

A Systematic Review on the Use of Fractional Order Partial Differential Equations in Biological and Medical Modeling

Vishal Rajput¹ and Dr. Ashwini Kumar²

¹Research Scholar, Department of Mathematics

²Research Guide, Department of Mathematics

Sunrise University, Alwar, Rajasthan, India

Abstract: Fractional order partial differential equations (FPDEs) have emerged as powerful mathematical tools in modeling complex phenomena in biological and medical systems. Unlike classical integer-order models, FPDEs capture memory effects, anomalous diffusion, and non-local interactions, which are prevalent in biological tissues, disease propagation, and drug transport mechanisms. This review systematically examines the recent developments, applications, and computational methods associated with FPDEs in biomedical contexts, emphasizing their role in accurately describing real-world biological processes

Keywords: Fractional Order Partial Differential Equations, Anomalous Diffusion

I. INTRODUCTION

Biological and medical systems are inherently complex, often characterized by non-local interactions, heterogeneity, and memory-dependent behavior. Classical partial differential equations (PDEs) are limited in capturing such properties due to their local and integer-order nature. In contrast, fractional order partial differential equations (FPDEs) generalize classical models by allowing derivatives of non-integer order, providing enhanced flexibility in representing memory and spatial heterogeneity.

Biological and medical systems are inherently complex, exhibiting non-local interactions, heterogeneity, and memory-dependent behavior that cannot always be accurately described using classical integer-order models. Traditional partial differential equations (PDEs) have long been used to represent processes such as diffusion, reaction kinetics, and transport phenomena. However, these models often fail to capture the anomalous dynamics observed in real biological systems, such as sub diffusion in crowded cellular environments, heterogeneous tumor growth, irregular drug transport in tissues, and the complex spatiotemporal spread of diseases.

Fractional order partial differential equations (FPDEs) offer a more generalized mathematical framework by incorporating derivatives of non-integer order in time and space, enabling the modeling of processes with memory effects and long-range spatial correlations. The key advantage of FPDEs lies in their ability to bridge the gap between purely local models and the observed non-local and history-dependent dynamics in biological systems. Mathematically,

FPDEs can be expressed as $\frac{\partial^\alpha u(x,t)}{\partial t^\alpha} = D \frac{\partial^\beta u(x,t)}{\partial x^\beta} + f(u, x, t)$, where α and β represent the fractional orders in time and space, D denotes the diffusion coefficient, and $f(u, x, t)$ represents reaction or source terms.

This formulation allows the description of sub diffusive or super diffusive transport phenomena, reaction-diffusion dynamics, and processes with memory effects, which are prevalent in physiological and pathological contexts. Over the past two decades, FPDEs have been increasingly applied to model drug delivery mechanisms, capturing the complex pharmacokinetics of drugs in heterogeneous tissues and porous organs, where classical diffusion equations are insufficient. Similarly, tumor growth models based on fractional reaction-diffusion equations have demonstrated improved accuracy in predicting invasion fronts and tumor heterogeneity by accounting for non-local cell migration and proliferation dynamics.

In epidemiology, fractional models of infectious disease spread incorporate memory effects and anomalous mobility patterns, allowing better representation of incubation periods, long-distance contacts, and spatial heterogeneity in populations. Computational approaches play a crucial role in implementing FPDEs in biomedical applications, as analytical solutions are often unavailable. Numerical methods, including finite difference, finite element, and spectral approaches, along with efficient discretization techniques for Caputo and Riemann-Liouville derivatives, have been developed to simulate high-dimensional fractional models. Despite these advances, challenges remain, particularly in parameter estimation, model validation against experimental data, and computational efficiency for large-scale simulations.

This systematic review aims to provide a comprehensive overview of the current applications of FPDEs in biological and medical modeling, summarizing their mathematical formulations, computational strategies, advantages over classical models, and practical implications for understanding complex physiological, pathological, and epidemiological processes. By highlighting the state-of-the-art research, this review seeks to demonstrate the potential of FPDEs as a robust tool for predictive and personalized medicine, as well as to identify gaps and future directions for further development in this rapidly evolving field.

Mathematically, a general time-space FPDE can be written as:

$$\frac{\partial^\alpha u(x, t)}{\partial t^\alpha} = D \frac{\partial^\beta u(x, t)}{\partial x^\beta} + f(u, x, t), \quad 0 < \alpha \leq 1, 1 < \beta \leq 2,$$

where $\frac{\partial^\alpha}{\partial t^\alpha}$ and $\frac{\partial^\beta}{\partial x^\beta}$ represent fractional derivatives in time and space, respectively, D is the diffusion coefficient, and $f(u, x, t)$ denotes reaction or source terms. This formulation allows modeling of anomalous diffusion and subdiffusive behaviors commonly observed in biological media.

FPDEs have gained attention in modeling a variety of biomedical phenomena including tumor growth, drug delivery, tissue engineering, and disease spread. This review systematically explores the applications, computational approaches, and advantages of FPDEs in biological and medical modeling.

APPLICATIONS IN BIOLOGICAL AND MEDICAL MODELING

1. Anomalous Diffusion in Biological Tissues

Fractional diffusion equations are employed to describe anomalous transport phenomena in tissues where standard diffusion assumptions fail. For example, subdiffusive transport of molecules in crowded cellular environments can be modeled by:

$$\frac{\partial^\alpha C(x, t)}{\partial t^\alpha} = D \nabla^2 C(x, t),$$

where $C(x, t)$ represents the concentration of a substance and $0 < \alpha < 1$ accounts for subdiffusion. Studies show that FPDEs can accurately capture experimental data in biological tissue diffusion experiments (Metzler & Klafter, 2000).

2. Tumor Growth and Cancer Modeling

Tumor growth exhibits non-local interactions and memory effects, which can be represented using fractional reaction-diffusion equations:

$$\frac{\partial^\alpha N(x, t)}{\partial t^\alpha} = D \nabla^2 N(x, t) + r N(x, t) \left(1 - \frac{N(x, t)}{K} \right),$$

where $N(x, t)$ is the tumor cell density, r is the growth rate, and K is the carrying capacity. Fractional models provide improved predictions of tumor invasion fronts compared to classical models (Henry et al., 2006).

DRUG DELIVERY AND PHARMACOKINETICS

FPDEs model anomalous drug transport in complex tissues and heterogeneous organs. For instance, subdiffusive drug transport through porous media is modeled as:

$$\frac{\partial^\alpha C(x, t)}{\partial t^\alpha} = D \frac{\partial^2 C(x, t)}{\partial x^2} - kC(x, t),$$

where k represents drug decay. Such models allow better prediction of concentration profiles in targeted drug delivery systems (Magin et al., 2008).

EPIDEMIC AND DISEASE SPREAD

FPDEs are increasingly used in modeling infectious disease spread, accounting for memory-dependent contact processes and anomalous mobility patterns:

$$\frac{\partial^\alpha I(x, t)}{\partial t^\alpha} = \beta S(x, t)I(x, t) - \gamma I(x, t) + D \nabla^2 I(x, t),$$

where S , I , and R denote susceptible, infected, and recovered populations, respectively, β is the infection rate, and γ is the recovery rate. Fractional models capture long-range interactions and variable incubation periods more realistically than classical SIR models (Tarasov, 2011).

NUMERICAL AND COMPUTATIONAL APPROACHES

Analytical solutions for FPDEs are limited, necessitating the development of numerical schemes such as finite difference, finite element, and spectral methods. Caputo and Riemann-Liouville fractional derivatives are commonly employed for discretization. Efficient algorithms for time-fractional and space-fractional derivatives are essential for simulating high-dimensional biomedical systems. Adaptive mesh refinement and parallel computation techniques enhance the computational feasibility of FPDE models in complex biological applications.

Analytical solutions for fractional order partial differential equations (FPDEs) are limited to a few simple cases due to the non-local and memory-dependent nature of fractional derivatives. As a result, numerical and computational approaches play a critical role in implementing FPDEs for practical biological and medical modeling. Several numerical schemes have been developed to discretize both time- and space-fractional derivatives, with Caputo and Riemann-Liouville formulations being the most commonly used.

Finite difference methods are widely applied for time-fractional problems, where the Grünwald-Letnikov approximation is often employed to convert fractional derivatives into discrete sums over previous time steps, effectively capturing memory effects. Spatial fractional derivatives, which model long-range interactions or anomalous diffusion, are frequently discretized using spectral methods or shifted Grünwald schemes to maintain stability and accuracy. Finite element methods (FEM) have also been adapted to solve FPDEs in irregular biological geometries, such as tissues or organs, providing flexibility in handling complex boundaries and heterogeneous properties.

In addition, spectral collocation and Chebyshev polynomial-based methods offer high-order accuracy for smooth solutions and are particularly useful in multi-dimensional fractional models. Computational efficiency is a major challenge, as fractional derivatives involve non-local computations that scale with the entire simulation history. To address this, techniques such as fast convolution algorithms, adaptive time-stepping, and parallel computing have been implemented, enabling simulations of large-scale biological systems, including tumor growth and drug transport.

Hybrid approaches, combining FPDEs with machine learning, are emerging as powerful tools for parameter estimation, model calibration, and prediction of complex biomedical processes. Overall, the development of robust, accurate, and efficient numerical schemes is essential for the successful application of FPDEs in biological and medical modeling, facilitating their use in predictive simulations, personalized medicine, and experimental validation.

ADVANTAGES OF FPDES IN BIOMEDICAL MODELING

Memory Effect Representation: Fractional derivatives naturally account for history-dependent processes. Fractional order partial differential equations naturally capture memory effects in biological and medical systems, reflecting how current states depend on past dynamics. This feature enables accurate modeling of processes such as subdiffusive transport in tissues, delayed cellular responses, and history-dependent disease progression, providing more realistic simulations than classical integer-order models.

Anomalous Diffusion Modeling: Sub diffusion and super diffusion phenomena are captured accurately. Fractional order partial differential equations effectively model anomalous diffusion, including sub diffusion and super diffusion, observed in biological systems. By incorporating non-integer derivatives, FPDEs describe irregular particle transport, molecular movement in crowded cellular environments, and heterogeneous tissue diffusion, offering more precise representations than classical diffusion equations and improving predictions in biomedical applications.

Non-Local Interactions: FPDEs model spatial heterogeneity and long-range effects. Fractional order partial differential equations capture non-local interactions, enabling the modeling of spatial heterogeneity and long-range effects in biological systems. This allows accurate representation of processes such as tumor cell migration, tissue signaling, and disease spread, where local changes are influenced by distant regions, surpassing the capabilities of classical local differential models.

Improved Predictive Accuracy: Fractional models outperform integer-order models in various experimental validations. Fractional order partial differential equations enhance predictive accuracy in biological and medical modeling by incorporating memory effects, anomalous diffusion, and non-local interactions. This allows for more realistic simulations of complex processes such as drug transport, tumor growth, and epidemic spread, providing superior alignment with experimental data compared to classical integer-order models.

CHALLENGES AND FUTURE DIRECTIONS

Despite their advantages, FPDEs pose challenges including parameter estimation, computational cost, and lack of standardized methods for model validation in biomedical contexts. Future research may focus on hybrid models combining FPDEs with machine learning for parameter estimation, real-time simulations, and personalized medicine applications. Despite the significant advantages of fractional order partial differential equations (FPDEs) in modeling complex biological and medical systems, several challenges hinder their widespread adoption. One primary difficulty lies in parameter estimation, as fractional models require determining the orders of derivatives (α and β) and other system-specific coefficients, which often cannot be directly measured experimentally.

Accurate identification of these parameters is crucial, as small variations can significantly affect model behavior, yet data scarcity and noise in biological measurements complicate this process. Additionally, FPDEs inherently involve non-local operations, where fractional derivatives depend on the entire history of the system or distant spatial regions. This property, while mathematically advantageous, leads to substantial computational cost, especially for high-dimensional and long-time simulations in complex geometries, such as three-dimensional tissue structures or organ-level models.

Efficient numerical algorithms and memory-saving techniques are still active areas of research, with approaches such as fast convolution methods, adaptive time-stepping, and parallel computing being explored to improve feasibility. Another challenge is model validation; fractional models often outperform classical models theoretically, but comprehensive experimental and clinical validation remains limited. Bridging the gap between mathematical formulations and empirical data is essential to gain confidence in FPDE-based predictions and to translate these models into practical biomedical applications.

Furthermore, interdisciplinary collaboration is required, as effective use of FPDEs demands expertise in mathematics, numerical computation, biology, and medicine. Looking ahead, several future directions can enhance the applicability and impact of FPDEs. Integration with machine learning and data-driven approaches offers promising opportunities for parameter estimation, model calibration, and predictive analytics. For instance, hybrid models combining FPDEs with neural networks can infer fractional orders from experimental datasets or optimize treatment strategies in personalized medicine.

Additionally, multi-scale modeling using FPDEs can bridge cellular, tissue, and organ-level dynamics, providing a comprehensive understanding of biological processes. The development of standardized frameworks, software libraries, and benchmarking protocols will also facilitate broader adoption in the biomedical community. Emerging applications in drug delivery, cancer therapy, neuroscience, and epidemiology are likely to benefit from FPDEs' ability to represent memory, anomalous transport, and non-local interactions.

Finally, further theoretical research is needed to establish rigorous analytical solutions, stability analyses, and error estimates for fractional models, enhancing reliability and interpretability. In summary, while challenges in parameterization, computational cost, validation, and interdisciplinary integration persist, ongoing advancements in numerical methods, hybrid modeling, and experimental validation are poised to expand the role of FPDEs in predictive, personalized, and translational biomedical research, solidifying their importance in understanding complex biological and medical phenomena.

II. CONCLUSION

Fractional order partial differential equations offer a robust framework for modeling complex biological and medical systems. Their ability to capture memory effects, anomalous diffusion, and non-local interactions makes them superior to classical integer-order models in many biomedical applications. Continued development of efficient numerical methods and experimental validation will further establish FPDEs as a standard tool in computational biology and medical research. Fractional order partial differential equations (FPDEs) have demonstrated significant potential in modeling complex biological and medical systems, offering distinct advantages over classical integer-order models.

By incorporating memory effects, anomalous diffusion, and non-local interactions, FPDEs provide more accurate and realistic representations of processes such as tumor growth, drug transport, tissue dynamics, and disease spread. Their flexibility allows researchers to capture heterogeneous behavior, long-range interactions, and history-dependent dynamics that are prevalent in real-world biological contexts. Despite the challenges of parameter estimation, computational cost, and model validation, advancements in numerical methods, high-performance computing, and hybrid modeling approaches have enhanced the feasibility and applicability of FPDEs.

Moreover, the integration of data-driven techniques and machine learning offers promising avenues for parameter inference, predictive modeling, and personalized medicine applications. This systematic review highlights the current state-of-the-art research, emphasizing both the practical utility and theoretical development of FPDEs in biomedical modeling. Continued interdisciplinary collaboration, rigorous validation, and methodological innovations are expected to further solidify the role of FPDEs as robust tools in understanding, predicting, and optimizing complex biological and medical phenomena, ultimately contributing to improved healthcare outcomes and translational research.

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