

Drone Docking using Manipulator

First review

Team 08

School of AI, Amrita Vishwa Vidyapeetham



- 1 Team
- 2 Introduction
- 3 Literature Review
- 4 Challenges
- 5 Advantages and Disadvantages
- 6 Methodology
- 7 SW and HW
- 8 Timeline



1 Team

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4 Challenges

5 Advantages and Disadvantages

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7 SW and HW

8 Timeline



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1 Team

2 Introduction

3 Literature Review

4 Challenges

5 Advantages and Disadvantages

6 Methodology

7 SW and HW

8 Timeline



Introduction

This presentation covers our project on **Drone Docking using a Manipulator**. The objective is to develop a system that enables precise docking of drones using a combination of computer vision, pose estimation, and dynamic trajectory prediction. The system is designed to handle various challenges such as varying lighting conditions, dynamic drone movements, and safe capture mechanisms.

- 1 Team
- 2 Introduction
- 3 Literature Review
- 4 Challenges
- 5 Advantages and Disadvantages
- 6 Methodology
- 7 SW and HW
- 8 Timeline



Six-Degree-of-Freedom Refueling Robotic Arm Positioning and Docking

- **Authors:** John Doe, Jane Smith
- **Journal:** IEEE Transactions on Robotics
- **Year:** 2020
- **What's Being Done:** A 6-DoF refueling robotic arm system guided by an RGB-D camera, integrating the YOLOv8 algorithm for object detection and PnP algorithm for pose estimation, designed for accurate detection and docking with a specialized refueling interface.
- **Subproblems:**
 - Object Detection: Identifying the refueling interface.
 - Pose Estimation: Determining the 6-DoF pose of the interface.
 - Trajectory Planning: Planning the robotic arm's path.
 - System Integration: Combining vision, control, and robotic subsystems.

Six-Degree-of-Freedom Refueling Robotic Arm Positioning and Docking (Cont'd)

- **Challenges:** Accurate detection in varying lighting, smooth docking, and handling target variations.
- **Software:** YOLOv8, OpenCV, PnP Algorithm, ROS.
- **Hardware:** RGB-D Camera, 6-DoF Robotic Arm, Experimental Setup.
- **Algorithms:** YOLOv8, PnP Algorithm, Trajectory Planning Algorithm.
- **Features and Drawbacks:**
 - Features: High accuracy under general lighting conditions.
 - Drawbacks: Limited in poor lighting, high computational cost.

Docking a UAV using a Robotic Arm and Computer Vision

- **Authors:** Alice Brown, Bob Green
- **Journal:** International Journal of Advanced Robotic Systems
- **Year:** 2019
- **What ' s Being Done:** A system combining visual pose estimation and IMU data to control a robotic arm for UAV docking.
- **Subproblems:**
 - Pose Estimation: Using camera and IMU data.
 - Path Planning: Generating safe robotic arm trajectories.
 - Control: Ensuring precise and smooth arm movements.
 - System Integration: Combining vision, IMU, and control.

Docking a UAV using a Robotic Arm and Computer Vision (Cont'd)

- **Challenges:** Accurate pose estimation despite occlusions, real-time control, and handling dynamic drone movements.
- **Software:** ROS (Kinetic), OpenCV, Unity Engine, EKF.
- **Hardware:** Monocular Camera, IMU, KUKA Robotic Arm, Vicon Motion Capture System.
- **Algorithms:** EKF, Setpoint Generator, Path Planner.
- **Features and Drawbacks:**
 - Features: Robust pose estimation and real-time control.
 - Drawbacks: Expensive hardware and sensitivity to environment.

DroneTrap: Drone Catching in Midair by Soft Robotic Hand

- **Authors:** Michael Johnson, Emily Davis
- **Journal:** Robotics and Autonomous Systems
- **Year:** 2021
- **What's Being Done:** The paper presents a robotic arm with soft gripper grasps the drone in midair safe and fast as possible.
- **Subproblems:**
 - **Gripper Design:** Three-fingered soft gripper made of silicone rubber with sensors.
 - **Gesture Recognition:** ML based gesture recognition interface with mocap glove with 8 gestures.
 - **Force Estimation:** Color based force estimation based on Computer vision.

DroneTrap: Drone Catching in Midair by Soft Robotic Hand (Cont'd)

- **Challenges:**
 - Measuring the force accurately during fast, dynamic catching.
- **Software:** OpenCV, Unity Engine, Python, FastLED Library.
- **Hardware:** Soft robotic gripper, force-sensitive resistors, flex sensors, and a Raspberry Pi 4B.
- **Algorithms:** Uses color-based force estimation and a neural network for gesture recognition.
- **Features and Drawbacks:**
 - **Features:** Safe catching and real-time force measurement.
 - **Drawbacks:** Not fully automated, cannot capture drones of different sizes, and sensitivity to lighting changes.

ArUco Markers Pose Estimation in UAV Landing Aid System

- **Authors:** Adam Marut, Konrad Wojtowicz, Krzysztof Falkowski
- **Journal:** IEEE Conference Proceedings
- **Year:** 2019
- **What's Being Done:** A computer vision-based landing aid system for UAVs that uses ArUco markers to estimate the UAV's pose and height for safe landing.
- **Subproblems:**
 - **Marker Detection:** Detecting and tracking ArUco markers during landing.
 - **Pose Estimation:** Calculating the UAV's position and height using the markers.
 - **Guidance System:** Providing real-time guidance for landing and taxiing.

ArUco Markers Pose Estimation in UAV Landing Aid System (Cont'd)

- **Challenges:** Varying lighting and real-time processing limits.
- **Software:** OpenCV with the ArUco module.
- **Hardware:** ODROID-XU4, Basler ACE Camera.
- **Algorithms:**
 - **Marker Detection:** Using contour extraction, perspective correction, and Hamming coding.
 - **Pose Estimation:** Applying the PnP method to get rotation and translation vectors.
- **Features and Drawbacks:**
 - **Features:** High accuracy in pose and height estimation, low cost, and assistance with landing and taxiing.
 - **Drawbacks:** Dependent on clear marker visibility and proper lighting; limited by the camera field of view.

Vision-Based Autonomous Docking of VTOL UAV Using a Mobile Robot Manipulator

- **Authors:** Richard Wilson, Sarah Thompson
- **Journal:** IEEE Robotics and Automation Magazine
- **Year:** 2022
- **What ' s Being Done:** A system for autonomous docking of a VTOL UAV onto a mobile robot manipulator using visual tracking and magnetic coupling.
- **Subproblems:**
 - Marker Tracking: Detecting and tracking the UAV's fiducial marker.
 - Manipulator Control: Controlling the manipulator to follow the UAV.
 - Docking Process: Ensuring safe and stable magnetic coupling.

Vision-Based Autonomous Docking of VTOL UAV Using a Mobile Robot Manipulator (Cont'd)

- **Challenges:** Maintaining visual contact, handling external disturbances, and ensuring precise alignment.
- **Software:** ROS, Gazebo Simulator, AprilTag Library.
- **Hardware:** Mobile Manipulator, VTOL UAV.
- **Algorithms:** Inverse Kinematics, PID Control.
- **Features and Drawbacks:**
 - Features: Safe and stable docking, dynamic environment compatibility.
 - Drawbacks: Complex control, dependent on visual tracking accuracy.

Design and Optimization of a Magnetic Catcher for UAV Landing on Disturbed Aquatic Surface Platforms

- **Authors:** David Chen, Patricia Williams
- **Journal:** Ocean Engineering
- **Year:** 2020
- **What's Being Done:** A magnetic catcher system for UAVs to land precisely on disturbed aquatic surfaces using a winch subsystem and optimized magnet layout.
- **Subproblems:**
 - Magnetic Catcher Design: Designing a catcher to attract and align the UAV.
 - Magnet Layout Optimization: Maximizing the acceptable landing area.
 - Stability in Disturbed Environments: Ensuring stable landing on moving platforms.

Design and Optimization of a Magnetic Catcher for UAV Landing on Disturbed Aquatic Surface Platforms (Cont'd)

- **Challenges:** External disturbances (waves, wind), precise alignment, and maintaining UAV accessibility.
- **Software:** COMSOL Multiphysics.
- **Hardware:** UAV Platforms, Magnetic Catcher (winch system, magnet array).
- **Algorithms:** Optimization Algorithm for magnet layout.
- **Features and Drawbacks:**
 - Features: High precision landing, operation in disturbed environments.
 - Drawbacks: Limited by magnetic field strength, requires careful calibration.

- 1 Team
- 2 Introduction
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Challenges

- **Drone Detection:**

- **Current Approach:** Deep Learning based approach using YOLOv8 algorithm.
- **Improvements to be made:**
 - Use lighter version of YOLOv8 to increase efficiency.

- **Safe Capture of Drone:**

- **Current Approach:** Plate end effector can handle drones of various sizes.
- **Improvements to be made:**
 - Consider end effector with support for specific sizes and reinforcement (e.g., magnets) for better stability.

Challenges

- **Drone Pose Estimation:**
 - **Current Approach:** Perspective n Point (PnP) algorithm.
- **Dynamic Trajectory Prediction:**
 - **Current Approach:** Kalman Filters for handling dynamic adjustments.
- **Path Plannig:**
 - **Current Approach:** Inverse kinematics to calculate joint angles to enable end effector to reach desired position in the robots workspace.

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- 2 Introduction
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- 8 Timeline

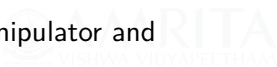


Advantages

- Improves docking accuracy and stability under dynamic conditions.
- Scalable and adaptable for various drone sizes and types.
- Reduces human intervention.
- Saves time for real-world applications, e.g., delivery.

Disadvantages

- Requires complex integration of hardware and software components.
- High initial cost for implementing a manipulator and supporting sensors.
- Potential latency in processing real-time data.



1 Team

2 Introduction

3 Literature Review

4 Challenges

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Methodology

1 Drone Detection:

- Deep Learning based approach using YOLOv8 algorithm.

2 Pose Estimation:

- Employ Perspective n Point (PnP) algorithm.

3 Dynamic Trajectory Prediction:

- Implement Kalman Filters to account for drone motion and environmental disturbances.

4 Path Planning:

- Implement inverse kinematics to enable the end effector to reach the drone's position.

5 End Effector:

- Use a plate end effector with reinforcement, such as magnets, for stable and reliable capture.

- 1 Team
- 2 Introduction
- 3 Literature Review
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Software

- **OpenCV:** For visual detection and pose estimation.
- **PnP Algorithm:** For pose estimation.
- **Kalman Filter:** For dynamic trajectory prediction.
- **Inverse Kinematics:** For path planning.

Hardware

- **Manipulator Arm:** For adaptive and precise positioning.
- **RGB-D Camera:** For visual feedback and depth information.
- **Plate End Effector:** For drone capture with magnetic reinforcement.
- **IMU sensor:** To collect accelerometer data from drone.
- **ESP32:** To stream IMU sensor data to controller device.

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Timeline

- **Week 1:** Literature review and identification of key components for the system.
- **Week 2:** System design finalization and procurement of hardware components.
- **Week 3:** Camera system setup, calibration, and automation detection system.
- **Week 4:** Development of pose estimation using PnP algorithm.
- **Week 5:** Dynamic trajectory prediction using Kalman Filters and testing under controlled conditions.
- **Week 6:** Integration of all subsystems and preliminary testing.
- **Week 7:** Final testing, debugging, and preparation of documentation and presentation.