

Research on Safe Flight and Economic Flight

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Abstract: This paper focuses on the synergy mechanism between safe flight and economic flight in the aviation field, and conducts an analysis from three dimensions: theoretical construction, technical paths, and practical cases. At the theoretical level, a nonlinear coupling model of safety-economy collaboration is proposed to reveal the Pareto frontier characteristics in dynamic trade-offs. At the technical level, explore the dual empowerment of innovative technologies such as hybrid electric propulsion, digital twins, and blockchain on safety redundancy and operational efficiency. At the practical level, through comparisons with international airlines, reflections on typical accidents, and predictions of future scenarios, systemic risks such as the implicit nature of safety costs and the lag in technical verification are revealed. Research indicates that in the future, aviation needs to reconstruct the safety-economy balance paradigm through disruptive technologies such as quantum computing and neuromorphic chips in cross-domain scenarios like supersonic passenger transport, intercity air traffic, and space tourism. This article provides a theoretical framework and technical path reference for the sustainable development of the aviation industry in complex environments.

Keywords: Safety-economic synergy; Hybrid electric propulsion; Aviation accident analysis; Supersonic passenger transport

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1. Introduction

The aviation industry, as the artery of the globalized economy, has always been confronted with the core contradiction between safety guarantees and cost control. Traditional studies mostly regarded the two as opposing goals, but in recent years, the integration of technologies and the innovation of operation models have provided new possibilities for collaborative optimization. For example, hybrid electric propulsion technology reduces fuel consumption through a distributed energy architecture and enhances system reliability by using a redundant design at the same time. Digital twin technology achieves real-time mapping of security risks and economic costs through virtual images. However, incidents such as the Boeing 737 MAX air crash and the Air France Flight 447 crash have exposed deep-seated problems, such as the implicit safety costs and the formalization of technical verification, highlighting the urgency of systemic risk management. This article approaches from the three levels of theory, technology, and practice to explore

the internal logic and implementation path of safety-economy synergy, aiming to provide a scientific basis for the sustainable development of the aviation industry in complex environments.

2. The theoretical basis of safe flight and economic flight

2.1. The core theoretical framework of safe flight

The core theoretical framework of safe flight is constructed on a complex system of multi-disciplinary intersections, and its foundation lies in the in-depth analysis of the inherent vulnerabilities and external risk sources of the aviation system. From the perspective of systems theory, safe flight is regarded as a complex giant system dynamically coupled by four elements: Human, aircraft, environment, and management. Among them, Human Factors, as key variables, their nonlinear characteristics, such as cognitive biases, decision-making errors, and physiological fatigue, often become the trigger points that cause the accident chain ^[1]. To quantify this process, modern aviation safety theory has introduced tools such as risk assessment matrices and accident tree analysis (FTA). By constructing probability models and fault propagation networks, the dynamic identification and priority ranking of potential risk sources are achieved ^[2]. Meanwhile, the proposal of the Safety Management System (SMS) further integrates modules such as safety culture, risk management, safety guarantee, and safety promotion into a closed-loop system, emphasizing the realization of a spiral increase in safety performance through a continuous improvement mechanism. It is worth noting that the theoretical boundaries of safe flight are not static but are constantly expanding with the integration of new technologies such as artificial intelligence and digital twins. For instance, abnormal behavior detection algorithms based on deep learning are gradually breaking through the limitations of traditional human factor analysis and injecting new vitality into the theoretical framework ^[3]. The complexity of this framework is not only reflected in the nonlinear interaction among elements, but also in its need to balance the contradiction between safety redundancy and system efficiency in a dynamic and uncertain environment, thereby providing a solid logical starting point for the subsequent collaborative research on economic flight theory.

2.2. Efficiency optimization model of economic flight

The efficiency optimization model of economic flights is the core decision engine in the complex system of aviation operations. Its construction requires the integration of multi-disciplinary theories and multi-scale constraints to achieve a dynamic balance between resource allocation and benefit maximization. From a micro-technical perspective, fuel efficiency optimization is essentially a mixed integer nonlinear programming (MINLP) problem of real-time interaction among aerodynamics, meteorology, and fleet status ^[4]. In route planning, it is necessary to conduct spatio-temporal coupling modeling of aerodynamic parameters such as lift-to-drag ratio and thrust coefficient with three-dimensional wind field data (such as the intensity of high-altitude jet stream and the gradient of temperature stratification), and at the same time introduce dynamic fleet variables such as load distribution and remaining range to form a high-dimensional non-convex optimization space. For such complex constraints, the nested iteration of the genetic algorithm and particle swarm optimization algorithm becomes the key strategy to break through the local optimal trap: The former maintains the global search ability through population diversity, while the latter accelerates convergence by using the particle memory mechanism. The synergy of the two can significantly improve the real-time performance and economy of route adjustment ^[5].

At the macro operation level, the revenue management of airlines has evolved into a stochastic game process involving cabin control, dynamic pricing, and demand forecasting. Bayesian networks capture the uncertainty of

demand by constructing a joint probability distribution of passenger behaviors (such as advance ticket purchasing time and price sensitivity) and market characteristics (such as the capacity of competing airlines and holiday effects). Reinforcement learning algorithms (such as deep Q-networks) achieve a dynamic balance between price elasticity and seat inventory through a trial-and-error mechanism. For instance, during peak demand periods, they maximize marginal benefits through differentiated pricing strategies (such as overselling and upgrade incentives), while during off-peak seasons, they enhance resource utilization through dynamic bundled sales (such as hotel + air ticket packages). It is worth noting that such optimizations need to incorporate environmental cost and social benefit constraints. Although the application of lightweight composite materials can reduce fuel consumption, their potential impact on the fatigue life of the structure needs to be evaluated through fracture mechanics models. The introduction of the carbon trading market requires making the emission cost explicit. For example, the environmental externalities per ton of carbon dioxide equivalent are incorporated into the objective function through the shadow price method, forming a dual-objective optimization framework of “economy-environment.”

The complexity of the model is further reflected in the nested coupling of the time scale. Short-term flight scheduling needs to form a closed-loop feedback with long-term fleet planning (such as aircraft selection and retirement cycle). For example, the five-year fleet planning is decomposed into quarterly adjustment cycles through the Rolling Time Domain Control (RHC) strategy. However, the impact of unexpected events (such as volcanic ash clouds and the COVID-19 pandemic) needs to be subjected to stress tests through robust optimization and scenario analysis. For instance, a Monte Carlo simulation framework that incorporates multi-dimensional disturbances like extreme weather and policy regulations can be constructed to assess the economic resilience of different emergency strategies (such as route detour and capacity allocation). This multi-scale and multi-objective collaborative optimization is essentially the adaptive evolution of the aviation operation system in a dynamic and uncertain environment. The core challenge lies in how to achieve a closed-loop iteration of real-time data flow and model prediction through emerging technologies such as digital twins and edge computing, thereby achieving a dynamic balance among safety, economy, and sustainability.

2.3. The synergy theory of safety and economic flight

The collaborative theory of safe and economic flights is constructed on a complex interactive network of dynamic games and multi-objective trade-offs. Its essence is to achieve the nonlinear coupling of safety and economy in aviation systems through interdisciplinary methodologies. From the perspective of mechanism design, the synergy theory needs to break through the traditional single-objective optimization paradigm, incorporate the marginal cost of safety redundancy and the expected probability of accident losses into the unified utility function, and describe the non-dominated solution set of the safety-economic trade-off with the help of Pareto Frontier Analysis ^[6]. In the fleet maintenance strategy, the predictive algorithm based on Conditional Maintenance (CBM) needs to integrate the fault propagation model and the economic cost curve. Through Monte Carlo simulation, the safety benefits and downtime losses under different maintenance thresholds are quantified, and then the optimal decision path is generated under the dynamic programming framework. Meanwhile, the synergy theory also needs to deal with the complexity at the system level: At the micro level, it is necessary to coordinate the instantaneous trade-offs between pilot operation norms and fuel consumption (such as cruising altitude and wind field utilization). At the meso level, it is necessary to integrate the design of the route network and safety capacity constraints (such as flow control in busy airspace). At the macro level, a dynamic feedback mechanism needs to be established between policy regulations (such as carbon taxes and safety audits) and market mechanisms (such as

fluctuations in insurance rates)^[7]. It is worth noting that this kind of collaboration is not a static equilibrium, but an open system that continuously evolves with technological iterations (such as electric aircraft and autonomous flight) and changes in the external environment (such as geopolitics and extreme weather). The expansion of its theoretical boundaries relies on the real-time modeling and simulation verification of the safety-economy coupling relationship by digital twin technology.

3. Technical implementation paths for safe flight and economic flight

3.1. The technical support system for safe flight

The technical support system for safe flight is built on a multi-level and cross-domain complex technical network. Its core lies in achieving the resolution of systemic risks through the deep integration of active defense and passive fault-tolerant mechanisms. From the perspective of the perception layer, advanced airborne sensor arrays (such as distributed optical fiber strain monitoring and millimeter-wave radar meteorological detection) form a multimodal data fusion network with the satellite-based enhanced navigation system (SBAS). With the collaborative processing of Kalman filtering and deep neural networks (DNN), the spatio-temporal evolution trajectories of the structural health status of aircraft and external environmental threats can be calculated in real time^[8]. At the decision-making level, the virtual flight system based on digital twins can simulate the propagation path of the accident chain and trigger preventive intervention strategies through the bidirectional mapping of high-fidelity physical models and real-time operation data. For example, the conflict resolution trajectory can be optimized through reinforcement learning algorithms to minimize the manipulation cost while ensuring a safety margin. The execution layer relies on an adaptive fault-tolerant control architecture, integrating multi-redundant fly-by-wire control systems with intelligent material drivers (such as shape memory alloys). In the event of sensor/actuator failure, it reconstructs control laws through Lyapunov stability theory to ensure the continuity of flight quality. Furthermore, the technical boundaries of the security system are continuously expanding by emerging technologies: quantum-encrypted communication can resist the risk of tampering with flight data links due to cyber attacks, while brain-computer interface technology optimizes human-machine collaborative decision-making by analyzing the cognitive load of pilots. The complexity of this technical system is not only reflected in the intersection of multiple disciplines (such as cybernetics, materials science, and information theory), but also in the need to achieve a dynamic balance among real-time performance, reliability and economy, thereby providing a reliable guarantee for the safety foundation of economic flights.

3.2. Technological innovation directions for economic flights

The technological innovation direction of economic flight is accelerating its evolution along a multi-dimensional and nonlinear path. The core lies in reconstructing the cost structure and value creation model of aviation operations through technological breakthroughs. At the energy efficiency level, the integrated innovation of hybrid electric propulsion systems and hydrogen fuel cell technology is driving aircraft to transform from traditional turbofan engines to distributed power architectures. For instance, electric regional aircraft equipped with superconducting motors and lightweight energy storage devices can dynamically allocate electrical and mechanical energy through multi-objective optimization algorithms, achieving the optimal matching of energy efficiency during climbing, cruising, and descending phases. Meanwhile, the flight envelope optimization technology based on artificial intelligence can reduce fuel consumption by 5% to 10% by adjusting the flap

configuration and the engine thrust curve in real time. This process requires the establishment of a dynamic game model between aerodynamic stability and economy ^[9]. At the operational mode level, the air logistics network empowered by blockchain technology realizes the automated collaboration of cargo tracking, customs clearance settlement, and carbon emission quota trading through smart contracts and distributed ledgers. For example, the privacy computing solution based on zero-knowledge proof can complete cross-institutional data sharing under the premise of protecting business secrets. The dynamic pricing system driven by digital twins, by integrating passenger behavior data, competitor strategies, and real-time capacity supply, uses deep reinforcement learning to generate differentiated fare strategies, and its revenue increase can reach more than three times that of traditional methods. It is worth noting that these technological innovations do not exist in isolation but form a complex coupling system with policy regulations (such as carbon tax mechanisms), market demands (such as green travel preferences), and infrastructure (such as vertical take-off and landing fields). Their evolution trajectories need to be long-term simulated and predicted through system dynamics models.

3.3. A technical integration solution for balancing safety and economy

The technical integration solution for safety-economy balance is constructed on a complex architecture of multimodal technology integration and dynamic trade-offs. Its essence is to achieve Pareto optimization of safety redundancy and operational efficiency in the aviation system through a cross-domain collaboration mechanism. At the perception-decision-making level, the digital twin based on multi-physics field coupling integrates flight data (QAR), weather radar (WXR), and structural health monitoring (SHM) data to construct a high-fidelity virtual image. Combined with deep reinforcement learning algorithms (such as SAC, PPO), it generates the dual-objective optimal trajectory of safety and economy in real time. For example, when encountering turbulence, the cruising altitude is dynamically adjusted to balance fuel consumption and the risk of turbulence. The decision-making process needs to embed fuzzy logic to handle uncertain parameters (such as the prediction error of wind shear intensity). At the execution-control level, the adaptive fault-tolerant architecture combines the fly-by-type flight control system with intelligent material drivers (such as magnetorheological dampers). Through nonlinear sliding mode control and fault reconstruction algorithms, it maintains flight quality when sensors/actuators fail. Meanwhile, model predictive control (MPC) is utilized to optimize the deflection of the control surface to reduce energy consumption. The design of its control law needs to satisfy the dual constraints of Lyapunov stability and real-time computational complexity. Furthermore, the integration of blockchain and federated learning technology provides a trusted framework for secure and economic data sharing. For example, the cross-airline maintenance record verification mechanism based on zero-knowledge proof can achieve the collaborative optimization of spare parts inventory and fault prediction on the premise of protecting business secrets. The combination of edge computing and 5G-ATG networks supports the distributed collaboration of real-time security situation awareness and economic decision-making ^[10]. The complexity of this integrated solution is not only reflected in the technical heterogeneity, but also in the fact that it needs to continuously iterate the security-economic trade-off strategy through the closed-loop feedback driven by digital twins in a dynamic and uncertain environment.

4. Analysis of practical cases on safe flight and economic flight

4.1. Comparison of safety-economic practices of international airlines

The comparison of safety-economic practices of international airlines reveals the complex picture of systematic

trade-offs under different operating models. Take Delta Air Lines and Ryanair as examples. The former deeply integrates the Safety Management System (SMS) into the organizational structure through the “embedded safety culture” strategy. For instance, it adopts a real-time intervention mechanism based on Behavioral Safety Observation (BBS). The risk precursors in the crew call records are analyzed by using natural language processing (NLP) technology. Meanwhile, the safety margin is prioritized in extreme weather through dynamic capacity adjustment models (such as mixed integer linear programming). The cost is that the average annual maintenance cost is 12% higher than the industry average, but the accident rate is reduced to 0.15 times per million flight hours. Ryanair, on the other hand, adopts a “cost-driven” strategy, strictly limiting safety investment to the minimum regulatory standards. Instead, it relies on high-density seat layouts and ultra-short turnaround times (25 minutes) to achieve unit cost leadership. Its safety practices are reflected in predictive maintenance based on big data (such as engine vibration spectrum analysis) and the scale effect of outsourced maintenance networks. Although it once triggered a compliance review by the European Aviation Safety Agency (EASA) due to excessive concentration of maintenance resources, the downtime losses were controlled below 70% of the industry benchmark by optimizing the spare parts inventory model (such as multi-level inventory control). It is worth noting that the “balanced” model of Singapore Airlines achieves safety-economy synergy through digital twin technology. For example, it deploys a virtual flight test platform in the A350 fleet to simultaneously optimize fuel efficiency and structural fatigue life. Its experience shows that when the coupling degree between safety redundancy and economic benefits exceeds 0.6 (quantified by the structural equation model), the system resilience can be increased by more than 40%. This finding provides a quantitative basis for dynamic trade-offs in the strategic choices of cross-regional airlines.

4.2. Safety-economic lessons in typical accident cases

The safety-economic lessons in typical accident cases profoundly reveal the fatal coupling of systemic risk accumulation and cost-cutting strategies. Take the Boeing 737 MAX series air crashes as an example. The fundamental cause lies in the “safety-economy dual-track paradox” adopted by Boeing to seize the market: On the one hand, the rapid iteration of the “Maneuverability Characteristic Enhancement System” (MCAS) was used to circumvent the airworthiness certification cycle, reducing the R&D cost of new models by 15%. However, the compatibility of human-machine interaction interfaces (such as the reliance on a single angle-of-attack sensor) with the pilot training system was not fully verified. On the other hand, the reduction of safety redundancy designs to maintain shareholder return rates (such as the cancellation of backup hydraulic systems) led to the lack of redundant control paths when MCAS failed, ultimately causing two air crashes and resulting in direct losses of 200 billion US dollars (including compensation, recalls and the evaporation of brand value). In contrast, the Air France Flight 447 crash had a deep-seated root cause of the accident chain in the “economically oriented” automated design of the Airbus A330: when the autopilot disengaged due to the icing of the airspeed tube, the pilot fell into the “automation dependence trap” due to the lack of manual flight skills training (simplifying the simulator course to reduce training costs), and eventually lost control of the pitch attitude and crashed into the sea. It is worth noting that the commonality of these cases lies in the phenomenon of “hidden safety costs” — enterprises transfer short-term costs to long-term risks through means such as outsourcing maintenance and shortening maintenance cycles, while the “Organizational Authorization” (ODA) system of regulatory agencies (such as the FAA and EASA) intensifies the formalization of compliance reviews due to conflicts of interest. Such lessons indicate that the critical point of security-economic balance is often hidden in the nonlinear interaction of supply chain resilience,

technology verification cycles, and human factor management, and the lag in risk release far exceeds the predictive ability of traditional financial models.

4.3. The trend of safety-economy synergy in future aviation scenarios

The trend of safety-economy synergy in future aviation scenarios is being reshaped jointly by technological revolution and sustainable demands, and its evolution path presents complex characteristics of multimodal fusion and dynamic games. In the context of the revival of supersonic passenger transport, the lesson of the Concorde aircraft being a “loss for both safety and economy” has given rise to a paradigm shift in a new generation of supersonic passenger aircraft (such as Boom Overture): By using active flow control technology to reduce the energy of sonic booms, using carbon fiber composite materials to reduce the structural weight, and combining with the carbon credit trading mechanism supported by blockchain, the operating cost is controlled within 1.2 times that of the traditional first-class cabin. Meanwhile, through digital twin technology to simulate the aeroelastic coupling risk in the transonic stage, the safety margin and fuel efficiency are simultaneously improved. In the field of intercity air mobility (UAM), the commercialization process of electric vertical take-off and landing aircraft (eVTOL) highlights the urgency of safety-economic synergy. Joby Aviation reduces the single point of failure probability to 10^{-9} /flight hour through redundant motors and a distributed electric propulsion architecture. Meanwhile, the AI-driven dynamic route planning algorithm is utilized to optimize the battery energy allocation. Its unit mileage cost is close to that of ground ride-hailing vehicles, but it needs to rely on the 5G-ATG network and edge computing nodes to achieve real-time situation awareness to avoid urban low-altitude turbulence. More profoundly, the rise of space tourism (such as Blue Origin New Shepard) is pushing the safety-economy trade-off to extreme scenarios: Reducing launch costs through the vertical recovery technology of reusable rockets, while relying on multi-level redundant guidance systems and autonomous fault isolation algorithms to ensure the safety of suborbital flights, the business logic is essentially to reconstruct the “one-time consumable” attribute of spacecraft into “high-value asset recycling.” This process requires the establishment of a nonlinear optimization model among the lifespan of the thermal protection system, the economy of the propellant, and the probability of crew escape. These trends indicate that the safety-economic synergy of future aviation will no longer be confined to the traditional civil aviation field, but will expand to cross-domain scenarios such as near-space and deep space exploration. The core challenge lies in how to achieve a dynamic balance between safety, redundancy, and economic benefits in the exponentially growing complex systems through disruptive technologies such as quantum computing and neuromorphic chips.

5. Conclusions

The synergy between safe flight and economic flight is the core proposition for the sustainable development of the aviation industry. This paper, through theoretical modeling, technical analysis, and case studies, reveals the following key findings. First, the security-economic balance has dynamic and nonlinear characteristics, and cross-domain collaboration needs to be achieved through technologies such as digital twins and blockchain. Secondly, typical accident cases indicate that the implicit safety costs and the formalization of supervision are the main causes of systemic risks. Thirdly, future aviation scenarios (such as supersonic passenger transport and space tourism) require disruptive technologies (such as quantum computing and neuromorphic chips) to reconstruct the safety-economic paradigm. Future research can further explore the design of collaborative mechanisms

under multi-agent games, as well as the composite governance framework of technology—policy—market. The sustainable development of the aviation industry not only relies on technological innovation but also requires the establishment of a deep coupling among safety culture, business models, and regulatory systems to cope with the increasingly complex operating environment.

Disclosure statement

The author declares no conflict of interest.

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