# INTRO TO SLAM & NAV

Presented by Victoria Rotich

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### OVERVIEW

**O1** What is Navigation

Why Navigation matters

Why SLAM is needed

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**SLAM Approaches** 

ROS2 Tools for SLAM & Navigation

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Path Planning in Detail

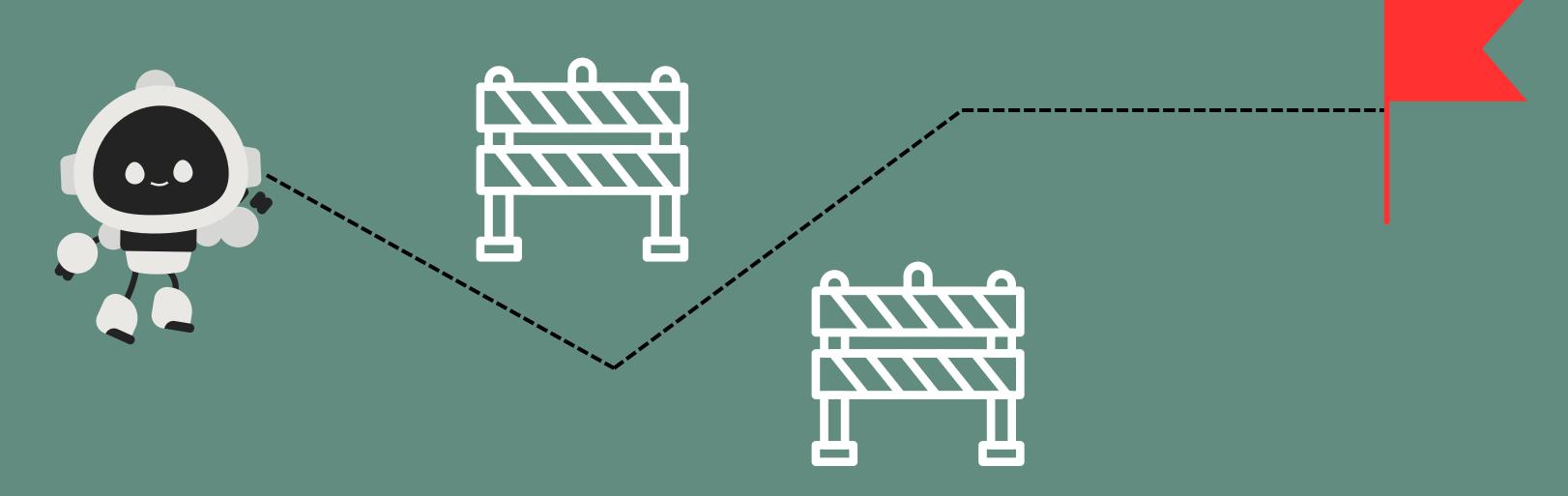
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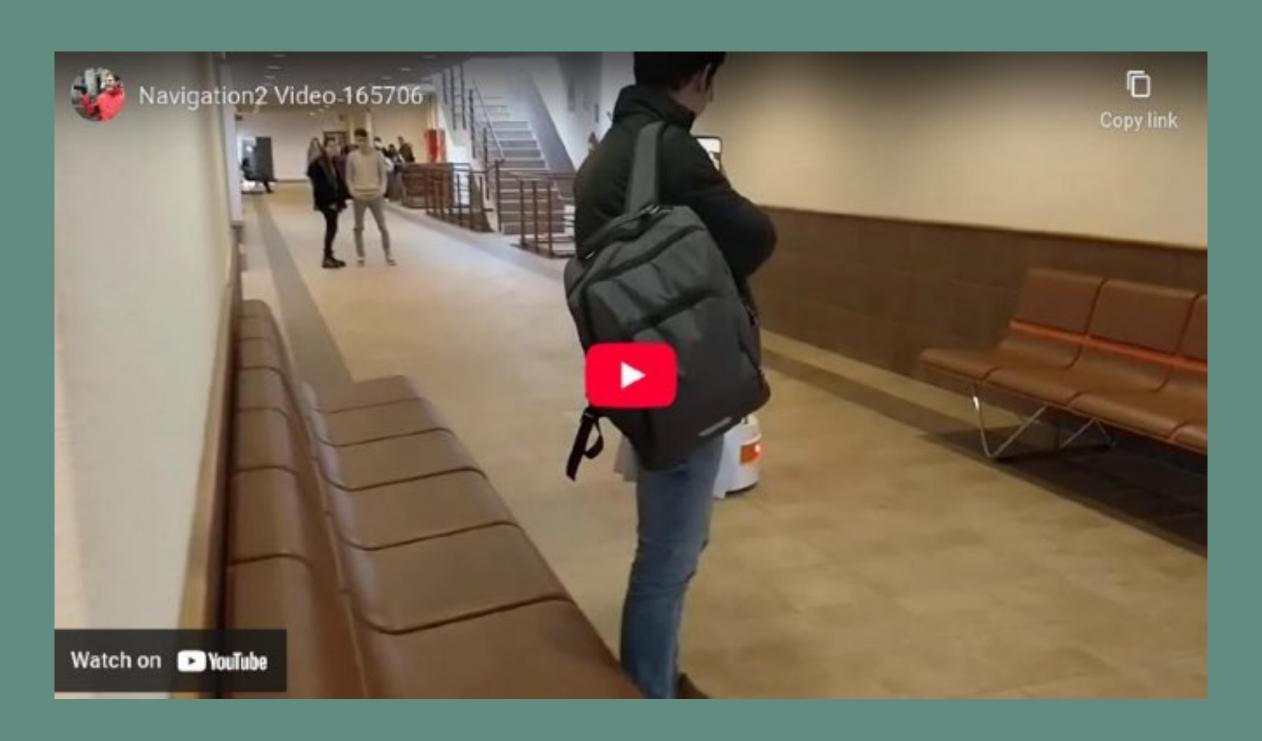
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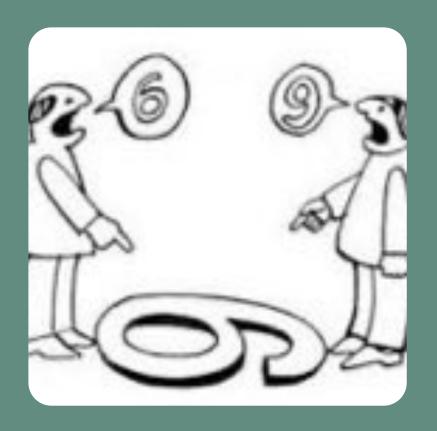
SLAM\_Toolbox vs AMCL

Robot navigation = moving from a start → goal, safely & efficiently

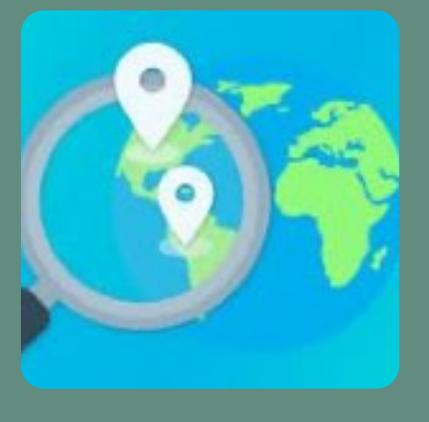




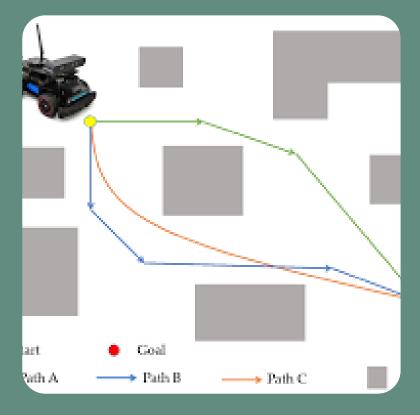
It requires the following:



Perception (sense world)



Localization (know where you are)



Path Planning ( decide where to go)



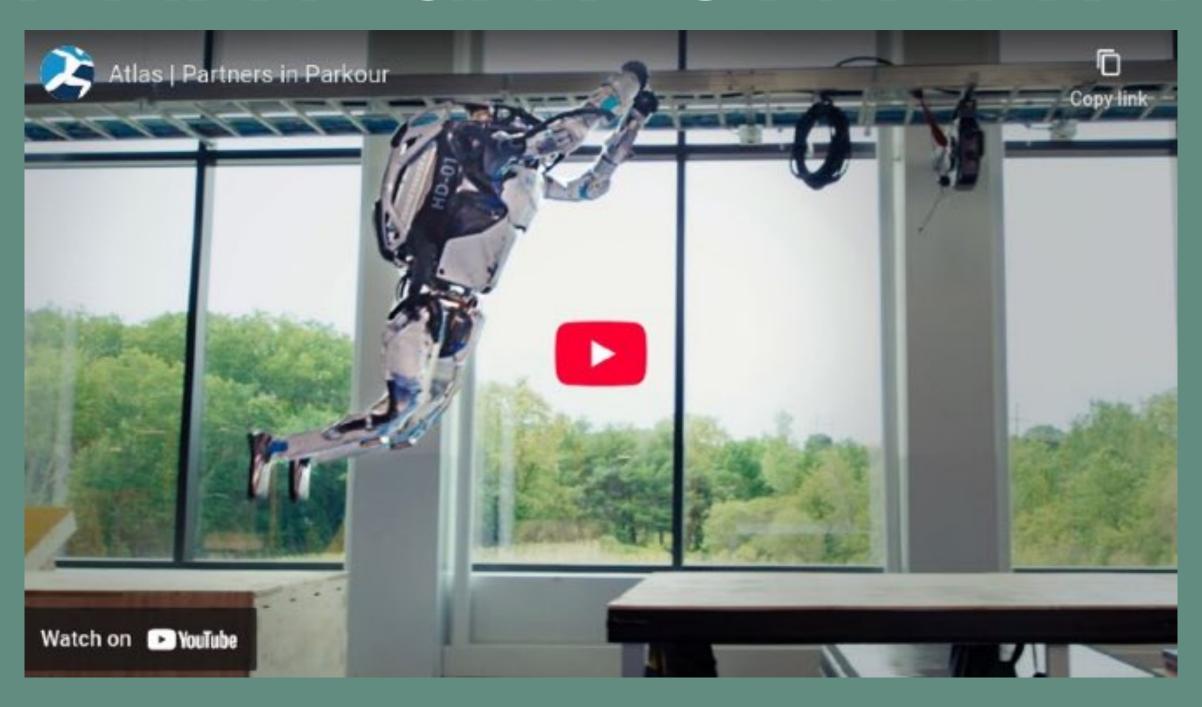
Control (execute movement)

Q: Why do you think navigation matters?

### WHY NAVIGATION MATTERS

- Autonomous Operation
- Obstacle Avoidance and Safety
- Efficient and Optimized Paths
- Adaptability to changing Environments
- Enabling Complex Tasks
- Foundation for Advanced Capabilities

# WHY NAVIGATION MATTERS



### WHY SLAM IS NEEDED

- SLAM Simultaneous Localization and Mapping
- Refers to creating a map of an unknown environment while simultaneously determining location
- Robots need maps for navigation in unknown spaces
- GPS is unreliable indoors

# APPLICATIONS OF SLAM



Automated vehicles & robotics



AR & VR



Agriculture



Medicine

# SLAM IS DIFFICULT



# SLAMIS DIFFICULT

In order to accurately localize, we need good maps, but to have good maps, we need to accurately localize

### SLAM APPROACHES

We'll break them down into 4 types:

-Filter Based SLAM

-Graph Based SLAM

-Deep Learning Based SLAM

- Modern Methods

Q: What types of filters have you worked with before?

### Overview

A class of algorithms that use a probabilistic filter, to recursively estimate a robot's pose and simultaneously build a map of an unknown environment from sensor data

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A class of algorithms that use a probabilistic filter, to recursively estimate a robot's pose and simultaneously build a map of an unknown environment from sensor data

#### How it works

- 1. State Estimation
- 2. Prediction Step
- 3. Measurement Update (Correction Step)
- 4. Recursion

#### Kalman Filter-based SLAM

- Extended Kalman Filter (EKF) SLAM
   →For non-linear systems
- Unscented Kalman Filter (UKF) SLAM
   →An improvement over EKF

### Particle Filter (PF) SLAM

- Also known as FastSLAM
- Uses a collection of particles to represent the robot's state distribution

# 2. GRAPH BASED SLAM

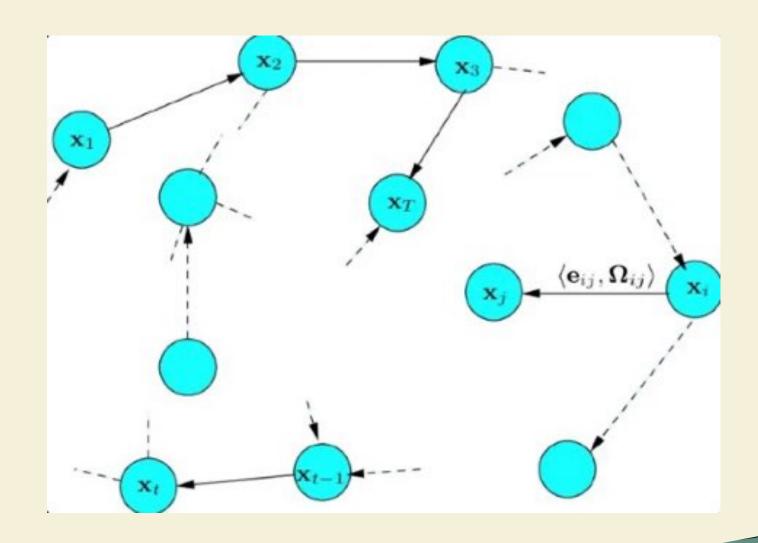
#### Overview

A technique that represents the problem as a graph where nodes are the robot's poses over time and edges are constraints from sensor measurements and odometry.

### How it works

- 1. Graph Construction2. Error Minimization

# 2. GRAPH BASED SLAM



# 3. DEEP LEARNING BASED SLAM

#### **Overview**

Attempts to solve the SLAM problem using neural networks and deep learning.

### **Existing Algorithms:**

They include:

- 1. ŘatSLAM
- 2. LIFT-SLAM
- 3. EnvSLAM

### A. Visual SLAM

Uses camera-based visual input to simultaneously build a map of an unknown environment and track its own position and orientation within that map.

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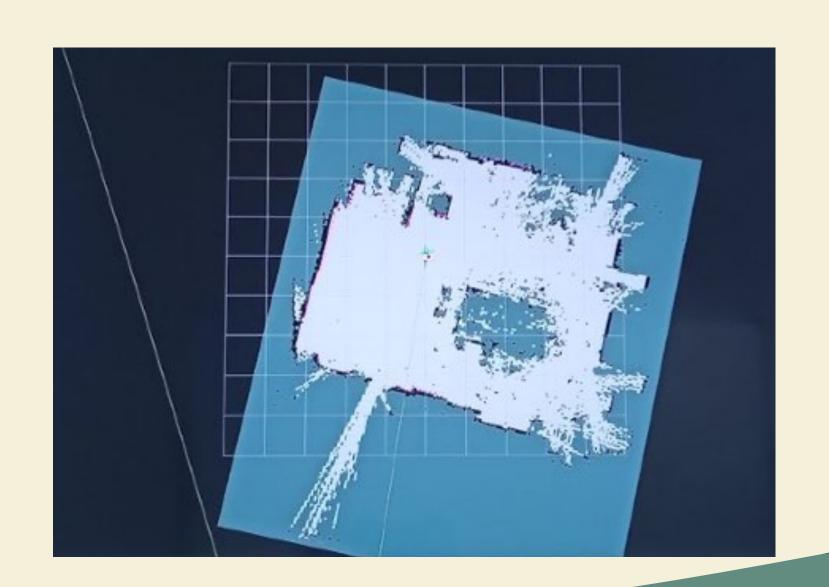
#### **B. LiDAR-based SLAM**

Uses a laser sensor to generate a 3D map of an environment while simultaneously determining the sensor's precise location within that map.

A. Visual SLAM



**B. LiDAR-based SLAM** 



# ROS2 TOOLS FOR SLAM & NAVIGATION

1.slam\_toolbox → 2D LiDAR-based mapping

# ROS2 TOOLS FOR SLAM & NAVIGATION

1.slam\_toolbox

2.Nav2 (Navigation2) → stack for path planning, control, recovery behaviors

# ROS2 TOOLS FOR SLAM & NAVIGATION

- 1.slam\_toolbox
- 2.Nav2 (Navigation2)
- 3. TurtleBot4 → mobile robot intended for education and research

Synchronous and Asychronous

Online & offline

### Synchronous SLAM

Requires all operations to complete in strict order before the next one can begin, acting as a blocking architecture

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Requires all operations to complete in strict order before the next one can begin, acting as a blocking architecture

### **Asynchronous SLAM**

Allows operations to run in parallel or be deferred, not blocking the main process and providing more flexibility for tasks like map updates

#### **Online SLAM**

A real-time process, used by robots during operation to build maps while simultaneously navigating by prioritizing speed and immediate control, even if it means a less precise or incomplete map.

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#### Offline SLAM

A post-processing technique that takes place after data collection, allows for computationally intensive algorithms, and focuses on creating a high-quality, comprehensive map and closing loops to ensure accuracy

Prerequisites:

RViZ setup

<u>Ignition Gazebo Fortress already installed</u>

Installing slam\_toolbox and Nav2:

sudo apt install ros-humble-slam-toolbox ros-humble-navigation2 roshumble-nav2-bringup

For WSL2 Users

Setting Environment Variables:

export LIBGL\_ALWAYS\_SOFTWARE=1
export MESA\_GL\_VERSION\_OVERRIDE=3.3
export MESA\_GLSL\_VERSION\_OVERRIDE=330

For WSL2 Users

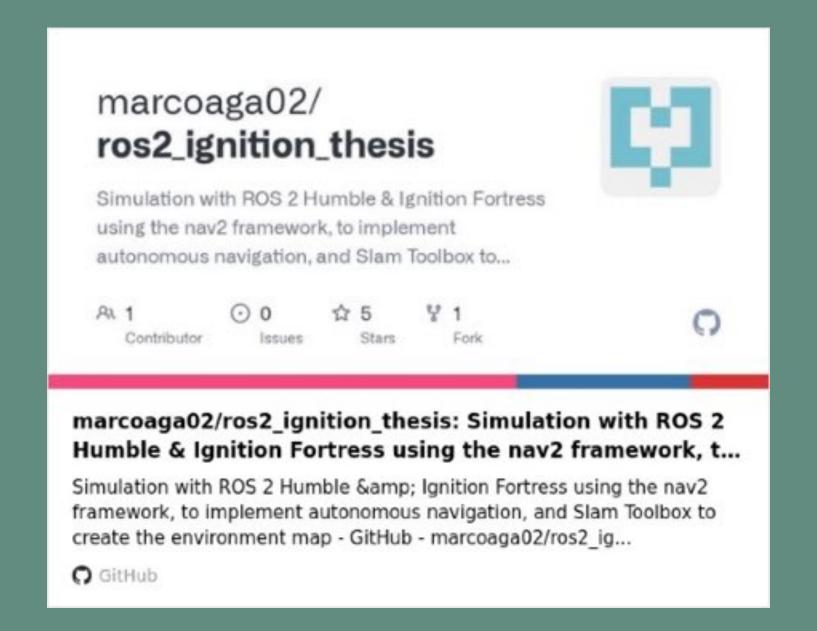
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# SIMULATION DEMO

Watch a simulation

# SIMULATION DEMO



A. Global vs Local planners

- A. Global vs Local planners
  - 1.Global planner → generates a high-level, long-term path on a known map

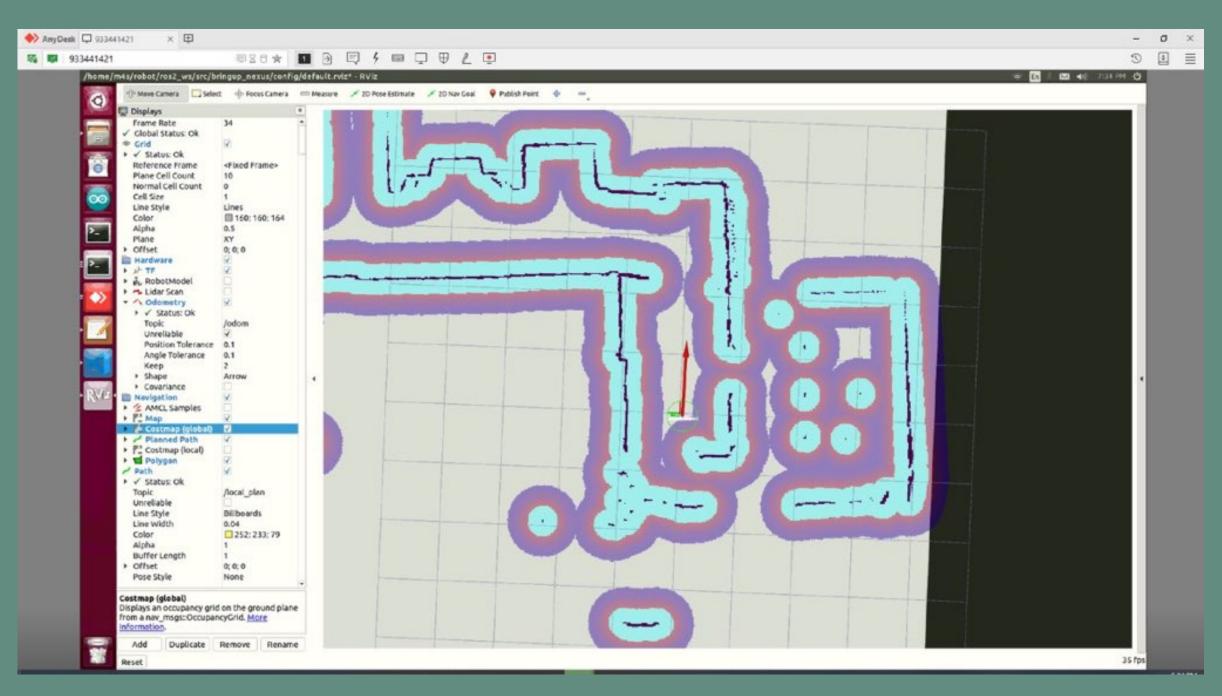
- A. Global vs Local planners
  - 1. Global planner
  - 2. Local planner → uses real-time sensor data to navigate the robot along this path, dynamically avoiding immediate, previously unknown obstacles and making adjustments to avoid collisions

The global planner provides the "big picture" route, and the local planner handles the "in-the-moment" path execution and obstacle avoidance.

Helps us generate a <u>costmap</u>

The costmap holds information about the environment, such as obstacles or areas that the vehicle cannot traverse.

To check for collisions, the costmap inflates obstacles using the inflation radius specified in the CollisionChecker property.



B. Algortithms

- B. Algorithms
- 1.Dijkstra's Algorithm:
  - Finds the shortest path between nodes in a graph
  - Can be computationally expensive in large environments.

- B. Algorithms
- 2. Sampling-Based Algorithms:
  - Rapidly- exploring Random Tree (RRT) and its variants
  - Probabilistic Roadmaps (PRM)

- B. Algorithms
- 3. Intelligent Algorithms (Inpired by Bionics):
  - Ant Colony Optimization
  - Dolphin SLAM

- B. Algorithms
- 4. Optimization-Based Algorithms:
  - Trajectory Optimization
  - Time Elastic Band
  - Greedy Optimization

## RVIZ PLUGIN SETUP

Launch RViZ by typing "rviz2"

# AMCL

- AMCL- Adaptive Monte Carlo Localization
- A probabilistic algorithm that uses a particle filter to estimate a robot's position and orientation (pose) on a known 2D map by matching laser scan data from sensors like LIDAR with the map.

# SLAM\_TOOLBOX VS AMCL

 AMCL → A localization-only system that uses a preexisting map to determine a robot's position

# SLAM\_TOOLBOX VS AMCL

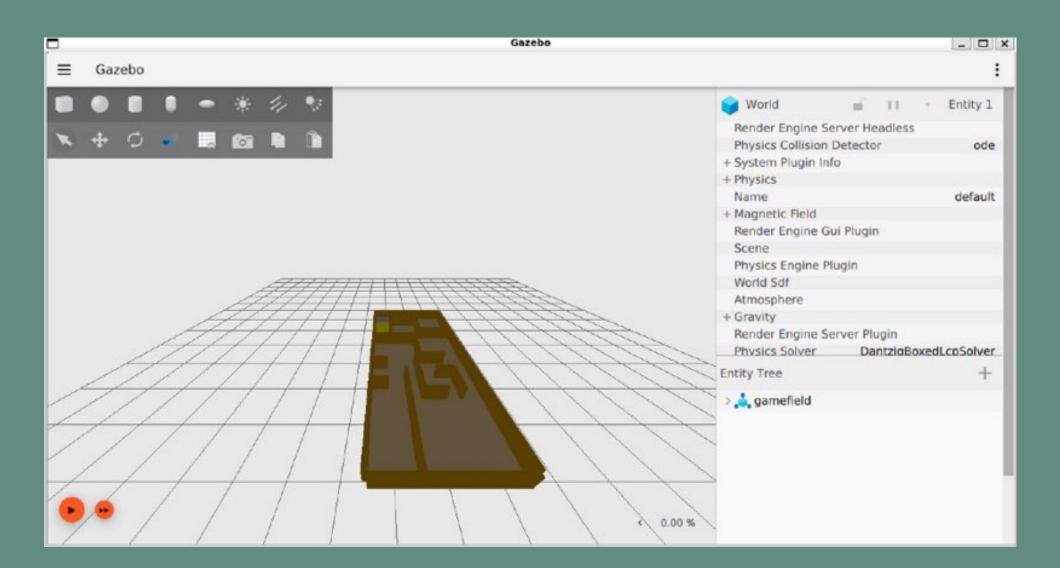
- AMCL → A localization-only system that uses a preexisting map to determine a robot's position
- Slam\_toolbox → system that creates a map and localizes the robot within it simultaneously, and can also be used for long-term mapping and updating existing maps

# SLAM\_TOOLBOX VS AMCL

In essence, AMCL needs a map to work, whereas SLAM\_Toolbox builds the map itself.

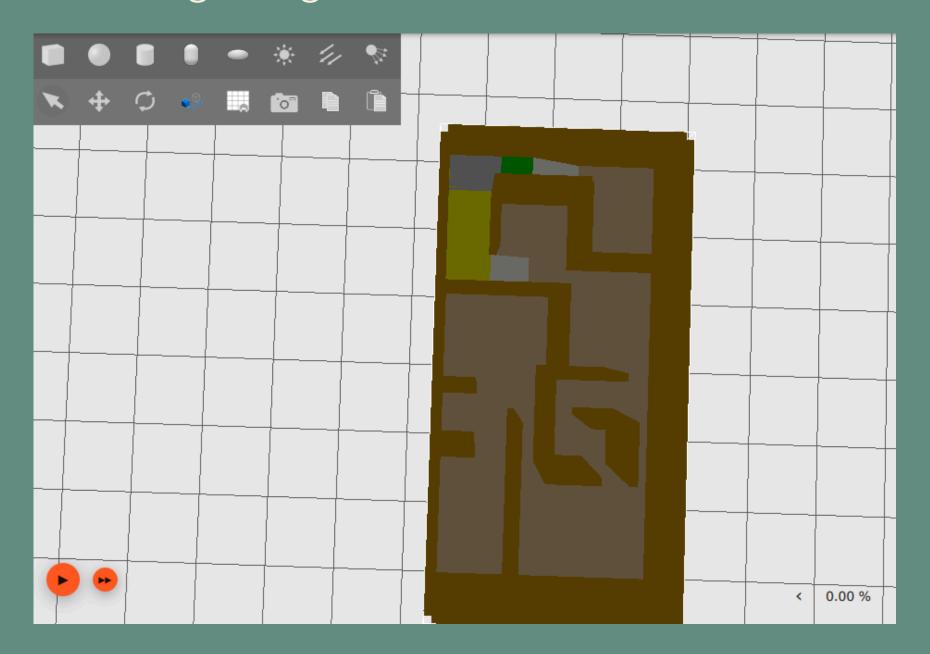
# NEXT STEPS

Sharing the gamefield world next week



# NEXT STEPS

Sharing the gamefield world next week



# RESOURCES

- Rviz Setup
- Intro to Gazebo Fortress
- Intro to SLAM
- Intro to Navigation
- Worlds in simulation
- Why simulate

# Thank You

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