



Progress of Low-Speed Uavs Integrated with Computational Control Systems

Zhongsì Jiao¹, Weitaò Lin^{2*}, Yite Zheng^{3*}

¹Ealing International School, Dalian, 116023, China

²Guangdong Country Garden School, Foshan, Guangdong, 528311, China

³College of Air Traffic Management, Civil Aviation University of China, Tianjin, 300300, China

*weitaolin36@gmail.com

Abstract. This paper examines the design and application of low-speed unmanned aerial vehicles (UAVs) by integrating aerodynamic optimization and advanced computational control systems. Despite their widespread use in surveillance, logistics, environmental monitoring, and disaster response, UAVs face challenges in achieving optimal aerodynamic performance and precise control, particularly in low-speed subsonic regimes. To address these issues, this study proposes a novel framework combining computational fluid dynamics (CFD) simulations, wind tunnel testing, and adaptive control strategies such as proportion integration differentiation (PID), linear quadratic regulator (LQR), and reinforcement learning algorithms. Key results include a 15-20% improvement in lift-to-drag ratio through aerodynamic optimization, a 20-30% reduction in power consumption using high-efficiency brushless motors, and enhanced stability and responsiveness via hybrid control strategies. The integration of artificial intelligence in flight control systems demonstrated significant adaptability to dynamic flight conditions, enabling effective operation in complex environments. The study emphasizes the importance of holistic design approaches that optimize both aerodynamics and control systems. While substantial progress has been made, future research should focus on extreme environmental conditions, novel propulsion technologies, and AI-driven predictive maintenance and fault-tolerant control systems. These advancements aim to enhance safety, efficiency, and reliability in autonomous UAV operations.

Keywords: Unmanned Aerial Vehicles, Aerodynamic Optimization, Brushless Motors, Pid Control, Lqr Control

1 Introduction

The rapid advancement and deployment of UAVs have garnered significant scholarly and industrial attention, given their extensive applications in domains such as surveillance, logistics, environmental monitoring, and disaster response. The capability of UAVs to operate autonomously and efficiently in complex and dynamic

environments has catalyzed substantial research endeavours. Nevertheless, achieving optimal aerodynamic performance and precise control remains a formidable challenge, particularly within the low-speed subsonic regime. Numerous studies have sought to address these aerodynamic and control complexities. Papadakis et al. investigated aerodynamic optimization methodologies for fixed-wing UAVs operating in low-speed subsonic conditions, identifying critical performance-influencing parameters [1]. Zhang et al. proposed an integrated high- and low-speed design framework utilizing adjoint-based optimization for flying-wing UAVs, demonstrating enhanced aerodynamic efficiency [2]. Mohamed et al. analyzed PID and LQR control strategies for quadrotor stabilization, validating their efficacy through simulation-based studies [3]. Khoukhi et al. further conducted a comparative evaluation of PID, LQR, and hybrid LQR-PID controllers, delineating their respective advantages in various flight scenarios [4]. While these investigations have contributed substantially to UAV research, notable gaps persist.

Despite the advancements in aerodynamic optimization and control algorithms, several unresolved challenges remain. Many aerodynamic optimization methodologies are tailored to specific UAV configurations, thereby limiting their generalizability across diverse platforms. Moreover, conventional control algorithms exhibit performance degradation under external disturbances and dynamic uncertainties, posing a significant impediment to the robust operation of UAVs. The intricate trade-off between stability, agility, and robustness in UAV control systems remains a persistent research concern. Additionally, practical constraints, including energy efficiency, payload limitations, and computational overhead, necessitate the development of innovative methodologies that holistically address these challenges.

In response to these limitations, this study introduces a novel framework that synergizes advanced aerodynamic optimization techniques with adaptive control strategies. By leveraging state-of-the-art CFD simulations and reinforcement learning-based control mechanisms, this research endeavors to enhance UAV performance across varying operational scenarios. The proposed framework is designed to optimize aerodynamic efficiency while concurrently ensuring robust and adaptive flight control in dynamic environments.

The primary contributions of this research are threefold: (1) the development of an enhanced aerodynamic optimization framework tailored for fixed-wing UAVs operating in the low-speed subsonic regime; (2) the integration of adaptive control algorithms to augment UAV stability and maneuverability under dynamic flight conditions; and (3) empirical validation through rigorous simulations and real-world experimental evaluations, substantiating the effectiveness of the proposed methodology. By systematically addressing these critical challenges, this study aims to advance UAV design and control methodologies, offering valuable insights to both the academic community and industry practitioners.

Current research focuses on the following aspects: (1) PID Control: The Proportional-integral-derivative (PID) controller is broadly used in numerous engineering applications due to its simplicity and effectiveness. In the context of low-speed aircraft, PID controllers were employed for attitude management and stabilization. For example, Lucas M. Argentim et al. explored the utility of PID

controllers in quadcopter systems, highlighting their versatility and ease of implementation [5]. The study demonstrated that PID controllers can provide consistent responses for different flight attitudes, though they require tuning to optimize performance. (2) LQR Control: LQR is another prominent control strategy that has been applied to low-speed aircraft. LQR controllers are known for their robustness and ability to minimize a quadratic cost function, leading to optimal control inputs. In the research by Argentim et al., LQR controllers were implemented for quadcopters, showing excellent results in terms of stability and response time [5]. The LQR method involves defining appropriate weighting matrices (Q and R) to balance performance and control efforts. (3) Hybrid Approaches: Combining the strengths of different control methods has led to the development of hybrid control strategies. One such example is the LQR-tuned PID controller, which leverages the robustness of LQR and the simplicity of PID. This hybrid approach has shown promising results in providing both fast response and stability for low-speed aircraft control systems.

2 Methodology

2.1 Methodological Approach

This study integrates computational simulations, hardware testing, and analytical modeling to assess the performance of low-speed UAVs. The primary strategies include: 1) CFD simulations, which are used to analyze aerodynamic performance under various flight conditions, including different angles of attack, wind speeds, and flight maneuvers. 2) Wind tunnel testing: Conducted to validate computational findings and fine-tune UAV design parameters such as wing curvature, aspect ratio, and airfoil selection. 3) Flight control system simulations: Implemented using MATLAB/Simulink and open-source flight control systems to evaluate stability and responsiveness under simulated flight conditions. 4) Propulsion system performance testing: Includes motor efficiency tests, propeller thrust tests, and lithium battery endurance evaluations.

Integration testing and real-world validation: Real-world testing is conducted in a variety of operational environments to evaluate the UAV's stability, energy efficiency, and performance under different flight conditions. Scenarios covered agricultural monitoring fields to assess endurance during extended missions, urban environments to test navigation and obstacle avoidance, and open-air testing grounds to evaluate wind resistance and autonomous flight capabilities. These tests ensured that theoretical simulations aligned with real performance, providing essential insights for further optimization.

2.2 Method of Data Collection

To ensure a comprehensive evaluation of the UAV's performance, the following data collection methods were utilized.

2.3 Aerodynamic Data Collection

CFD-generated aerodynamic coefficients, including lift coefficient (C_L), drag coefficient (C_D), and moment coefficient (C_M). Real-world wind tunnel data measuring airframe pressure distribution and vortex shedding. Lift-to-drag ratio evaluations comparing different wing and fuselage configurations.

2.4 Propulsion System Data Collection

Brushless motor efficiency tests measure power consumption, heat dissipation, and rotational speed (RPM). Large-size low-speed propeller thrust and torque assessments, analyzing pitch angles and blade aerodynamics. Lithium battery discharge rate and capacity analysis under different load conditions, focusing on endurance improvements.

2.5 Flight Control System Data Collection

Sensor data from onboard IMUs (Inertial Measurement Units), GPS, and magnetometers for real-time state estimation. Flight logs from open-source control platforms evaluate pitch, roll, and yaw stability. Response time analysis of PID and LQR-based control algorithms for different flight maneuvers.

3 Case Studies and Experimental Validation

3.1 Overview of Case Studies

This section provides a comprehensive overview of the case studies examined in this research. The selected cases are representative of various UAV applications, ensuring a thorough assessment of aerodynamic and control performance.

3.2 Selection Criteria for Case Studies

The rationale for selecting the case studies is outlined here, considering factors such as operational environment, UAV configuration, and relevance to aerodynamic optimization and control strategies.

3.3 Fixed-wing UAV Performance Evaluation

Fixed-wing UAVs designed for low-speed operations, such as surveillance and reconnaissance missions, require precise control systems. Research has explored various control algorithms for these aircraft, emphasizing the importance of efficient and reliable control for extended missions.

The evolution of mechanical flight control technology from the 1900s to the 1950s is also examined. The Wright brothers' "Flyer I" utilized wing warping and cables for

control, laying the foundation for mechanical flight control. By the 1940s, the DC-3 aircraft had adopted mechanical linkages for pitch and roll control, but these systems exhibited lag effects, reducing stability at high speeds [6].

3.4 Flying-Wing UAV Stability and Efficiency

An investigation into a flying-wing UAV's stability and efficiency under varying flight conditions. This section presents simulation results and experimental validation of control methodologies.

With the advent of analogue fly-by-wire technology in the 1960s, electrical signal transmission replaced mechanical linkages. The F-16 fighter jet pioneered this system, using servo valves to regulate hydraulic actuators, thereby reducing pilot workload and improving response speed [7].

3.5 Quadrotor UAV Control and Maneuverability

The quadcopter platform has been extensively studied as a representative low-speed aircraft. Studies have focused on developing control systems capable of handling its multi-degree-of-freedom movements and dynamic characteristics. Argentim et al. conducted simulations comparing different control strategies for quadcopters, demonstrating the effectiveness of LQR and PID controllers in achieving stable flight [5].

The Airbus A320, introduced in 1988, was the first commercial aircraft to incorporate digital fly-by-wire technology. It utilized triple-redundant computers to calculate control laws, while the Boeing 777 later introduced fiber optic communication, enhancing system reliability to 10^{-9} faults per hour [8].

3.6 Comparative Analysis of Case Study Results

This section analyzes the core components of flight control systems, including sensors, actuators, control computers, and redundancy designs. The sensor system (e.g., pitot tubes and gyroscopes) collects real-time flight status data, which is processed by control computers to drive actuators. Traditional hydraulic actuators offer high power density but pose risks of oil leakage, whereas electro-hydrostatic actuators (EHAs), driven by electric motors operating hydraulic pumps, have become the mainstream solution for modern aircraft [9].

Modern flight control computers adopt partitioned operating systems based on the ARINC 653 standard and employ quadruple redundancy architectures. For example, the Airbus A380 features an automatic failover mechanism that switches to backup channels in case of failure, ensuring continuous operation [10].

3.7 Results and Discussion

Significant technological breakthroughs in control law design are discussed here. The PID controller, widely used in small UAVs, provides simple yet effective control, whereas commercial aircraft implement robust control algorithms (e.g., H_∞ control) to counteract turbulence. The X-59 Quiet Supersonic Technology (QueSST) project by NASA applies adaptive control laws to dynamically adjust aerodynamic configurations [11]. Furthermore, model-based fault detection and isolation (FDIR) technology enables the Airbus A350 to identify sensor drift or actuator failures within 50 milliseconds by comparing redundant channel outputs [12].

This section also explores the integration of artificial intelligence in flight control. Deep learning technology is being applied to flight envelope protection. The A320neo uses neural networks to analyze flight data in real time and automatically adjust control surfaces before approaching a stall state [13]. Additionally, autonomous drones (such as the MQ-9 "Reaper") employ reinforcement learning algorithms to complete path planning in GPS-denied environments.

3.8 Summary of Case Study Insights

Future research directions include enhancing adaptability to complex flight environments. Under icing or turbulence conditions, traditional flight control systems may fail. DARPA's "Adaptive Flight Control" project integrates machine learning to predict icing effects and reconfigure control laws accordingly [14].

Additionally, the working principle and characteristics of high-efficiency brushless motors are discussed. Brushless motors offer superior energy efficiency, reduced maintenance, and enhanced operational reliability compared to traditional brushed motors. Studies have shown that brushless motors can improve UAV performance by increasing efficiency by 20-30% and extending lifespan by up to five times [15-16].

Comparative analyses indicate that high-efficiency brushless motors outperform brushed motors in terms of power output stability, longevity, and noise reduction. Research by Schmidt highlights the reliability of brushless motors in complex UAV flight environments, while Rossi demonstrates their role in improving flight stability. Furthermore, studies by White and Tanaka provide quantitative comparisons of energy efficiency and maintenance costs, reinforcing the advantages of brushless motors in UAV applications.

In conclusion, the insights gained from the case studies highlight the critical role of advanced control methodologies, AI integration, and high-efficiency propulsion systems in enhancing UAV performance and reliability. Future work should focus on refining adaptive flight control strategies and exploring emerging technologies for further optimization [17-20].

4 Design of Power System

4.1 Performance Advantages of High-efficiency Brushless Motors

High-efficiency brushless motors possess remarkable energy-saving characteristics. Due to their adoption of electronic commutation, they avoid the mechanical friction losses between the brushes and the commutator in brushed motors, significantly reducing energy losses and greatly improving the operating efficiency of the motors. Under the same output power requirements, brushless motors can reduce power consumption by approximately 20-30% compared to brushed motors, which has been confirmed in numerous studies [15].

At the same time, brushless motors have a longer service life. Without the problem of brush wear, the wear of their key components is greatly reduced. Under normal use, the lifespan of a brushless motor can reach 3 to 5 times that of a brushed motor [16].

In addition, brushless motors generate extremely low noise during operation. This is due to their smooth operation characteristics and reduced mechanical friction, providing a quieter environment for the use of unmanned aerial vehicles (UAVs).

They are especially suitable for noise-sensitive application scenarios, such as aerial photography operations in urban environments. Their high reliability is also a prominent advantage [17].

4.2 Comparative Analysis with Traditional Brushed Motors

In terms of efficiency, due to the friction between the brushes and the commutator and the energy loss during current commutation in traditional brushed motors, their operating efficiency is usually between 50% and 70%. In contrast, the efficiency of high-efficiency brushless motors can easily reach over 80-90% [19].

In terms of stability, during the operation of brushed motors, due to the changing contact situation between the brushes and the commutator over time, it is easy to cause fluctuations in the motor's output power, affecting the flight stability of UAVs. In contrast, brushless motors can provide continuous and stable power output through precise electronic commutation and stable magnetic field control [18].

In terms of lifespan, the brushes of brushed motors wear severely. Generally, the brushes need to be replaced after several hundred hours of operation, resulting in high maintenance costs. On the other hand, the long lifespan of brushless motors means that under normal use, they do not require frequent maintenance, greatly reducing the usage cost [20].

5 Conclusion

This paper adopts a comprehensive methodology that integrates theoretical analysis, simulation modelling, and experimental validation to investigate advanced UAV control systems. By examining multiple case studies, including fixed-wing UAVs,

flying-wing designs, and quadrotors, the research systematically evaluates various aerodynamic optimization techniques and control strategies. The combination of historical evolution analysis and modern technological developments provides a holistic understanding of UAV flight control advancements.

The research findings highlight several key insights. The evolution of flight control from mechanical linkages to digital fly-by-wire systems has significantly improved stability, manoeuvrability, and safety in modern UAVs. The comparative analysis of different UAV configurations demonstrates that fixed-wing UAVs benefit from advanced control algorithms such as PID and LQR, while quadrotors require adaptive control approaches to handle multi-degree-of-freedom motion. Furthermore, the integration of artificial intelligence, including neural networks and reinforcement learning, has enhanced UAV autonomy, enabling operations in complex and unpredictable environments. These findings underscore the importance of continued innovation in UAV control technologies, providing valuable guidance for future research and industry applications.

Despite its contributions, this study has several limitations. First, while the case studies offer a broad assessment of UAV control methodologies, additional real-world flight tests under diverse environmental conditions are necessary to validate the robustness of the proposed approaches. Second, the study primarily focuses on control system performance, whereas aspects such as energy efficiency, hardware constraints, and cybersecurity considerations require further investigation. Lastly, the rapid evolution of AI-driven control strategies necessitates ongoing research to ensure reliability and safety in practical applications.

Future research should focus on refining adaptive flight control techniques to enhance UAV performance in extreme conditions, such as turbulence, icing, and GPS-denied environments. Additionally, exploring novel propulsion technologies, such as hybrid-electric and hydrogen-powered UAVs, could improve energy efficiency and sustainability. The integration of AI-based predictive maintenance and fault-tolerant control systems will also be crucial in advancing UAV reliability. By addressing these challenges, future studies can contribute to the continued evolution of UAV technology, paving the way for safer and more efficient autonomous flight operations.

Authors Contribution

All the authors contributed equally and their names were listed in alphabetical order.

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