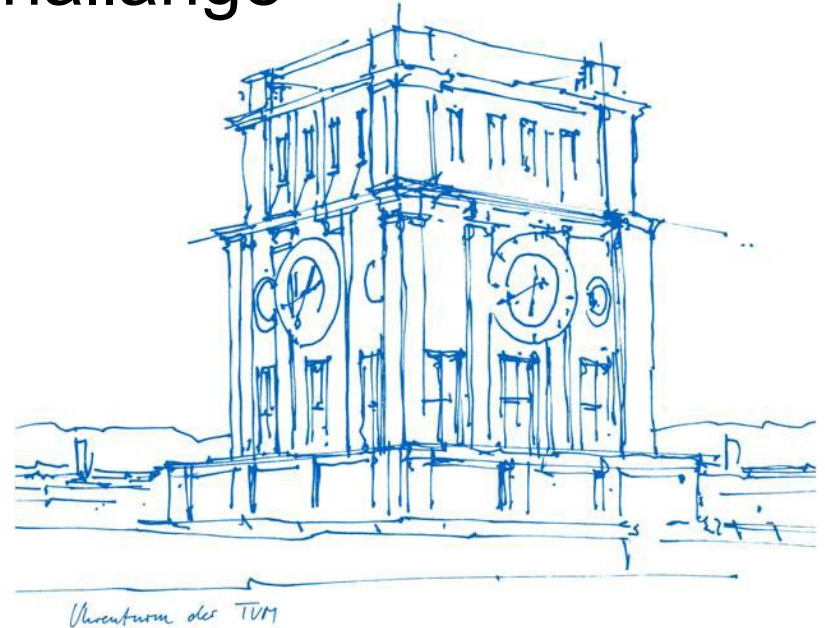


# Autonomous System: Group Project: Sub-Terrain Challenge

Team 12

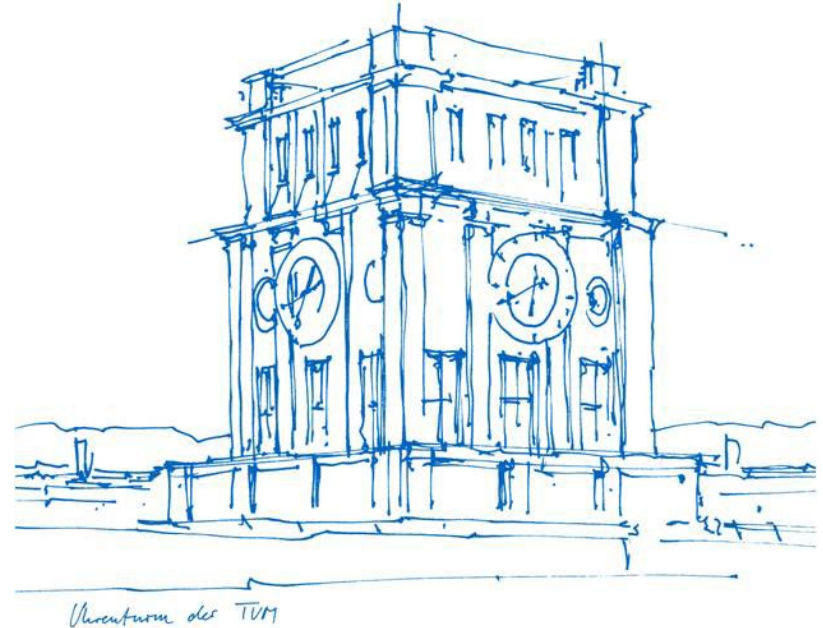
Members: Xuanyi Lin, Jiaxiang Yang, Enze Wang,  
Yicun Song, Mohammadali Rahmati

Munich, 04. March 2025



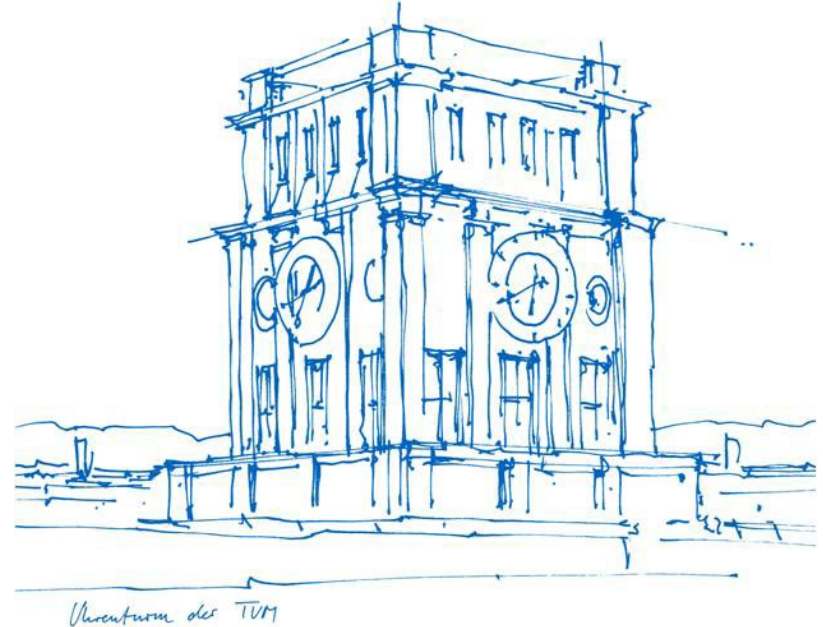
# Content

- Project Overview
- Packages
  - state\_machine\_pkg
  - Light\_detection\_pkg
  - mapping\_pkg
  - pathplanning\_pkg
  - trajectory\_generator
- Conclusion

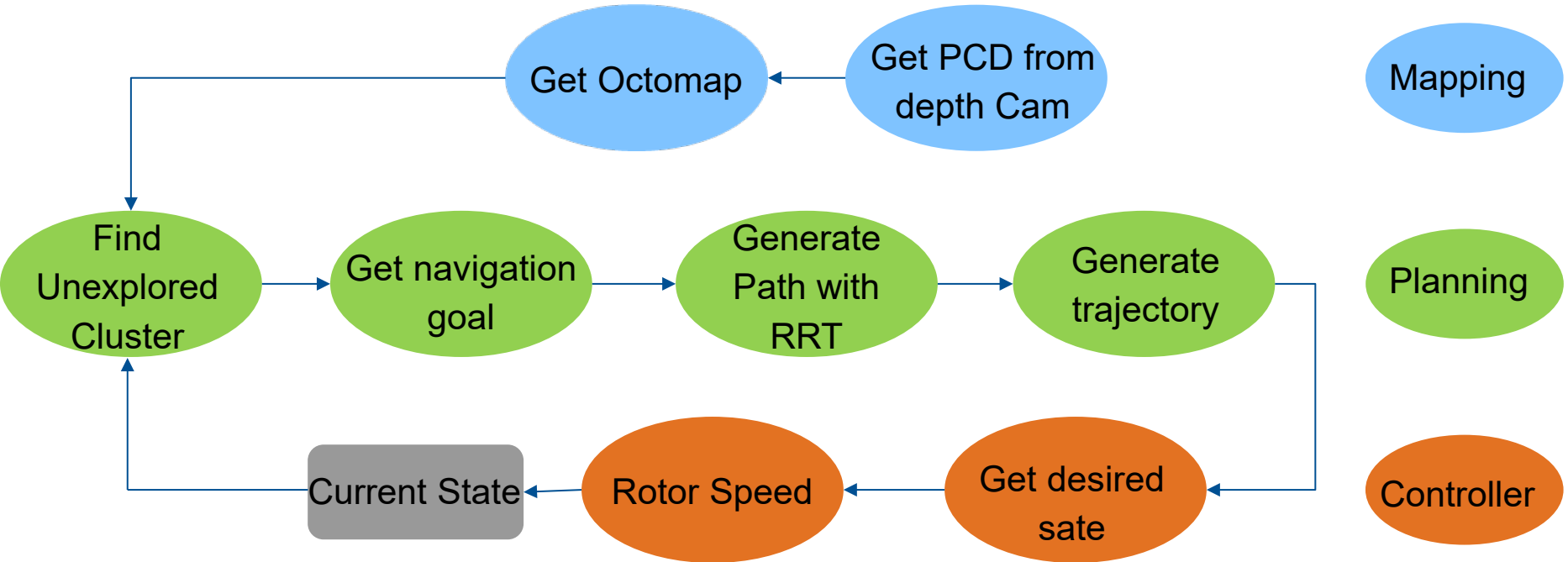


# Project Overview

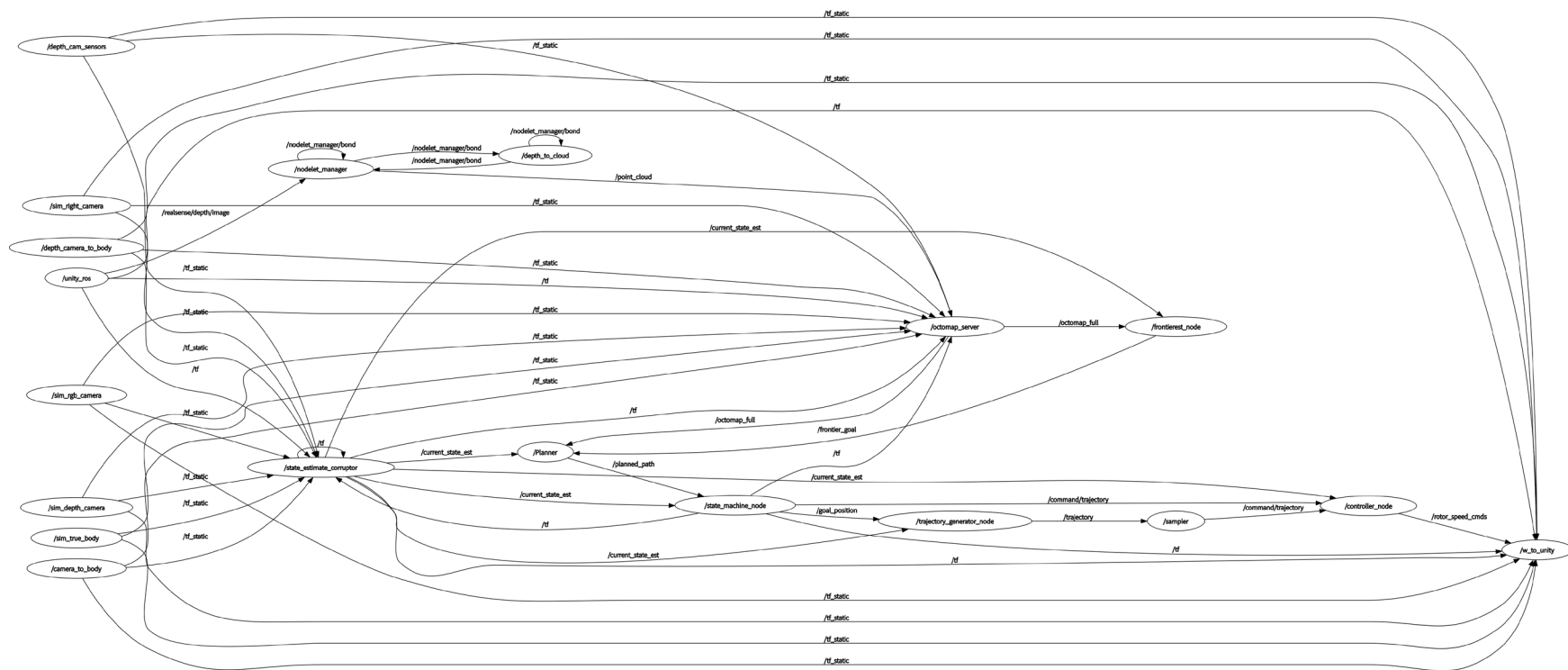
- Goal: Autonomous system operates in a cave environment
- Object detection, environment modeling, path planning
- Tech Stack: ROS + Unity Simulation
- Computer Vision, Robot dynamic, Path Planning



# System Architecture



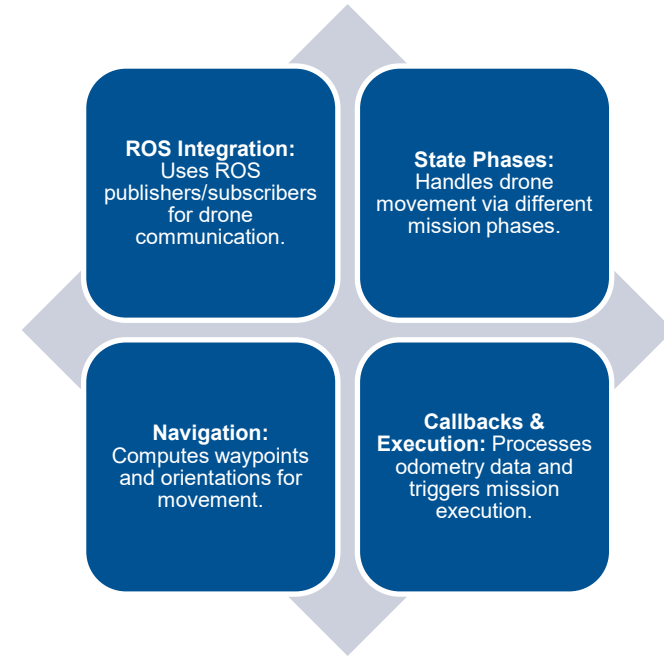
# RQT Graph



# State\_machine\_pkg

## Purpose and Key Components

- Manages a drone's mission using a structured state machine.
- Receives path data, sets goal waypoints, and executes sequential movements.
- **State Phases:**
  1. Takeoff
  2. Transit to Cave
  3. Hovering
  4. Descent
  5. Rotation
  6. Straight Movement



# State\_machine\_pkg

## Workflow and Execution Flow

### Initialization:

- Define **goal points** for navigation.
- Subscribe to **odometry & path updates**.
- Publish **desired drone states**.

### Mission Execution:

1. Fetch **next waypoint**.
2. Compute **yaw** for orientation.
3. Publish **goal position**.
4. Monitor **target fulfillment**.
5. Transition to **next phase**.

### Outcome:

- Enables autonomous drone navigation through defined waypoints.
- Ensures **smooth transitions** between mission states.
- Uses **ROS messaging** for real-time control.

## Semantic Mask Node

**Goal:** Identify the areas where lights are present and filter out everything else.

### 1.Receives Images:

1. It subscribes to two image topics:
  1. A **semantic image** (color-coded image showing objects detected by a model).
  2. A **depth image** (providing distance information for each pixel).

### 2.Extracts Light Mask:

1. It processes the **semantic image** to detect lights.
2. It creates a **mask** by filtering pixels of a specific color (representing lights).
3. This mask is applied to the **depth image**, keeping only the depth values where lights exist.

### 3.Publishes a Filtered Depth Image:

1. The modified depth image is sent as a **masked depth image**, which contains only light-related depth information.
2. This image is published so the second node can process it.



# Light\_detection\_pkg

## Light Detector Node

**Goal:** Convert the filtered depth image into 3D points and detect the location of the lights.

### 1.Receives the Masked Depth Image:

1. It subscribes to the **masked depth image** from the first node.

### 2.Converts Depth Image to 3D Points:

1. It processes the depth image to create a **Point Cloud** (a 3D representation of detected lights).
2. It transforms this into world coordinates (real-world positions).

### 3.Finds the Center of Lights:

1. It calculates the **center** of the detected light sources in 3D space.

### 4.Filters Out Duplicates:

1. It checks if this light was already detected before (to avoid duplicates).

### 5.Publishes the Light's 3D Position:

1. It publishes the **position of the light** as a PointStamped message.
2. This can be used by other systems (e.g., a robot that needs to navigate using detected lights).

# Mapping Package

## Key Optimization Strategies

### 3D Point Cloud Generation (depth\_image\_proc)

- Converts depth images into 3D point clouds for environment mapping.
- High-precision spatial representation aids navigation.

### Voxel Grid Mapping (OctoMap)

- Converts point clouds into occupancy grids for dynamic path planning.
- Supports real-time environment updates and obstacle avoidance.

### Semantic Camera for Object Detection

- Identifies key objects (e.g., lights) with reduced false positives.
- Provides semantic-level information to support path planning.

# Pathplanning Package

Efficient autonomous exploration in unknown environments by identifying boundary regions.

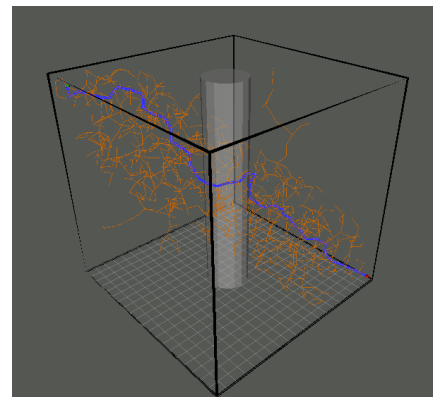
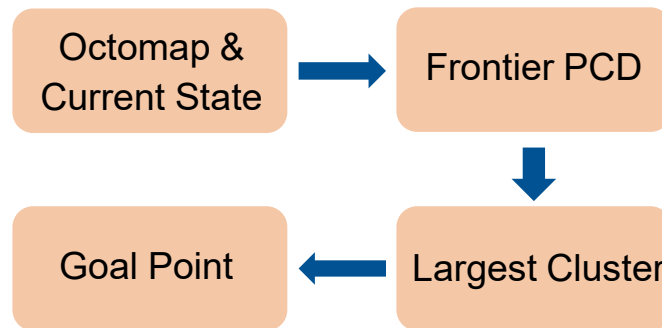
## Path Planning Workflow:

### 1. Identify Boundaries & Compute Target Point

- Use OctoMap occupancy grid and current state to detect the unexplored area's boundary.

### 2. Path Planning with RRT

- Set UAV's current position as the start and boundary center as the goal.
- Randomly sample feasible points, expanding the RRT tree while ensuring obstacle avoidance.



# Trajectory Generator

## Trajectory Generation with `mav_trajectory_generation`



### Objective:

- Generate smooth, optimized UAV trajectories using **`mav_trajectory_generation`**.
- Ensure feasible paths that respect velocity = 30 m/s and acceleration constraints =  $10 \text{ m/s}^2$ .

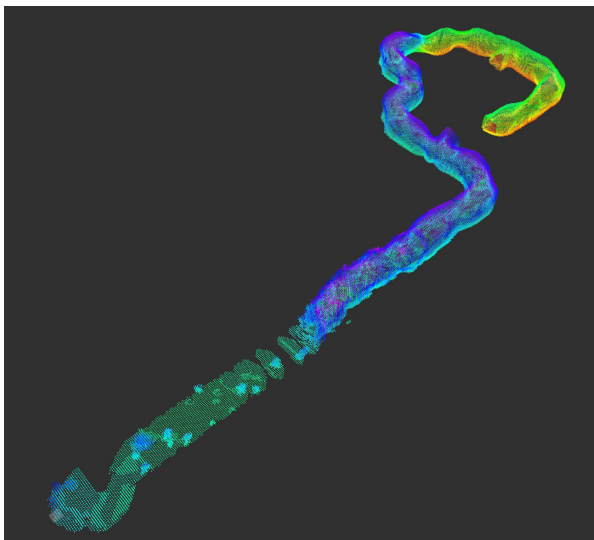
### Key Steps:

1. **Define Waypoints:** Specify UAV positions and constraints.  
goal\_position (from pathplanning) current position (from the state\_estimate\_corruptor).
2. **Set Optimization Parameters:** Choose polynomial degree (N=10) and derivative (order = 4/SNAP) for optimization.
3. **Optimize Trajectory:** Solve for a **minimum-snap** trajectory using nonlinear methods.
4. **Sample & Visualize Trajectory:** Extract trajectory points and visualize in **RViz**.

# Conclusion

We have completed the core parts expected in the assignment.

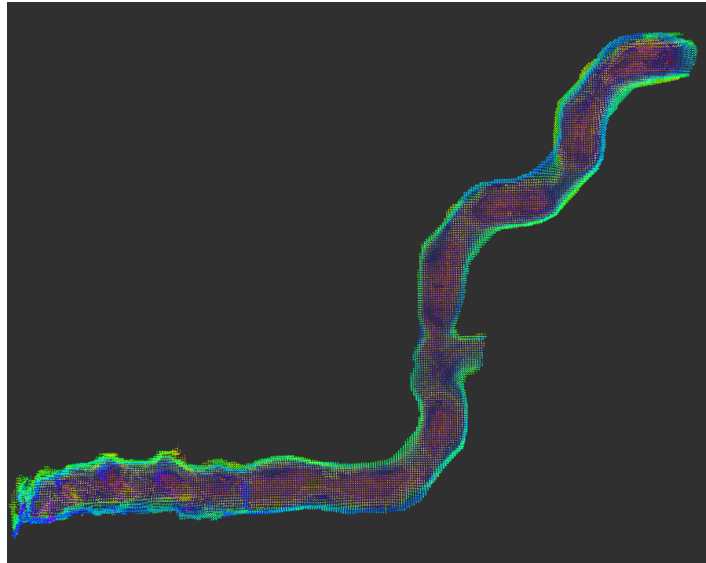
- Successfully working perception pipeline
- Successfully working path and trajectory planning
- Successfully arriving at the entrance of the cave
- Successfully finding two objects of interest
- Successfully building a voxel-grid/mesh representation of the cave environment



# Conclusion

But we also have some details should have to be completed

- Sometime the drone can not build the full voxel-grid/mesh representation of the cave environment
- We can only find two objects of interest
- The drone have a low speed in the cave to build the voxel-grid/mesh representation.



Thank you very much for your patience