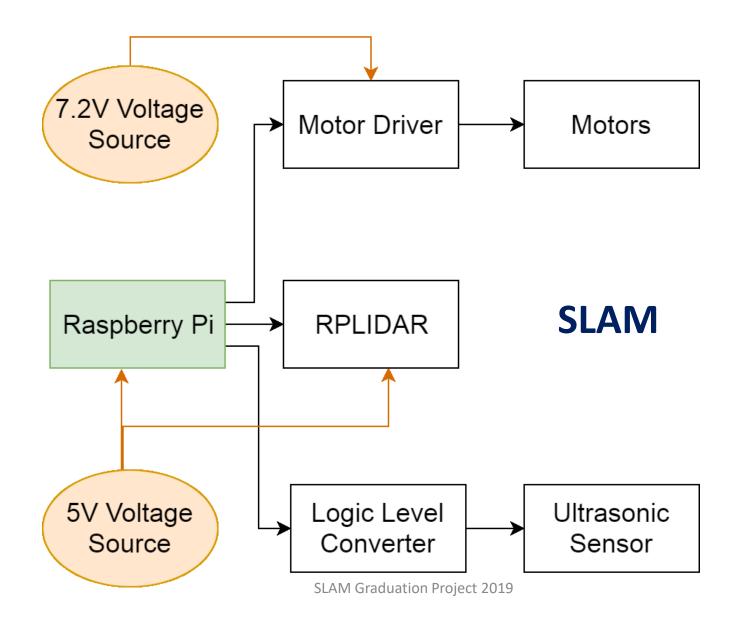
## Robot - Hardware Block Diagram



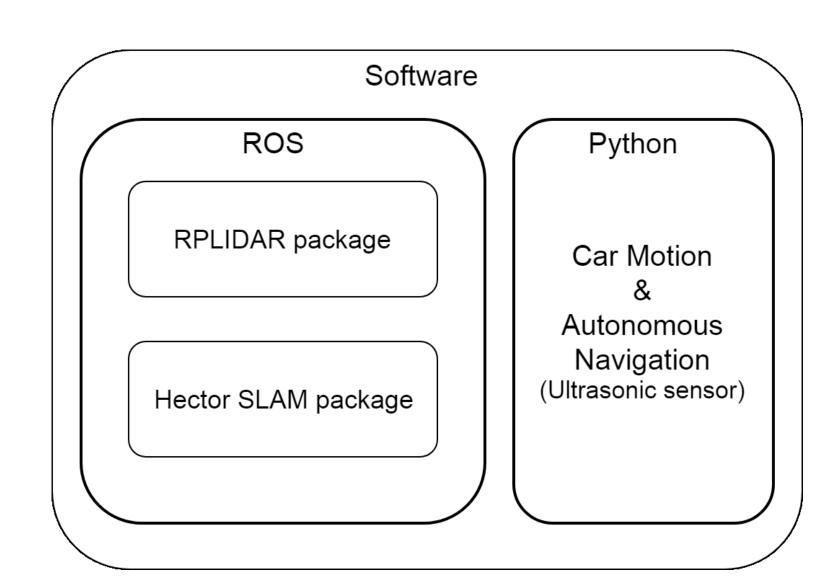
## Robot - Hardware Block Diagram



## Robot - Hardware Block Diagram

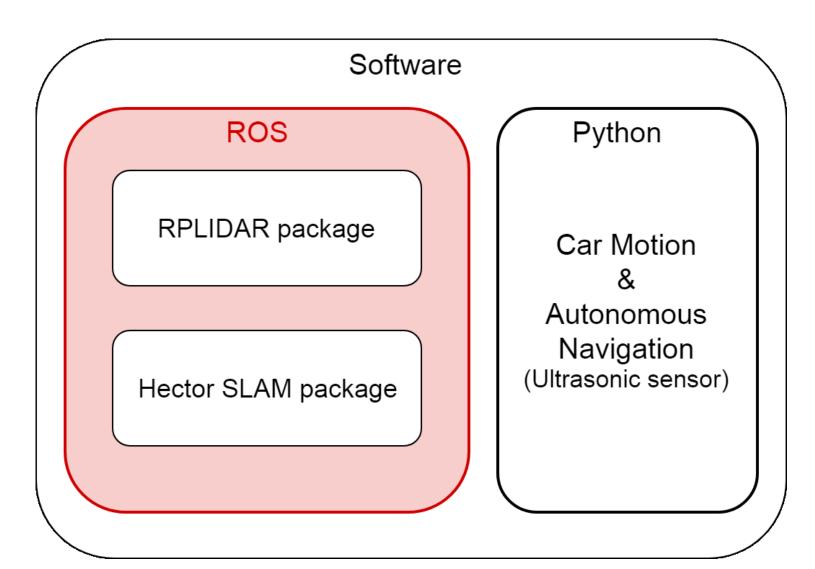


- Software runs on Raspberry Pi 3
- Two main components
  - ROS
  - Python code





- Robot Operating System
- Open-source, runs on UNIX
- Includes
  - Implementation of commonly-used functionality
  - Message-passing between processes
  - Package management
  - Etc.



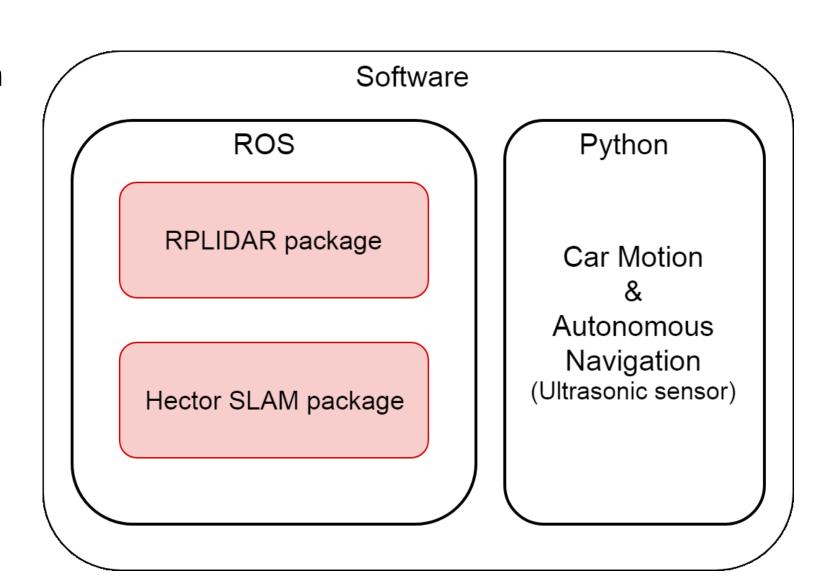
 We used 2 packages from ROS

#### 1. RPLIDAR

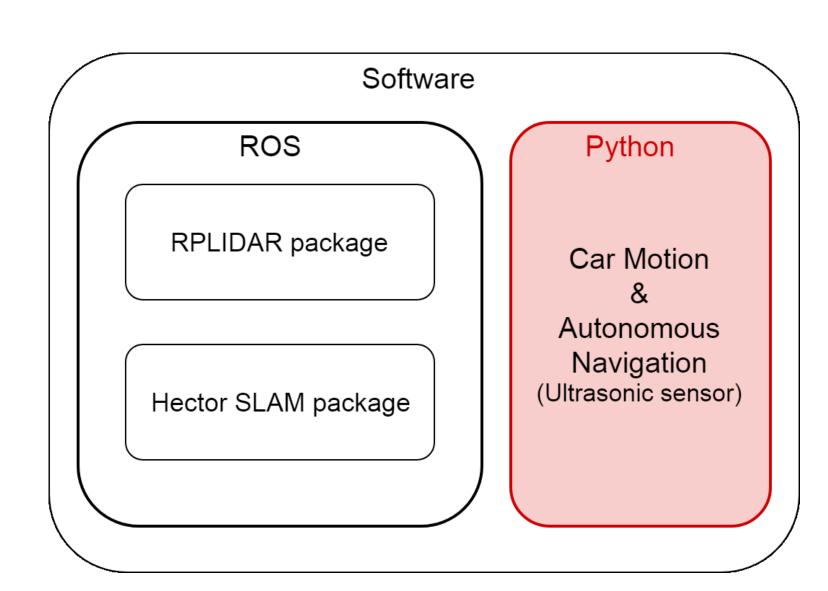
- Operates the LIDAR
- Publishes LIDAR readings

#### 2. Hector SLAM

- Subscribes to LIDAR readings
- Builds the map
- Estimates the robot's pose and trajectory



- We wrote separate python code to:
  - 1. Control the motor and move the car
  - 2. Drive the ultrasonic sensor for autonomous navigation



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## Autonomous Navigation

- Used Ultrasonic sensor
- Simple method: turn on detecting an obstacle
- Limited access to LIDAR
- Future work: use LIDAR instead of ultrasonic for navigation
  - LIDAR is more accurate
  - 360° ranging
  - Can implement Active SLAM
    - Active SLAM: deciding how should the robot moves in order to explore unvisited parts of the map
- Future work: improve mechanical build of the robot

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## Demo videos

## Thank You

#### **Contact us:**

Kirollos Henry

+201224959553

kirollostadros@gmail.com

**David Nader** 

+201226269655

david.nader.g@gmail.com