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Abstract: ...

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1. INTRODUCTION

Multi-agent systems (MASs) are nowadays used in a variety of applications from as they can perform complex tasks, which would otherwise impossible for a single robot. This is even more the case when considering heterogeneous fleet where the strenghts of one type of robot can compensate for the limitations of another. For example, to monitor a large area, multiple ground agents can move from one point to another being supervised from above by a drone with a wide field of view. To deepen the subject and understand both the potential tasks and challenges of MASs, the reader can refer to the work of Maldonado et al. (2024).

One of the fundamental cooperative tasks when working with multi-agents is formation control. This topic has been studied for years. For example, already Lee and Chong (2009) already proposed a decentralized control algorithm for a team of two-wheeled robots to achieve achieve a geometric pattern. However, formation control remains an hot challenging topic to this day. Tran et al. (2021) experimentally validated a robust distributed control based on negative imaginary systems consensus theory, using both ground robots and an air-ground fleet. In the work of Güler and İsa E. Yıldırım (2023), each agent computes its control action in a leader-follower manner using local extended Kalman filter's estimates to achieve a desired shape together with other wheeled robots and a drone. An optimal distributed formation control algorithm for double-integrator multi-agents has been presented and validated through simulations by Huang et al. (2023). Aditya and Werner (2023) defined the formation problem for a group of double-integrator agents as a discrete-time game whose solution is given by a state-dependent Riccati equation. A robust distributed consensus controller is presented by Restrepo et al. (2023) to address the rendezvous problem and applied in simulation to a swarm of drones. In the recent years, reinforcement learning has used in many areas including formation control. Wang et al. (2020) addressed multi-particle formation control by combining graph attention networks and multiple long short-term

memories to achieve the desired shape and avoid collisions respectively. A position and an orientation robust controllers based on reinforcement learning are proposed and validated through simulations by Zhong et al. (2025) for the formation of a term of quadrotors.

In this context is placed this work which proposes a novel formation control scheme for a team consisting of a quadcopter and two unicycles, as show in Fig. **put image**. when addressing the multi-agent formation, there are two main challenges to solve: achieving the desired shape and avoiding collisions. On one hand, for the first challenge, a distributed controller combining Feedback Linearization (FL) and an optimal linear controller based on Linear Matrix Inequalities (LMIs) is presented. LMIs are a powerful mathematical tool used in many applications among which also formation control. In the work of Trejo et al. (2023), an LMI is solved to design the controller and observer gains, guaranteeing stable formation flight for a group of quadcopters. Deshpande et al. (2011) propose a formation control with artificial fixed delays for multiple double-integrators and verify the asymptotic stability of the scheme by checking the solvability of an LMI. Semsar-Kazerooni and Khorasani (2009) present an LMI formulation of the LQR problem to guarantee stable state consensus with an optimal control effort in a multi-agent system. On the other end, to avoid collision avoidance, the chosen solution is Artificial Potential Field (APF). In analogy with a particle moving in an electrostatic or gravitational field, virtual repulsive force fields are simulated to ward off the robot from an obstacle. This approach has been widely used in formation control of ground vehicles, see for example the work of Yongshen et al. (2018). However, in more recent works, e.g. Han et al. (2024) and Piet et al. (2025), APFs are employed to navigate around obstacles and prevent collisions while attaining the desired shape.

This paper represents the continuation of the previous work by Morando et al. (2025), where a control architecture combining Feedback Linerarization (FL), a robust LMI-based controller and APFs was proposed and validated through simulations for the formation of three

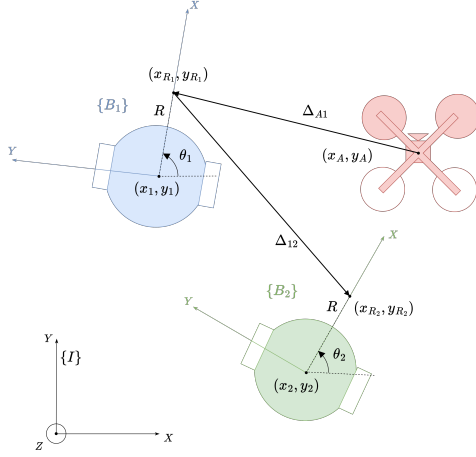


Fig. 1. Air-ground heterogeneous fleet.

unicycles. The successive developments sought to manage the challenge of a heterogeneous team (replacing a ground vehicle with a quadcopter) by adapting the methodology and then experimentally validating the new solution. The contributions of this work are: a) a controller involving FL, a LMI-based controller and APFs is presented for the stable formation of an air-ground fleet; b) the proposed controller has been validated through MATLAB Simulink simulations and a comprehensive set of experiments, including both static and dynamic obstacle scenarios. The remainder paper is structured as follows. In Section 2, the mathematical model used to describe the team is stated. The proposed formation controller is detailed in Section 3. The simulations and experiments results are reported and analyzed in Section 4 and in Section 5 respectively. Finally, conclusions are drawn in Section 6.

2. HETEROGENEOUS FLEET MODELING

3. FORMATION CONTROLLER

4. SIMULATIONS RESULTS

5. EXPERIMENTAL VALIDATION

6. CONCLUSION

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