# SLAM Based Autonomous Navigation of Mobile Robot

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### INTRODUCTION

- Autonomous Navigation
  - Navigation without human intervention
- Simultaneous Localization and Mapping (SLAM)
  - Chicken-egg problem of robotics
- Localizing, mapping and navigating from one location to another.

### **NAVIGATION SYSTEM**

SENSOR FUSION

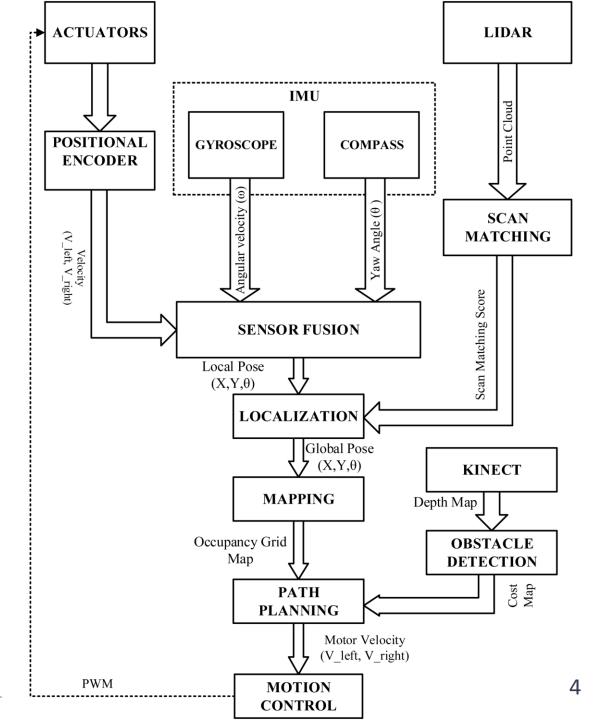
**LOCALIZATION** 

**MAPPING** 

**SLAM** 

PATH PLANNING MOTION CONTROL

# SYSTEM BLOCK DIAGRAM



### **APPLICATIONS**



Cleaning with Navigation (left)

Automatic
Transportation of load
(right)

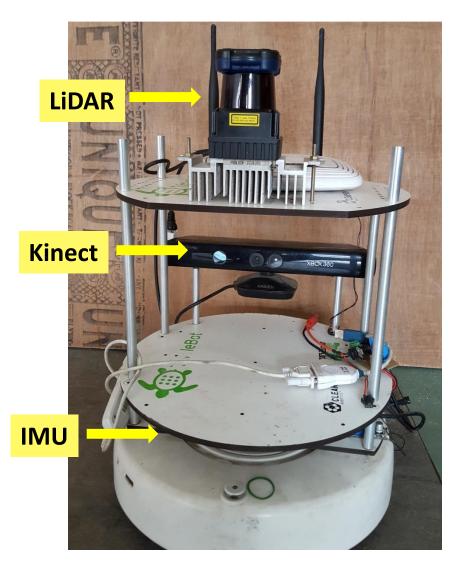
#### **TOOLS AND PLATFORMS**

#### Hardware

- Inertial Measurement Unit (IMU)
- Encoder
- LiDAR
- ARM
- Kinect

#### Software

- Robot Operating System (ROS)
- STM HAL
- Keil uVision IDE



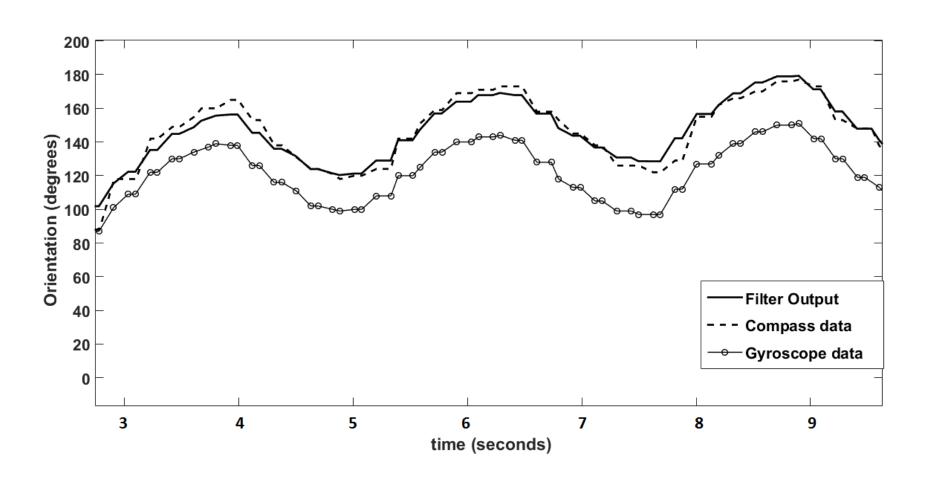
### **SENSOR FUSION**

- Sources of odometry
  - Encoder + IMU (Gyroscope + Compass)
- Fusion output is corrected orientation and position

Algorithm: Kalman filter

$$x(t-1|t-1)$$
  $\xrightarrow{predict} x(t|t-1)$   $\xrightarrow{measurem} x(t|t)$   $\xrightarrow{update} x(t|t)$   $\xrightarrow{priori}$   $\xrightarrow{nexstat}$ 

### **SENSOR FUSION: Kalman Filter**



### LOCALIZATION

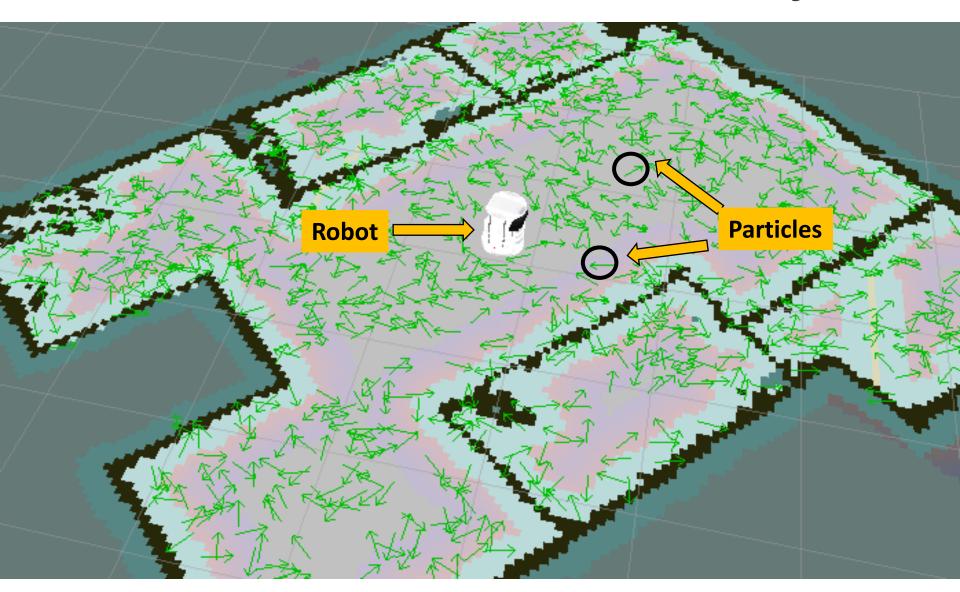
- Positioning the robot in world space.
- Appropriate algorithms to localize using multiple sensors data

#### **Algorithms**

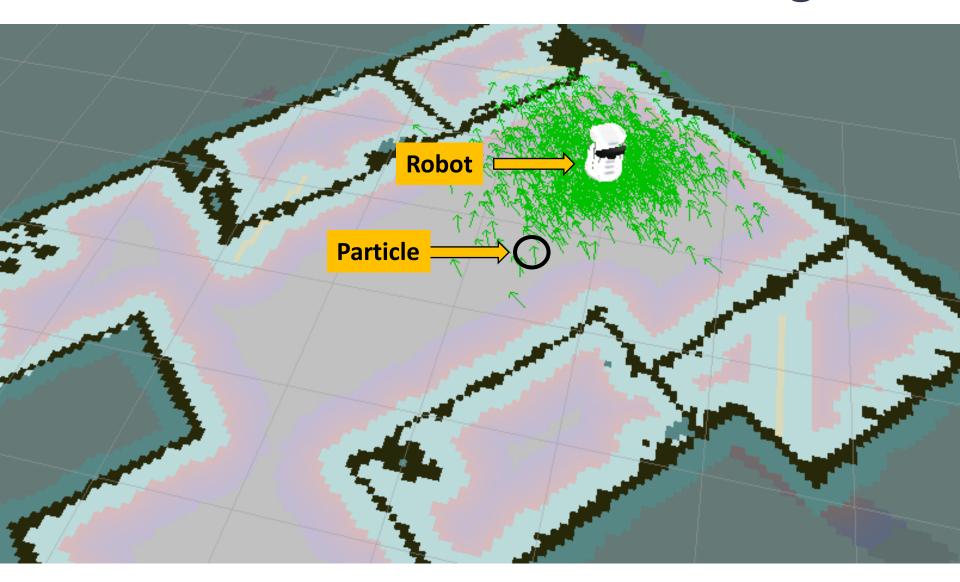
- Monte Carlo Localization (MCL)
- Adaptive Monte Carlo Localization(AMCL)

$$\begin{array}{c|c} p(x_{0t} \mid u_{1t}, z_{1t}, m) \\ \text{where,} \quad u_{1t} = \{u_1, u_2, \ldots u_t\} & \longrightarrow \text{ Control input} \\ z_{1t} = \{z_1, z_2, \ldots z_t\} & \longrightarrow \text{ Observations} \\ x_{0t} = \{x_0, x_1, \ldots x_t\} & \longrightarrow \text{ Path of the robot} \\ \mathcal{M} & \longrightarrow \text{ Map of the environment} \end{array}$$

## **LOCALIZATION:** Random Particle Injection



# **LOCALIZATION: Particle Converges**



### **MAPPING**

- Map Development of surrounding
- LiDAR fused with odometry data for reliability

Algorithm: Occupancy Grid Mapping

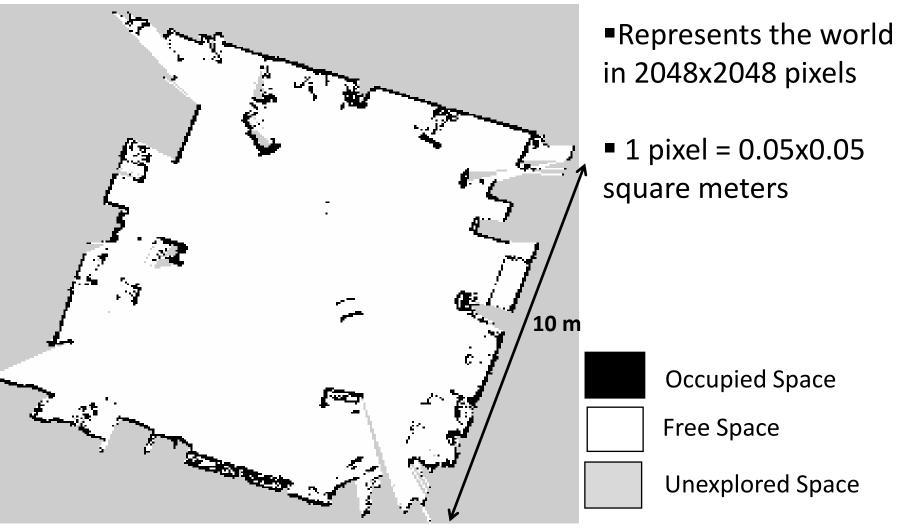
$$p(m|x_{lt},z_{lt})$$

Where, 
$$Z_{1:t} = \{Z_1, Z_2, \dots, Z_t\} \longrightarrow \text{Observations from time 1 to t}$$

$$X_{1:t} = \{X_1, X_2, \dots, X_t\} \longrightarrow \text{Positions of the robot}$$

$$M \longrightarrow \text{Map of the environment}$$

### **MAPPING: Occupancy Grid Map**



### **PATH PLANNING: Global**

- Optimum path planning between source and destination
- Graph Search for finding routes

**Algorithm**: A\*

$$f(n)=g(n)+h(n)$$

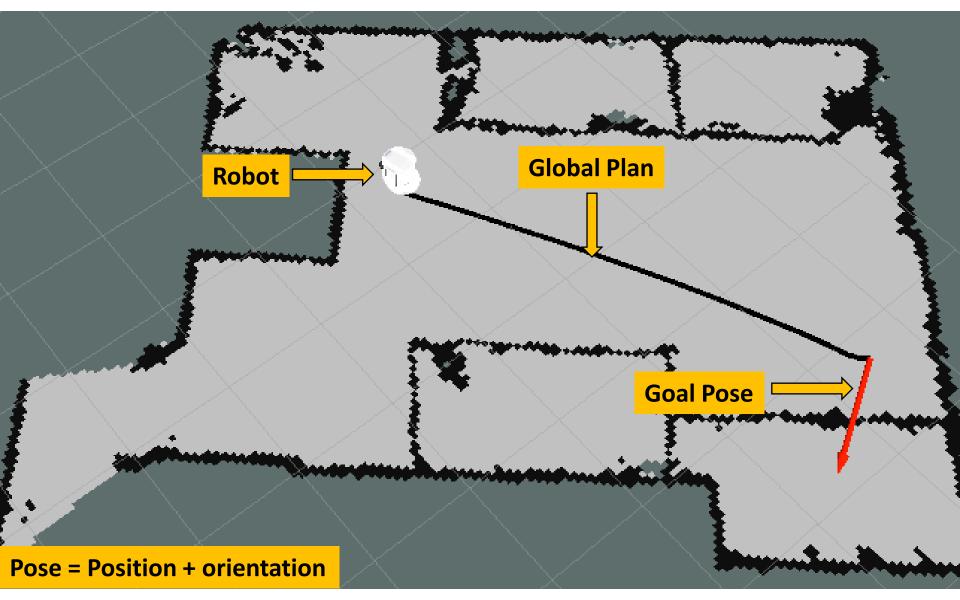
Where,

 $f(n) \longrightarrow \text{Total cost of node n}$ 

 $g(n) \longrightarrow \text{Cost of reaching node n from starting location}$ 

 $h(n) \rightarrow \text{Value of heuristic function at node n}$ 

## **PATH PLANNING: Global**



### **PATH PLANNING: Local**

Execute segments of the Global plan sequentially

Algorithm: Dynamic Window Approach (DWA)

$$Q(v,w)=a*headim (v)+b*veloc(v,w)+c*cleara(v,w)$$

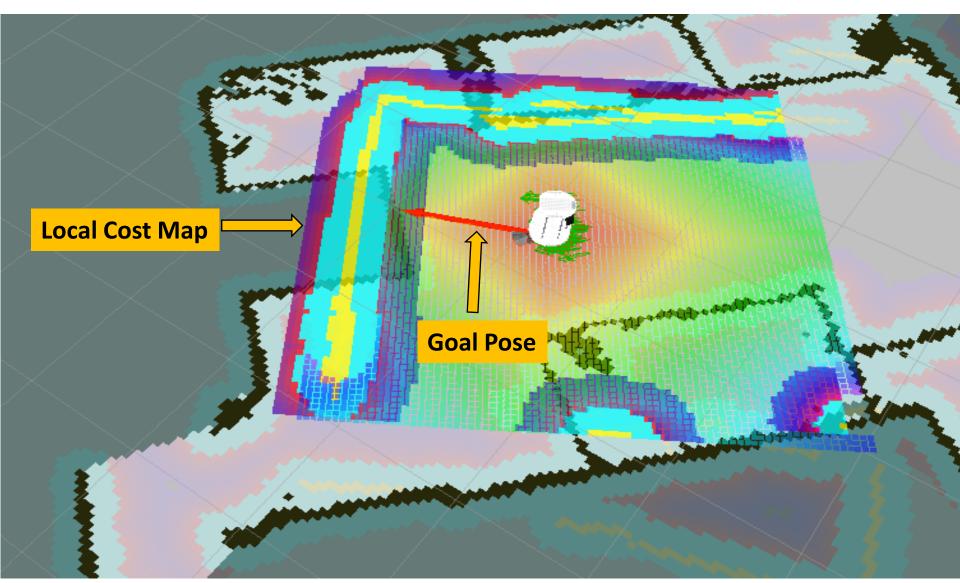
Where,  $Q(v, w) \longrightarrow$  Objective function for linear velocity v and angular velocity w

 $heading) \rightarrow$  Heading towards goal

cleara(v) Clearance from obstacle

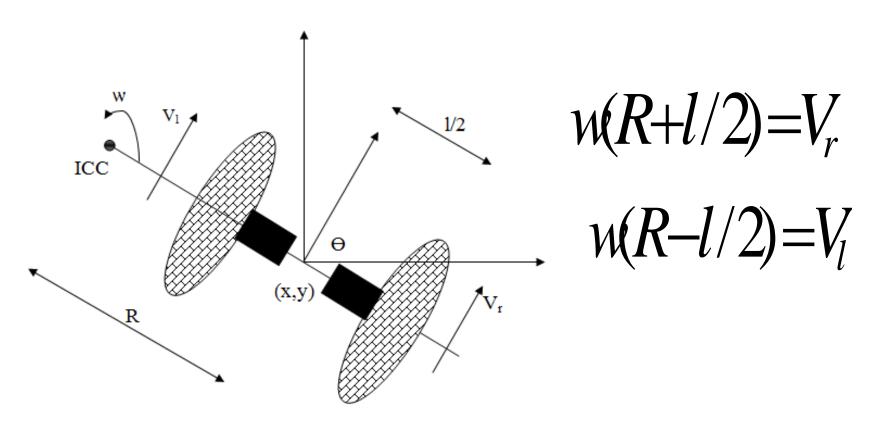
 $veloc(v) \longrightarrow$  Forward velocity of the robot

### **PATH PLANNING: Local**



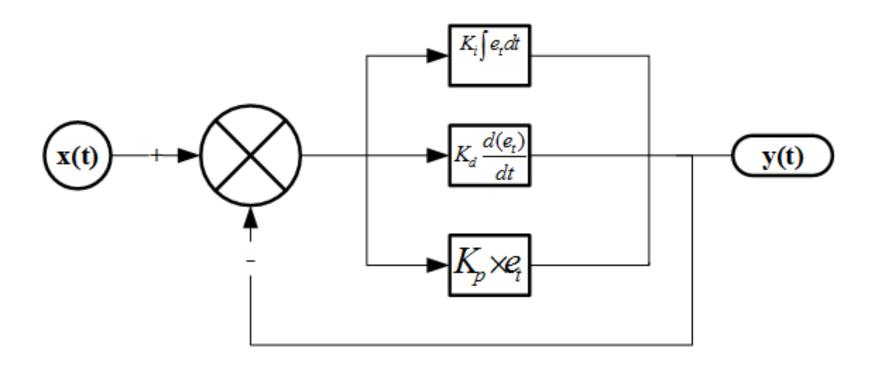
### **MOTION CONTROL: Kinematics**

Differential Drive Kinematics



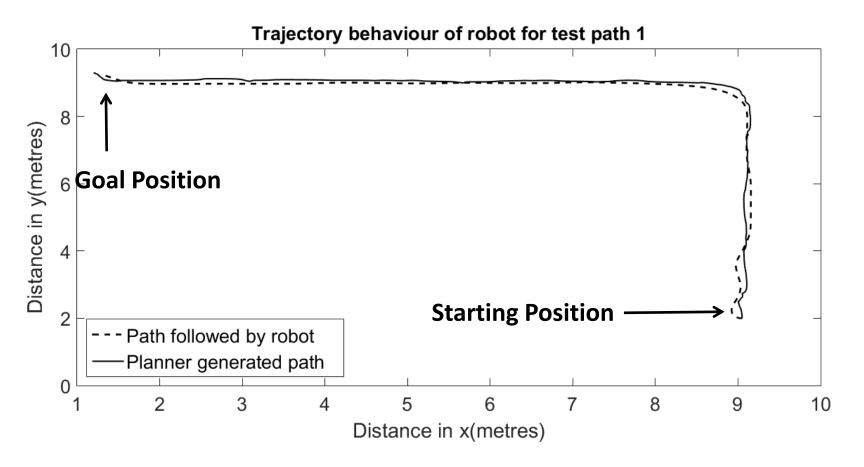
### **MOTION CONTROL: PID**

PID control on the motors to attain desired velocities



### **COMPLETE NAVIGATION**

Path Planned and actual trajectory followed by robot



### **COMPLETE NAVIGATION**

Test path	Final Plan position( $x_p, y_p$ )(m)	Final postion of robot( $x_r, y_r$ )(m)	Deviation at goal(m)
1	(1.367,9.194)	(1.346,9.211)	0.027
2	(4.296,1.984)	(4.45,2.05)	0.1675
3	(4.037,3.1)	(3.985,3.054)	0.0694
4	(9.15,4.5)	(9.25,4.38)	0.1562

Maximum deviation from goal in four experiments =
 16.75 cm

#### CONCLUSION

- Sensor Fusion of IMU data for orientation with 2 degree of maximum error
- Localization with MCL using 500 particles
- Mapping with grid size 2048x2048 and scale: 1 pixel equivalent to 0.05 meters
- Path planning with less than 0.2 meters deviation on final location during navigation

# Thank you!...