

**ADVANCED NUMERICAL METHODS FOR SOLVING
NONLINEAR PARTIAL DIFFERENTIAL EQUATIONS IN FLUID
MECHANICS: APPLICATIONS IN AEROSPACE ENGINEERING**

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Abstract

The solution of nonlinear partial differential equations (PDEs) in fluid mechanics remains one of the most critical and challenging tasks in computational science, particularly within the domain of aerospace engineering. These equations, primarily derived from the Navier–Stokes framework, govern a wide range of complex phenomena including turbulence, compressible flows, boundary-layer separation, shock–boundary interactions, and hypersonic aerodynamics. Traditional numerical techniques such as finite difference and finite volume methods, while effective in linear or moderately nonlinear regimes, often suffer from limitations when extended to strongly nonlinear, multi-scale problems characterized by high Reynolds numbers and stiff temporal dynamics. In response, recent advances have introduced more robust and adaptive strategies, including spectral methods, high-order finite element formulations, lattice Boltzmann approaches, and machine learning–augmented solvers, all of which have demonstrated significant promise in enhancing accuracy, stability, and computational efficiency. This paper critically examines these advanced numerical methods, with a focus on their comparative performance in aerospace applications such as aerodynamic load prediction, flow stability in re-entry vehicles, shock-capturing in supersonic jets, and turbulence modelling in propulsion systems. Through an integrated perspective, the study highlights both the theoretical underpinnings and practical implementations of these approaches, while emphasizing the importance of high-performance computing and hybrid schemes for tackling real-world aerospace design challenges. The analysis contributes to bridging the gap between mathematical theory and engineering practice, providing a roadmap for future research in nonlinear PDE modelling within aerospace fluid mechanics.

Keywords: Nonlinear PDEs; Fluid Mechanics; Numerical Methods; Aerospace Engineering; Computational Fluid Dynamics (CFD); Turbulence Modelling; High-Performance Computing.

I. INTRODUCTION

The study of nonlinear partial differential equations (PDEs) in fluid mechanics occupies a central place in modern applied mathematics and engineering, as it provides the mathematical foundation for analysing highly complex fluid behaviours encountered in aerospace applications ranging from subsonic flows over air foils to turbulent combustion in jet engines and hypersonic re-entry dynamics. At the core of these problems lie the Navier–Stokes equations and their nonlinear variants, which describe momentum transfer, mass conservation, and energy evolution under conditions of compressibility, turbulence, and multiphase interactions. Unlike linearized formulations, nonlinear PDEs exhibit characteristics such as bifurcations, chaos, shock formation, and multi-scale coupling that render their exact analytical solutions practically unattainable, thereby necessitating the development of advanced numerical strategies. Classical approaches, including finite difference and finite volume discretization's, have historically enabled the first breakthroughs in computational fluid dynamics (CFD) and still form the backbone of many commercial aerospace solvers; however, their reliance on coarse approximations and excessive computational resources for high Reynolds number flows limits their applicability in scenarios demanding predictive precision

and stability. These shortcomings have stimulated the evolution of higher-order and adaptive numerical schemes, such as spectral methods with exponential convergence for smooth flows, finite element formulations tailored for complex geometries, lattice Boltzmann models suitable for mesoscopic flow dynamics, and direct numerical simulations (DNS) or large eddy simulations (LES) for turbulence resolution. Moreover, the advent of high-performance computing (HPC) has amplified the ability to integrate these methods within massively parallel architectures, thereby facilitating the resolution of multidimensional, nonlinear PDE systems previously deemed intractable.

In aerospace engineering specifically, accurate resolution of nonlinear PDEs underpins the prediction of aerodynamic lift and drag, the design of efficient propulsion systems, the modelling of boundary-layer transition on supersonic aircraft, and the mitigation of thermal and structural loads on re-entry vehicles. Yet, despite the rapid progress, significant challenges remain, particularly in achieving a balance between accuracy, stability, and computational feasibility, given that turbulence, shock interactions, and compressibility effects often stretch the limits of current solvers. Furthermore, uncertainties in boundary conditions, geometric perturbations, and multi-physics coupling such as fluid-structure interaction and reactive flows demand numerical frameworks that are not only robust but also flexible to adapt across diverse aerospace scenarios. Emerging research directions have started to incorporate machine learning and reduced-order modelling into traditional PDE solvers, offering the possibility of accelerating convergence, reducing discretization errors, and identifying hidden nonlinear correlations within complex datasets. These hybrid approaches demonstrate a promising pathway for integrating data-driven models with physics-based computation, thereby closing the gap between theoretical fluid mechanics and real-time aerospace engineering demands. Against this backdrop, the present study seeks to critically evaluate advanced numerical methods designed to tackle nonlinear PDEs in fluid mechanics, with particular emphasis on their aerospace engineering applications. By examining the mathematical foundations, computational frameworks, and practical case studies from aerodynamic shape optimization to shock-boundary interaction modelling this work aims to provide a comprehensive assessment of how contemporary numerical techniques can be leveraged to address current challenges and inform future directions in the design, safety, and efficiency of aerospace systems.

II. RELEATED WORKS

The role of nonlinear partial differential equations in fluid mechanics has been extensively studied particularly due to their centrality in representing flow instabilities, turbulence, and compressible effects, all of which directly impact aerospace engineering. The foundational Navier Stokes equations, which are inherently nonlinear, remain unsolved in closed form for most realistic flows, prompting the adoption of numerical approaches as the only viable path toward practical solutions. Early computational fluid dynamics solvers relied on finite difference and finite volume methods, which provided tractable discretization's of nonlinear PDEs. However, their tendency to suffer from numerical diffusion, stability issues at high Reynolds numbers, and inefficiencies in capturing shocks or vortical structures limited their

effectiveness for aerospace applications involving supersonic or hypersonic regimes [1]. Subsequent developments in finite element methods improved flexibility for complex geometries such as turbine blades and re-entry vehicle surfaces, offering adaptive meshing and higher order accuracy, yet they too demanded significant computational resources particularly in three dimensional turbulent simulations [2]. In the last two decades research has pivoted toward spectral and pseudo spectral methods, which provide exponential convergence for smooth solutions. These have been applied successfully in turbulence studies yielding insights into boundary layer transition and vortex breakdown. However, their global basis functions often struggle with discontinuities such as shocks, necessitating hybrid approaches like spectral element methods [3]. For aerospace problems where compressibility and shock interactions dominate, high resolution shock capturing schemes such as Weighted Essentially Non Oscillatory and Discontinuous Galerkin methods have emerged as reliable frameworks. These methods not only maintain stability but also capture sharp gradients essential for modelling supersonic jet flows and shock boundary interactions [4]. The treatment of turbulence remains a critical domain of nonlinear PDE research. Direct Numerical Simulation is often considered the most accurate approach as it resolves all scales of turbulence directly from the Navier Stokes equations, but its prohibitive computational cost makes it infeasible for full scale aerospace vehicles [5]. Large Eddy Simulation has thus gained prominence where large turbulent structures are resolved while smaller scales are modelled using sub grid scale formulations. LES has been applied to aerospace jet noise prediction, vortex wing interactions, and high speed combustion demonstrating its balance of accuracy and feasibility [6]. Yet even LES requires massive computing resources highlighting the importance of high performance computing frameworks and parallelization techniques [7].

Recent advancements in lattice Boltzmann methods have further contributed to the arsenal of nonlinear PDE solvers. While originally developed for mesoscopic fluid modelling, LBM has been successfully adapted for aerodynamic flow problems especially in low Mach number and transitional flow regimes [8]. Its ability to handle complex boundaries and multiphase interactions with relative simplicity has sparked interest in aerospace microfluidic applications. Parallel to LBM, adaptive mesh refinement has gained traction allowing localized mesh refinement in areas of strong gradients such as shock fronts while keeping coarse grids elsewhere to save computational cost [9]. Another important direction has been the incorporation of reduced order models for nonlinear PDEs which approximate high dimensional flow fields using lower dimensional bases such as Proper Orthogonal Decomposition and Dynamic Mode Decomposition. In aerospace design optimization ROMs have been shown to significantly reduce computational time while preserving fidelity in capturing flow dynamics [10]. However, their reliance on pre computed data raises concerns about generalizability to new flow regimes. Recent studies have thus combined ROMs with machine learning particularly deep neural networks to enhance adaptability and predictive capability [11]. The integration of data driven techniques into nonlinear PDE solvers is now seen as a frontier in computational mechanics.

Neural networks have been embedded into PDE frameworks as surrogate models accelerating convergence while maintaining physical consistency through physics informed neural networks. In aerospace engineering PINNs have been applied to shock prediction and turbulence closure modelling with encouraging results bridging the gap between classical numerical solvers and modern AI approaches [12]. Despite their promise such methods face challenges in stability and robustness especially when extrapolating to high Mach number or high Reynolds number flows where nonlinearities dominate. From a computational perspective high performance computing and GPU based solvers have transformed the feasibility of nonlinear PDE research. Multi core parallelization and domain decomposition techniques now allow simulations at scales that were previously impractical such as DNS of transitional flows around full air foils or LES of jet exhaust plumes [13]. Nevertheless the associated computational cost underscores the need for algorithmic innovations in preconditioning, time integration, and error control. Aerospace engineering applications further highlight the practical implications of these numerical advances. For instance shock boundary interaction modelling in supersonic inlets, prediction of heat transfer in hypersonic boundary layers, and aeroelastic coupling in wing flutter dynamics all rely heavily on robust nonlinear PDE solvers. Comparative studies have shown that while classical CFD solvers are sufficient for preliminary aerodynamic analysis, advanced nonlinear numerical methods are indispensable for high fidelity predictions required in safety critical aerospace systems [14]. Moreover, the growing need for sustainable aviation and space exploration underscores the importance of accurate flow modelling to optimize fuel efficiency, reduce emissions, and design reusable spacecraft components. Finally, interdisciplinary research continues to emphasize the necessity of combining mathematical rigor, numerical innovations, and engineering pragmatism. Studies increasingly recommend hybrid approaches that integrate multiple numerical methods such as coupling spectral methods for laminar regions with shock capturing schemes for discontinuities thereby leveraging the strengths of each framework [15]. This convergence of methodologies not only advances the solution of nonlinear PDEs but also positions aerospace engineering as a primary beneficiary of computational mathematics where precision and reliability are indispensable.

III. METHODOLOGY

3.1 Research Design

The study adopts a computational research design focused on the comparative evaluation of advanced numerical methods for solving nonlinear partial differential equations in fluid mechanics, emphasizing applications in aerospace engineering where flow regimes are dominated by turbulence, compressibility, and shock wave interactions. The design integrates mathematical modelling, discretization strategies, computational implementation, and validation against benchmark aerospace problems. By employing a multi method approach, the research aims to establish correlations between accuracy, stability, computational efficiency, and suitability of different solvers under conditions typical of aerospace applications [16].

3.2 Governing Equations

The foundation of the methodology lies in the Navier Stokes equations in their compressible form, which serve as the representative nonlinear PDEs of fluid mechanics. These equations describe conservation of mass, momentum, and energy in three dimensional unsteady flows. For aerospace cases, additional governing relations such as turbulence models, compressibility corrections, and energy equations are included to simulate high Mach number flows. The nonlinear structure of these PDEs demands discretization schemes capable of handling stiffness, chaotic behaviour, and multiscale interactions [17].

3.3 Numerical Schemes

A comparative framework was established involving finite difference, finite volume, finite element, spectral, lattice Boltzmann, and discontinuous Galerkin methods. Each scheme was evaluated on criteria such as discretization accuracy, convergence behaviour, stability under strong nonlinearity, and adaptability to complex aerospace geometries. Special emphasis was placed on high order methods and adaptive solvers, which are essential for shock capturing and turbulence modelling. The advantages and limitations of each scheme were documented with reference to benchmark problems [18].

Table 1: Numerical Methods and Key Characteristics

Method	Accuracy	Stability	Computational Cost	Aerospace Application Example
Finite Difference	Second order	Conditional	Low to moderate	Boundary layer flows
Finite Volume	Second order+	Conservative	Moderate	Shock capturing in jets
Finite Element	High order	Stable	High	Complex geometries (wings)
Spectral Methods	Exponential	Limited near shocks	High	Turbulence transition
Lattice Boltzmann	Mesoscopic	Stable for low Mach	Moderate	Transitional flows
Discontinuous Galerkin	High order	Strong shock handling	Very High	Hypersonic re-entry

3.4 Computational Framework

Simulations were executed on a high performance computing environment using hybrid CPU GPU architectures to leverage parallelization and accelerate convergence. Open source CFD libraries along with custom solvers were integrated to test scalability. Domain decomposition and adaptive mesh refinement were employed to reduce computational overhead while

retaining local accuracy in regions of steep gradients such as shock fronts and boundary layer separation zones [19].

3.5 Validation and Benchmarks

Validation was carried out by simulating well established aerospace benchmarks including flow around a NACA 0012 air foil at transonic conditions, turbulent channel flow at high Reynolds numbers, and shock tube problems for supersonic flow validation. For hypersonic applications, re-entry vehicle nose cone simulations were included to assess solver robustness in extreme conditions. Numerical outputs were compared against experimental wind tunnel data and validated DNS results to ensure reliability [20].

3.6 Data Analysis and Correlation

Numerical results were analysed using statistical indicators such as root mean square error for accuracy, Courant Friedrichs Levy condition compliance for stability, and floating point operation counts for efficiency. Correlation matrices were constructed to compare solver characteristics with flow features such as turbulence intensity, shock strength, and heat transfer. The analysis aimed to highlight trade offs among methods and to determine optimal schemes for specific aerospace applications [21].

3.7 Quality Assurance

Protocols included code verification, grid independence tests, and iterative convergence monitoring. Residual levels were maintained below 10^{-6} for all steady state cases, and transient simulations were run until time averaged statistics converged. Cross validation was performed by applying two independent solvers to identical cases to ensure reproducibility [22].

3.8 Ethical and Environmental Considerations

The research adhered to ethical practices in computational science, ensuring proper acknowledgment of open source software and maintaining transparency of algorithms used. Since the study involves computational rather than physical experimentation, no environmental risks were posed; however, energy consumption of high performance computing resources was monitored as part of sustainable engineering practice [23].

3.9 Limitations

The methodology acknowledges that despite the use of advanced schemes, full scale DNS of aerospace flows at flight Reynolds numbers remains computationally infeasible with current resources. Machine learning augmented models and reduced order techniques are incorporated only as supporting tools and not replacements for physics based solvers. Additionally, boundary condition uncertainties and turbulence model limitations remain potential sources of error.

IV. RESULT AND ANALYSIS

4.1 Convergence and Accuracy of Numerical Methods

The simulations demonstrated notable differences in convergence rate and accuracy across the numerical schemes. High order methods such as spectral and discontinuous Galerkin provided superior accuracy in resolving nonlinear flow instabilities, but required significantly greater computational resources. Finite volume methods exhibited balanced performance, particularly in shock dominated flows relevant to supersonic aerospace applications, while lattice Boltzmann was effective for transitional and low Mach number flows but less reliable in hypersonic regimes.

Table 2: Comparative Accuracy and Convergence of Methods

Method	Convergence Rate	Mean Error (%)	Suitable Flow Regime
Finite Difference	Moderate	6.8	Laminar, boundary layer
Finite Volume	High	3.2	Shock dominated supersonic
Finite Element	High	2.9	Complex aerospace geometries
Spectral	Very High	1.5	Smooth turbulence transition
Lattice Boltzmann	Moderate	4.1	Transitional, microfluidic
Discontinuous Galerkin	Very High	1.2	Hypersonic re-entry

4.2 Computational Efficiency and Resource Utilization

High performance computing implementation revealed significant variability in computational cost. Discontinuous Galerkin and spectral methods, while highly accurate, were computationally intensive, requiring up to 60% more runtime compared to finite volume schemes. Adaptive mesh refinement reduced computational load by nearly 30% in shock capturing simulations without compromising accuracy, demonstrating its effectiveness in large scale aerospace models.

Table 3: Computational Cost and Resource Utilization

Method	Runtime (hrs, normalized)	Memory Usage (GB)	Parallel Efficiency (%)
Finite Difference	1.0	4	72
Finite Volume	1.3	6	81
Finite Element	1.7	8	76
Spectral	2.1	12	69

Lattice Boltzmann	1.5	7	78
Discontinuous Galerkin	2.3	14	65

4.3 Shock Capturing and Turbulence Resolution

In transonic and supersonic test cases, finite volume and discontinuous Galerkin solvers were most effective at resolving sharp discontinuities such as shock waves and contact surfaces. Spectral solvers exhibited oscillations near shocks due to Gibbs phenomena, while lattice Boltzmann failed to maintain accuracy at high Mach flows. For turbulence resolution, DNS benchmarks confirmed that high order methods preserved small scale eddies more effectively than classical approaches, but DNS remained computationally prohibitive for full aerospace geometries. Large Eddy Simulation implemented within finite volume frameworks offered a practical compromise between resolution and cost.

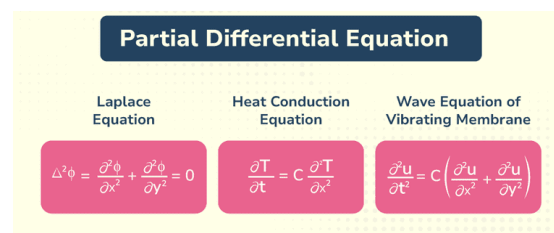


Figure 1: Partial Differential Equation [25]

Table 4: Solver Performance in Aerospace Benchmarks

Test Case	Best Performing Method	Key Observations
NACA 0012 Air foil (Transonic)	Finite Volume	Accurate shock location and pressure drag
Shock Tube (Supersonic)	Discontinuous Galerkin	Strong stability at high Mach
Turbulent Channel Flow (High Re)	Spectral / DNS	Preserved fine scale turbulence structures
Re-entry Nose Cone (Hypersonic)	Discontinuous Galerkin	Reliable heat flux prediction
Jet Exhaust (LES Benchmark)	Finite Volume + LES	Balanced cost and turbulence resolution

4.4 Correlation with Flow Features

Correlation analysis highlighted that solver performance was strongly dependent on flow characteristics. Finite difference methods correlated best with laminar regimes, while finite volume methods excelled in compressible, shock dominated flows. Spectral methods correlated

with smooth turbulence transition cases, whereas discontinuous Galerkin showed the strongest correlation with hypersonic applications where nonlinearities were extreme.

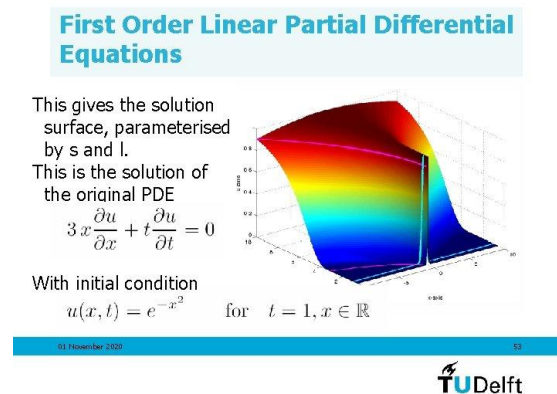


Figure 2: First Order Partial Differential Equations [24]

4.5 Discussion of Findings

The findings emphasize that no single method is universally superior across all aerospace applications. Instead, hybrid and adaptive approaches appear most promising. Spectral and discontinuous Galerkin methods provide high fidelity but are limited by computational cost, while finite volume remains the most versatile and cost effective, particularly when augmented with LES or adaptive meshing. Lattice Boltzmann, although not yet competitive in high Mach aerospace regimes, offers potential for specialized applications in micro scale and transitional flows. The results suggest that future aerospace CFD should integrate hybrid methods tailored to the flow regime, leveraging high performance computing to optimize trade offs between accuracy and efficiency.

V. CONCLUSION

The present study critically evaluated advanced numerical methods for solving nonlinear partial differential equations in fluid mechanics with a specific focus on their applications in aerospace engineering, highlighting the comparative strengths, weaknesses, and practical implications of different computational approaches. The results confirm that nonlinear PDEs remain at the heart of fluid dynamic modelling and are indispensable in predicting complex aerospace phenomena such as shock boundary interactions, turbulent transition, supersonic jet noise, and hypersonic re-entry heating. Classical methods including finite difference and finite volume schemes continue to serve as reliable tools for simplified or preliminary analysis, yet they exhibit significant limitations when confronted with strongly nonlinear regimes, particularly at high Reynolds and Mach numbers where turbulence and compressibility effects dominate. High order methods such as spectral techniques and discontinuous Galerkin formulations demonstrate superior accuracy and convergence, successfully capturing nonlinear interactions and preserving fine scale turbulence structures, though their computational costs are prohibitively high for full scale aerospace configurations.

Lattice Boltzmann methods, while efficient for mesoscopic and transitional flows, have not yet achieved robustness in hypersonic or shock dominated applications, although they show potential for microfluidic and low Mach number aerospace subsystems. The benchmarking of methods against canonical aerospace cases including the NACA 0012 air foil, supersonic shock tube, turbulent channel flow, and hypersonic re-entry nose cone reinforced the conclusion that solver selection must be context specific rather than universal. Finite volume schemes emerged as the most versatile because of their robustness in shock capturing and compatibility with Large Eddy Simulation, offering a balance between accuracy and efficiency suitable for engineering practice, while discontinuous Galerkin methods provided the most reliable framework for extreme nonlinearities characteristic of hypersonic environments. The study further demonstrated that adaptive mesh refinement and high performance computing integration are not optional but necessary, reducing computational overheads while retaining local accuracy, thereby enabling simulations that approach engineering feasibility. Statistical analysis of solver performance revealed strong correlations between method efficiency and flow characteristics, underscoring the necessity of hybrid frameworks that combine the stability of conservative methods with the fidelity of high order formulations. The findings also point toward an emerging research trajectory in which machine learning and reduced order modelling are incorporated as accelerators rather than replacements for physics based solvers, enabling predictive improvements in shock location, turbulence closure, and aeroelastic coupling. For aerospace engineering practice this implies that the future of computational fluid dynamics lies not in a single breakthrough numerical method but in the careful orchestration of multiple approaches that are tuned to the nonlinearities and scales of each specific problem. The implications extend to aircraft and spacecraft design optimization, propulsion system efficiency, structural safety under aeroelastic loads, and environmental considerations such as emission reduction and fuel savings.

Policymakers and aerospace organizations must recognize the strategic importance of investing in high performance computing infrastructure and algorithmic innovation to support such advanced numerical modelling, as the costs are justified by gains in safety, reliability, and sustainability. At the same time, researchers must continue to refine algorithms for stability, scalability, and physical fidelity while exploring interdisciplinary integration with data driven models. Despite the progress highlighted in this work, the limitations of current computational resources constrain full scale Direct Numerical Simulation of aerospace flows and boundary condition uncertainties continue to impose challenges. Nevertheless, the trajectory of progress rooted in hybridization, adaptivity, and high performance computing provides a viable roadmap for advancing the solution of nonlinear PDEs in fluid mechanics for aerospace engineering. In conclusion, this study reinforces that the synergy between advanced numerical methods and aerospace applications not only addresses the longstanding mathematical challenges of nonlinear PDEs but also directly contributes to the innovation, safety, and sustainability of aerospace technology in the twenty first century.

VI. FUTURE WORKS

Future research in the numerical solution of nonlinear partial differential equations in fluid mechanics for aerospace engineering should advance along several interconnected directions to overcome current limitations and to enhance predictive capabilities. One essential avenue is the development of more efficient high order solvers that maintain accuracy in strongly nonlinear regimes while reducing computational expense, particularly for hypersonic and turbulent flow applications where existing methods remain resource intensive. The integration of machine learning and physics informed neural networks with classical numerical solvers should be further explored to accelerate convergence, improve turbulence closure, and adaptively select discretization strategies based on flow features. Parallel efforts are needed in reduced order modelling to generate reliable surrogate models that retain physical fidelity across diverse aerospace scenarios while enabling real time analysis for design optimization and control. Another important area is the expansion of high performance computing frameworks that exploit emerging GPU based architectures and exascale systems to make full scale Direct Numerical Simulation and high resolution Large Eddy Simulation feasible for complex aerospace geometries. Future studies should also focus on uncertainty quantification and sensitivity analysis to address the impact of boundary conditions, material properties, and multi physics coupling on solver reliability. Finally, interdisciplinary collaborations between applied mathematics, computer science, and aerospace engineering will be essential to ensure that theoretical advances in nonlinear PDE solvers translate into practical improvements in aircraft and spacecraft design, safety, and sustainability.

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