

# *Stereo Visual SLAM System in a CARLA Simulation Environment*

*Integration of Stereo Vision, IMU Fusion & 3D Mapping*

*ECE 544 - Mobile Robots  
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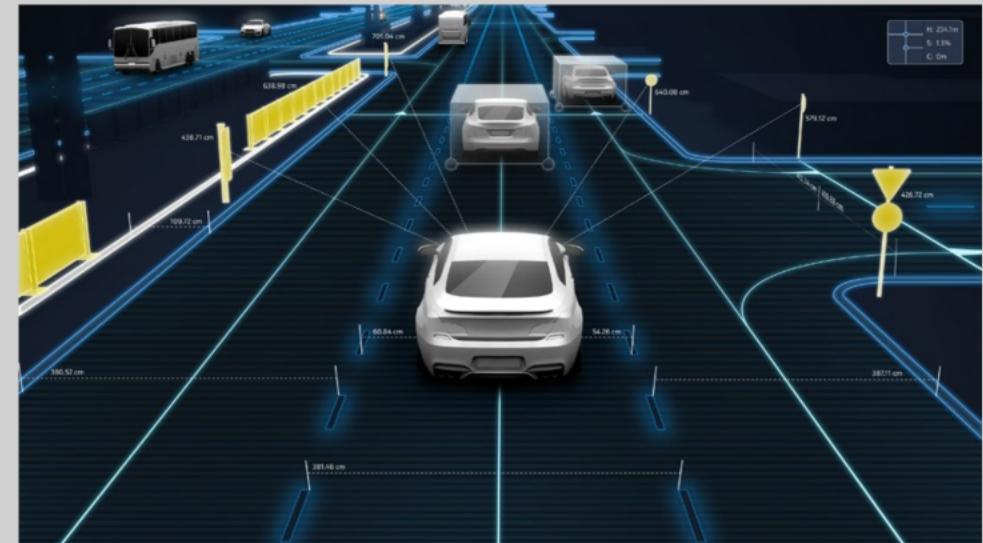
# Introduction

*Visual SLAM is the process of building a map and estimating trajectory using camera input*

*This project implements a full Stereo Visual SLAM pipeline with:*

- Depth estimation from stereo
- IMU integration using a 9D Kalman Filter
- Pose estimation via PnP
- Loop closure detection

*Simulated in CARLA using a self-driving vehicle in urban scenes*



*Key tools & libraries:*

- Python, OpenCV, NumPy, CARLA Simulator, Open3D

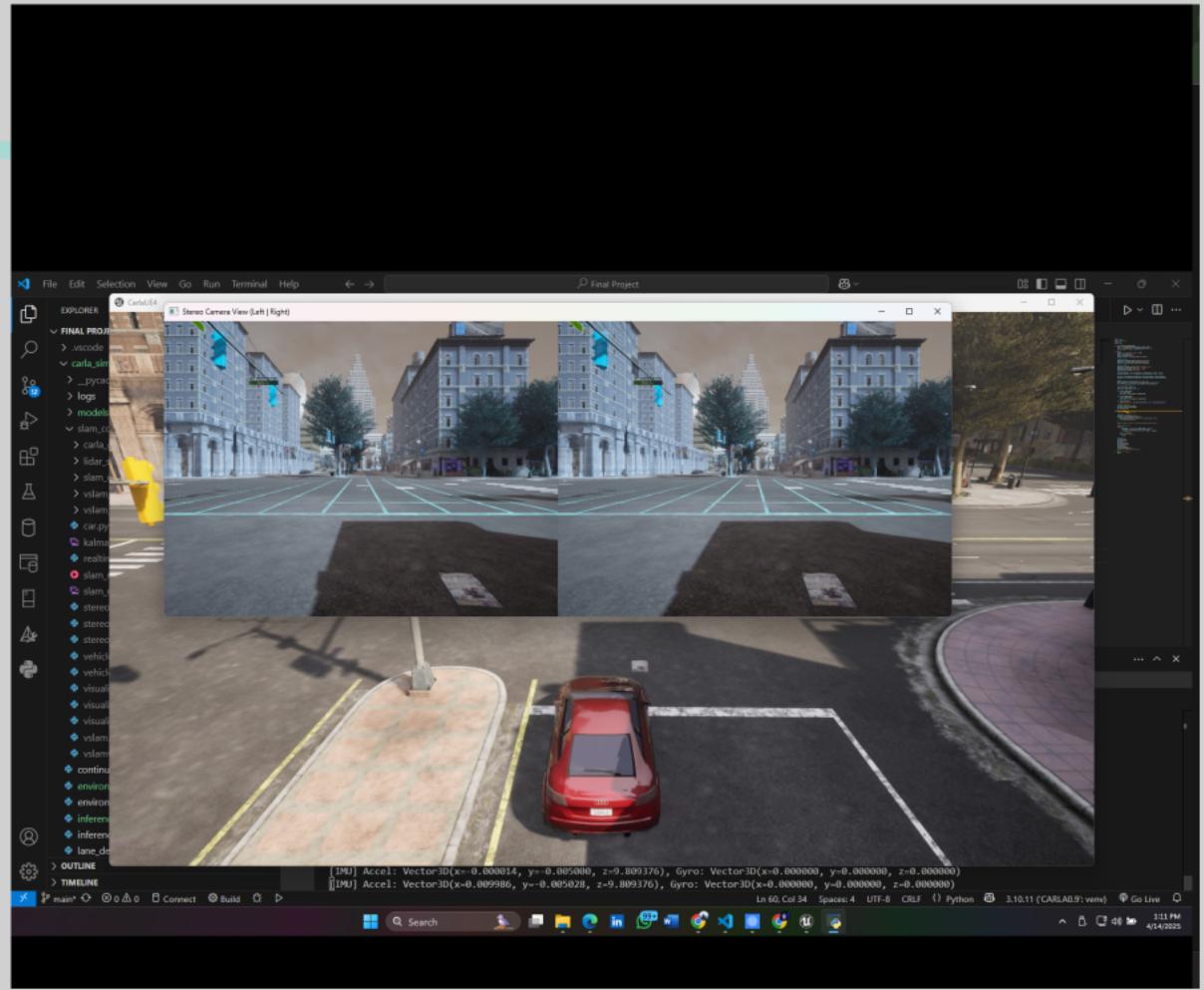
# Vehicle Sensor Configuration in CARLA

## Vehicle Setup:

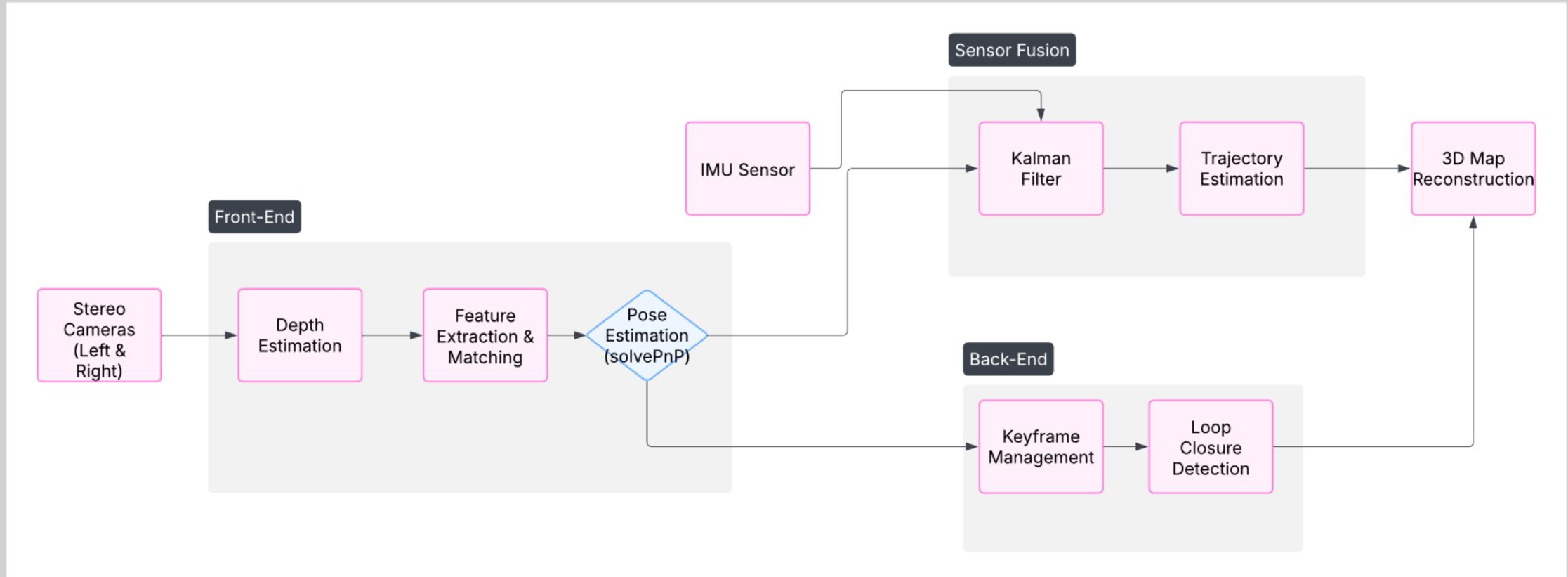
- Spawned a standard CARLA vehicle
- Enabled built-in autopilot

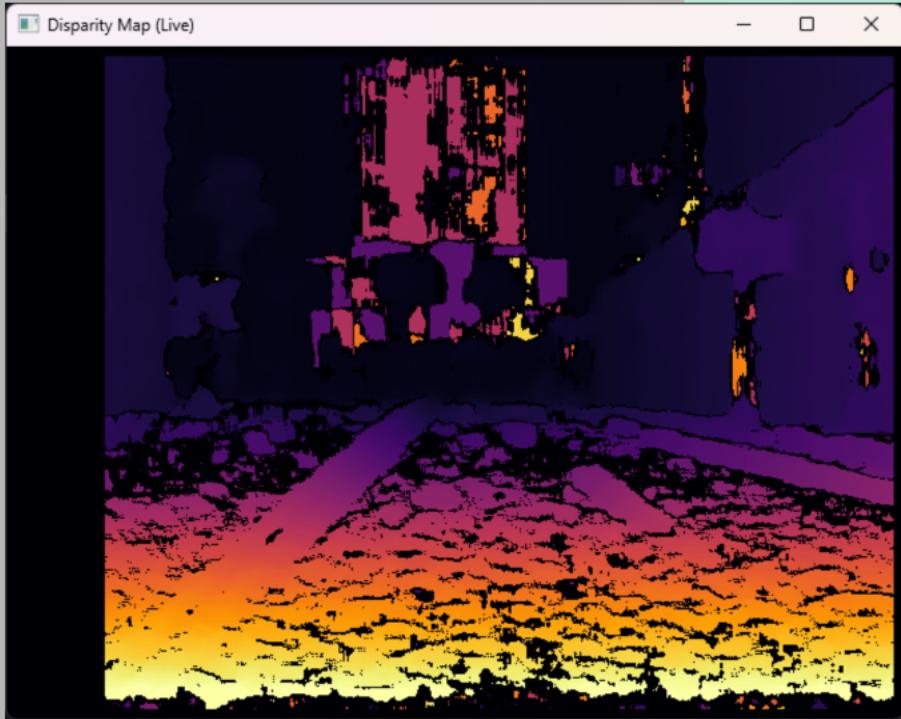
## Attached Sensors:

- Left RGB Camera
  - Resolution:  $640 \times 480$
  - Position: Front-left, focal baseline = 0.4 m
- Right RGB Camera
  - Synchronized with left for stereo vision
- IMU Sensor
  - Captures 3-axis acceleration and angular velocity
  - Used for Kalman Filter prediction



# System Architecture





# *Depth Estimation Methods*

*Depth estimation is achieved using StereoSGBM (Semi-Global Block Matching), which computes disparity maps from stereo images. The depth is calculated by the formula: Depth = (focal length \* baseline) / disparity, allowing for accurate real-time calculations essential for spatial understanding.*

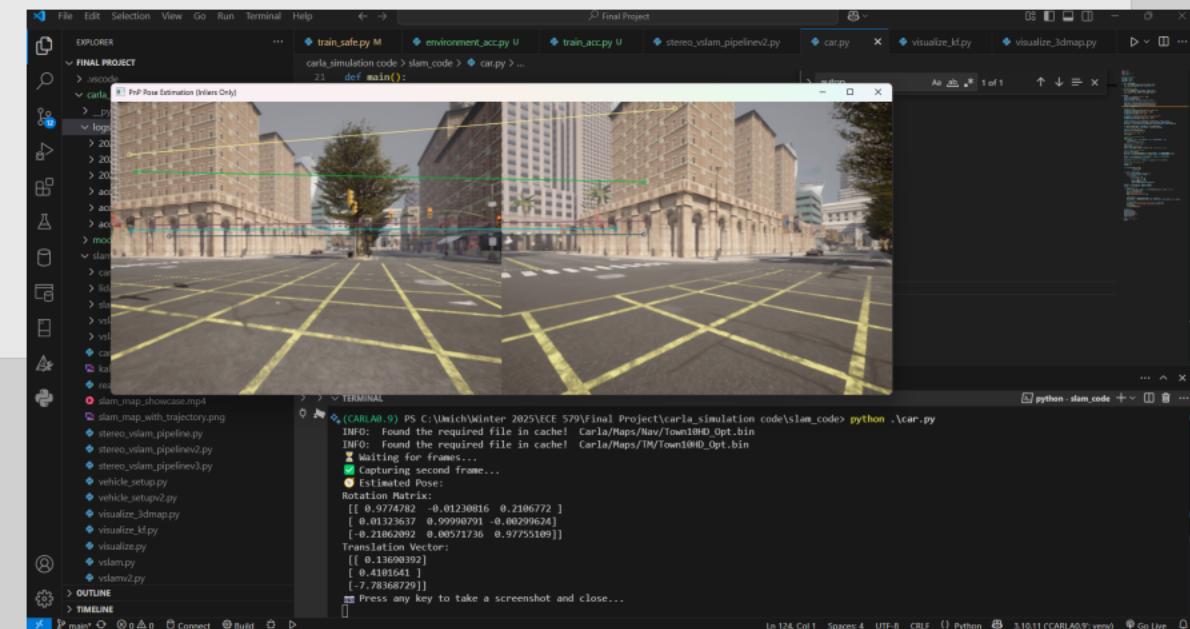
# *Feature Detection and Matching Algorithms*

- Keypoints extracted using *SIFT* and *ORB*
- Matched using *FLANN-based matcher* + *KNN ratio test*
- Ensures consistent tracking of features between frames
- Robust against scale and rotation changes



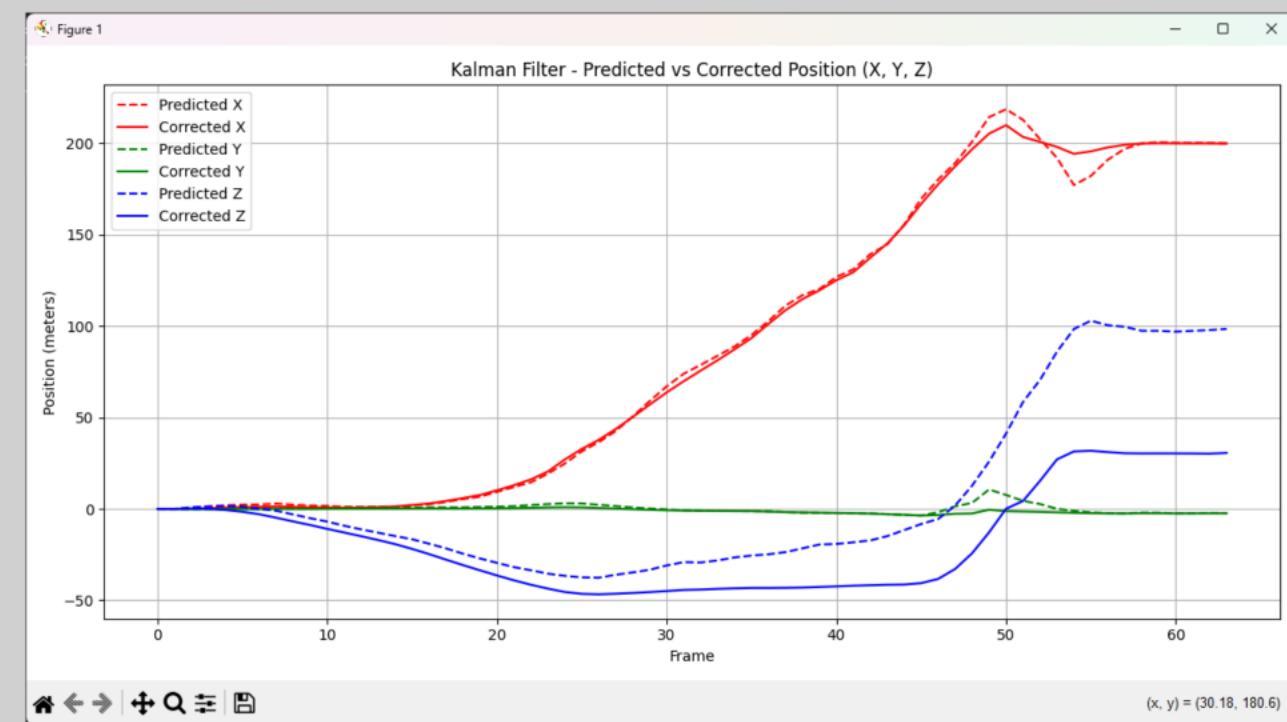
# Pose Estimation using PnP

- Matched 3D-2D correspondences across frames
- Depth from stereo used to recover 3D points
- Pose computed using PnP with RANSAC
- Returns rotation ( $R$ ) and translation ( $t$ ) of the current camera
- Used to update trajectory in real-world coordinates



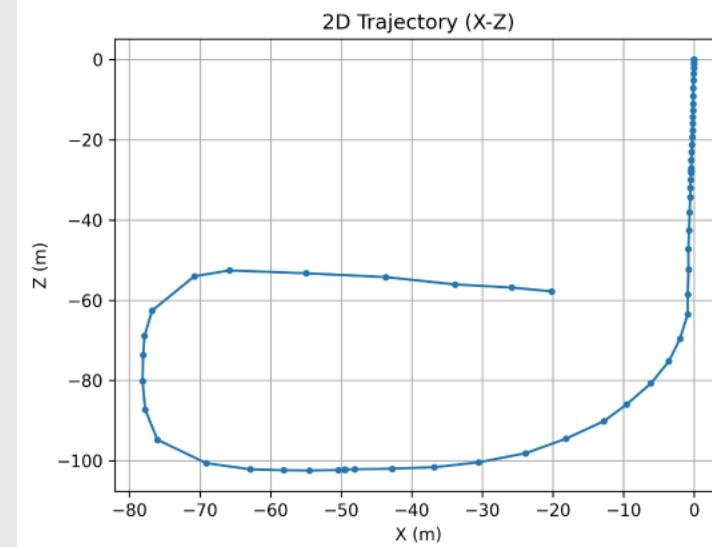
# Kalman Filter Implementation

- 9D Kalman Filter fuses:
  - Linear acceleration (IMU)
  - Angular velocity (IMU)
  - Visual pose estimates (SLAM)
- Predicts position, velocity, orientation in real time
- Visual data used to correct drift from inertial prediction



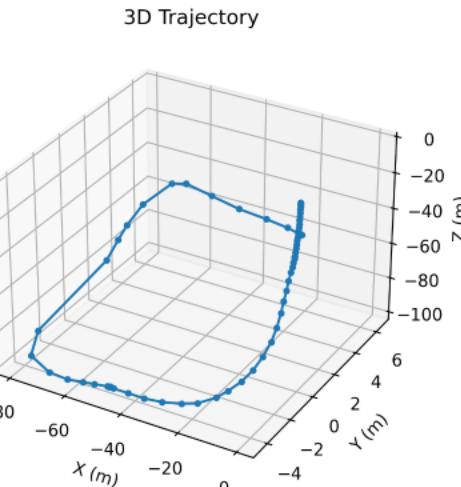
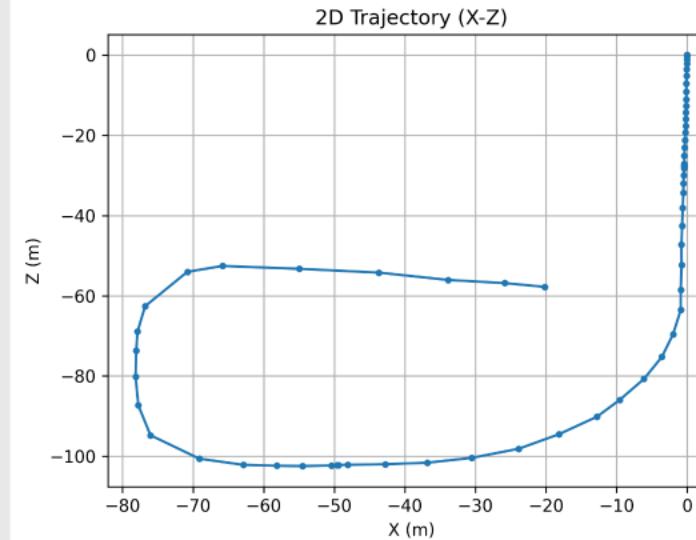
# *Keyframe Management and Loop Closure*

- Keyframes stored during significant motion
- Each new keyframe's descriptor is compared with all past keyframes
- If >30 good matches are found ; loop closure is flagged
- Enables drift detection when re-visiting previous locations
- Currently: loop is detected but not yet used for map correction

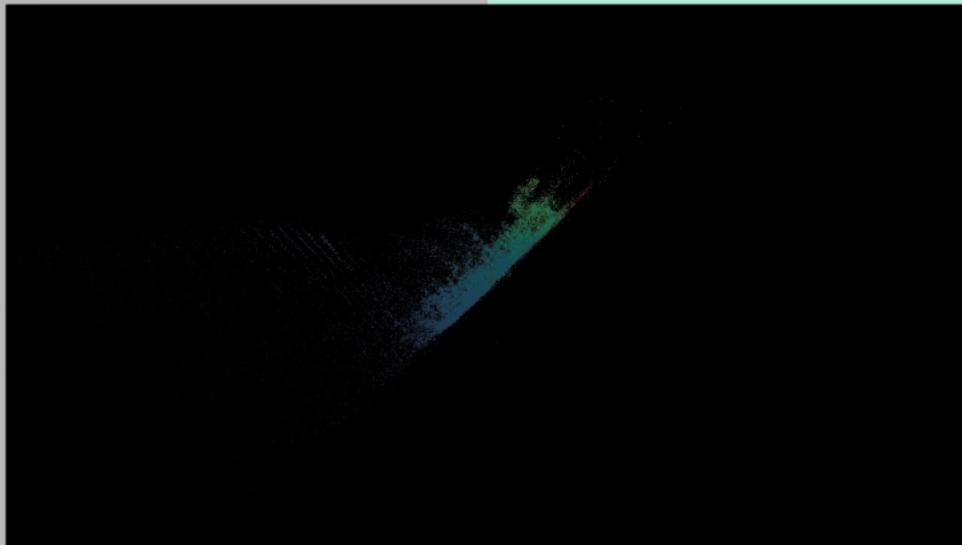


# Visualization of 2D and 3D Trajectories

- 2D trajectory shows lateral pose drift
- 3D trajectory confirms correct height tracking
- Consistent movement even without global optimization



# *3D Point Cloud Mapping*



- *Mapped environment using Open3D*
- *Colors indicate Z-height for clarity*
- *Data downsampled and cleaned for visualization.*



*Thank You!*