

# Abhishek Sen : Documentation Report

## Robust Navigation & Trajectory Control for Autonomous Mobile Robots

### 1. Overview

This project implements a complete navigation stack for a differential drive robot (TurtleBot3) operating in cluttered, unstructured environments. Unlike standard navigation stacks that rely on heavy global costmaps, this solution utilizes a lightweight, hybrid approach: combining **B-Spline Global Trajectory Generation** with a **Reactive Potential Field Controller**.

The system is capable of smoothing coarse global waypoints into a continuous path and robustly tracking that path while dynamically avoiding unmapped obstacles (walls, construction cones, barriers) using 2D LiDAR data.

### 2. Setup and Execution

#### 2.1 Prerequisites

- **OS:** Ubuntu 22.04 LTS (Jammy Jellyfish)
- **Middleware:** ROS 2 Humble Hawksbill
- **Simulation:** Gazebo Classic 11
- **Dependencies:** python3-scipy, python3-numpy

#### 2.2 Installation

1. **Clone the repository** into your ROS 2 workspace `src` directory:

2. **Install system dependencies:**
3. **Build the package:**
4. **Configure Environment:**

## 2.3 Execution Instructions

To replicate the full "Messy World" navigation scenario, use the following three terminals:

**Terminal 1: Simulation Environment** Launches Gazebo and spawns the robot along with randomized obstacles (cones, barriers).  
Using the command : `ros2 launch turtlebot3_gazebo empty_world.launch.py`

**Terminal 2: Visualization** Launches RViz configured to show the calculated path and sensor data.

Using the command : `ros2 run rviz2 rviz2`

**Terminal 3: Navigation Controller** Starts the core node for path smoothing and control.

Using the command : `ros2 launch turtlebot3_gazebo empty_world.launch.py`

## 3. Design & Architecture

### 2.2 Algorithms and Architectural Decisions

The system is architected as a single, high-frequency ROS 2 node (RobustNavigator) operating at 20Hz. The logic is divided into three distinct pipeline stages:

#### A. Path Smoothing (Global Planner)

- **Problem:** Raw waypoints produce piecewise linear paths with sharp  $C^0$  continuity corners, causing the robot to stop and rotate in place.
- **Solution:** Implemented **B-Spline Interpolation** using `scipy.interpolate.splprep`.
- **Design Choice:** A cubic spline ( $k=3$ ) was chosen to ensure  $C^2$  continuity (continuous velocity and acceleration). A smoothing factor ( $s=0.5$ ) allows the trajectory to deviate slightly from waypoints to minimize curvature energy, resulting in more natural motion.

## B. Trajectory Tracking (Controller)

- **Algorithm: Pure Pursuit** with Dynamic Lookahead.
- **Logic:** The controller calculates the curvature required to drive the robot from its current pose to a target point on the path.
- **Optimization - Sequential Search:** To handle looping paths (where Start == End), the lookahead search is restricted to a sliding window ahead of the robot's current index. This prevents the "shortcut problem" where the robot might mistakenly lock onto the finish line immediately after starting.
  - **Control Law:**  $W = v_{\text{target}}$ ,  $w = (2v \cdot \sin(a)) / (L_d)$

Where :

$W$  = Angular Velocity ,  $v_{\text{target}}$  = Linear Velocity ,  $a$  = Heading Error (or Lookahead Angle) ,  $L_d$  = Lookahead Distance.

## C. Reactive Obstacle Avoidance (Local Planner)

- **Algorithm: Artificial Potential Fields (APF).**
- **Logic:** The robot is treated as a particle moving through a force field.
  - **Attractive Force:** Pulls the robot along the Lookahead Vector toward the global path.
  - **Repulsive Force:** Generated by LiDAR points within a defined safety\_dist (0.6m). The magnitude follows an inverse-square law ( $1/d^2$ ) creating a "stiff" virtual spring against obstacles.
- **Result:** The final velocity vector is a weighted sum of these forces, allowing the robot to smoothly deviate from the path to avoid collision and naturally merge back when clear.

## 2.3 Extension to Real Hardware (TurtleBot3)

Migrating this node from Gazebo simulation to a physical TurtleBot3 requires three key modifications:

### 1. State Estimation (Localization):

- *Simulation:* Relies on /odom (ground truth from wheel encoders).
- *Real World:* Wheel encoders drift over time. I would integrate **sensor fusion** using an Extended Kalman Filter (robot\_localization package) fusing IMU and Encoder data. For long-term navigation, I would implement **AMCL** (Monte Carlo Localization) to correct drift against a known map.

## 2. Sensor Interface:

- The `scan_callback` assumes a 360-degree perfect sensor. On a real robot, I would implement a **median filter** on the LiDAR data to remove sensor noise and spurious readings before calculating repulsive forces.

## 3. Safety & Fail-safes:

- Network latency can cause control loops to hang. I would implement a **dead-man switch** in the node: if odom or scan data is stale ( $>0.5s$ ), the robot must command zero velocity immediately.

## 2.4 AI Tools Utilization

Generative AI tools (GPT-4) were utilized to accelerate the development workflow in the following capacities:

- **Boilerplate Generation:** Rapidly scaffolding the ROS 2 class structure, publishers, and subscribers to ensure adherence to ROS 2 object-oriented best practices.
- **Mathematical Verification:** Validating the vector transformation math required to convert Global Frame coordinates (Odom) into Robot Frame coordinates (Base Link) for the Pure Pursuit curvature calculation.
- **Debugging:** Assisting in identifying the logic flaw regarding circular paths, leading to the implementation of the "Sequential Index Search" to prevent premature goal completion.

## 2.5 Extra Credit: Obstacle Avoidance Implementation

The avoidance system is not a simple "stop and wait" logic; it is a **continuous active controller**.

### Implementation Details:

1. **Safety Bubble:** A `warning_dist` (0.6m) and `critical_dist` (0.35m) are defined.
2. **Force Vectoring:**
  - The LiDAR scan is segmented into a frontal cone ( $-60^\circ$  to  $+60^\circ$ ).
  - A repulsive vector  $F_{rep}$  is calculated by summing the inverse vectors of all points in this cone.
3. **Dynamic Speed Throttling:**
  - The linear velocity  $v$  is modulated by the magnitude of the repulsive force.
  - $v_{cmd} = v_{target} * (1.0 - |F_{rep}|)$

- This ensures the robot naturally slows down when navigating narrow gaps (like the house doorways) to allow for tighter turning radii.

#### 4. Stuck Recovery:

- If the robot's Euclidean displacement is  $< 0.01\text{m}$  for 2 seconds while moving, a **Recovery Behavior** is triggered: the robot reverses linear velocity ( $-0.15\text{ m/s}$ ) and rotates to dislodge itself from local minima.

### RQT Graph

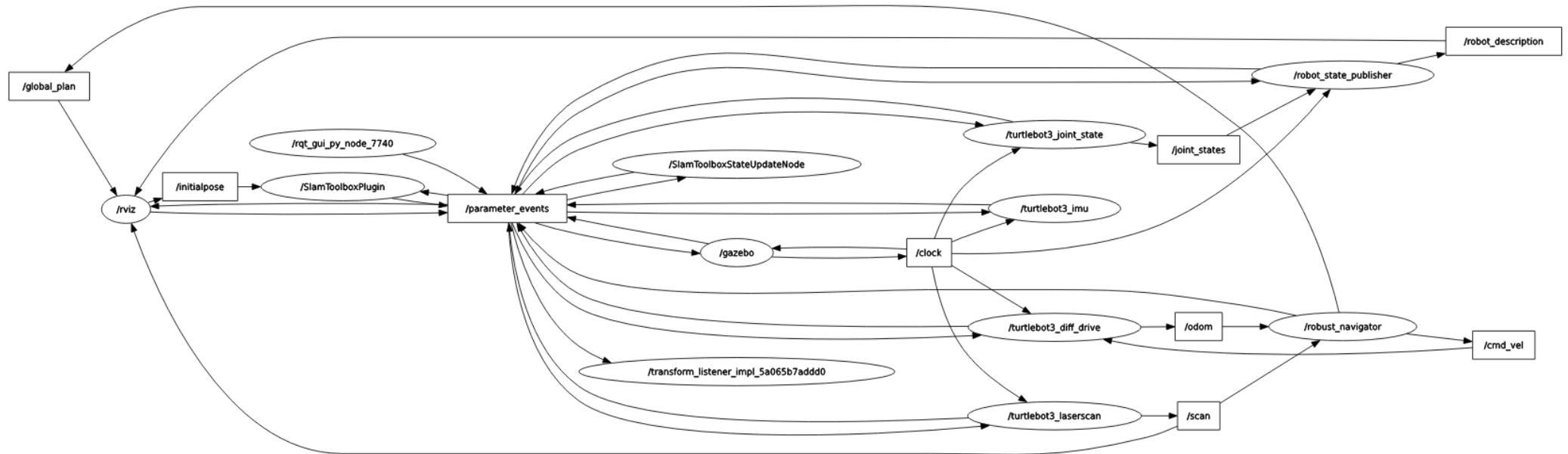


Fig :- This is the RQT graph