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EDITED AND REVIEWED BY Quan Zou, University of Electronic Science and Technology of China, China

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RECEIVED 23 November 2023 ACCEPTED 23 November 2023 PUBLISHED 29 November 2023

CITATION

Xi J, Yu Z and Shi W (2023), Editorial: Computational mechanism of genetic/ evolutionary operator and optimizations in genomic data applications. *Front. Genet.* 14:1343199. doi: 10.3389/fgene.2023.1343199

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Editorial: Computational mechanism of genetic/ evolutionary operator and optimizations in genomic data applications

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KEYWORDS

genetics, evolution, optimization, genomics, computational mechanism

Editorial on the Research Topic

Computational mechanism of genetic/evolutionary operator and optimizations in genomic data applications

Introduction

The exponential growth of genomic data, driven by advancements in high-throughput sequencing technologies, has precipitated the need for innovative data management solutions (Vandereyken et al., 2023). The challenge extends beyond mere storage to encompass the swift transmission and processing of large datasets, which are essential for timely and effective data analysis and interpretation (Caudai et al., 2021; Yan et al., 2022). The principles of genetics and evolutionary theory have fundamentally explained the emergence and progression of the biological realm, bringing transformative concepts of growth and variation to biology (Moczek et al., 2015). This evolutionary perspective has not only accelerated advancements in genetic research but also significantly propelled other scientific fields forward (Shi et al., 2021; Diaz-Flores et al., 2022). Motivated by genetic and evolutionary theories, researchers have developed numerous computational strategies rooted in genetic and evolutionary operations and stochastic search techniques (Ünal and Başçiftçi, 2022).

In recent times, the application of genomic data has encountered optimization challenges where conventional mathematical approaches fall short (Zhou et al., 2019). Genetic and evolutionary algorithms stand apart from traditional calculus-based and exhaustive methods due to their ability to achieve global optimization with remarkable robustness and broad applicability (Viriyasitavat et al., 2021; Shi et al., 2022). Characterized by self-organization, self-adaptation, and self-learning, these algorithms can adeptly handle complex optimizations irrespective of the problem's nature (Gheibi et al., 2021; Liu et al., 2022). Genomic data analysis often deals with intricate patterns and complex regulations unsuitable for traditional optimization methods (Hassija et al., 2023; Xi et al., 2023). Integrating genetic and evolutionary algorithm-based complex optimizations in genomic analysis can alleviate bottlenecks in bioinformatics tasks (Xi

et al., 2020a; Mandal et al., 2023). Therefore, this research theme focuses on exploring complex optimization challenges in genomic data applications using genetic and evolutionary algorithms (Jiao et al., 2023).

Advancements in genomic sequencing technology have led to an explosion of data, marking the advent of the genomics big data era (Xi et al., 2020b). This surge brings both possibilities and challenges, especially in terms of data storage, processing, and interpretation (Ahmed et al., 2022). The articles in this special edition, titled "Computational Mechanism of Genetic/Evolutionary Operator and Optimizations in Genomic Data Applications," collectively tackle these issues through cutting-edge computational methods. They delve into the complexities of genomic data and introduce innovative ways to optimize its application across various biological and clinical settings.

Optimizing genomic data storage and processing

In this Research Topic, the paper titled "Enhancing Genomic Mutation Data Storage Optimization based on the Compression of Asymmetry of Sparsity" tackles the formidable challenge of managing the deluge of genomic data. It presents a novel compression algorithm, CA_SAGM, specifically designed for sparse asymmetric gene mutations (Ding et al.). This development is particularly pertinent for massive genomic databases like The Cancer Genome Atlas (TCGA), where efficient data handling is paramount. The study's comparative analysis of CA_SAGM with other algorithms underscores the critical role of data compression in navigating the complexities of large-scale genomic datasets.

Advancing genomic research through computational estimation techniques

Building on the theme of computational innovation, the paper "A Noise-tolerance Learning Method for Efficiently Estimating Open Chromatin Regions via cfDNA sequencing data" focuses on open chromatin regions (OCRs), crucial for understanding cellular functions and gene expression (Ren et al.). The introduction of OCRFinder, a learning-based, noise-tolerant approach, marks a significant stride in addressing the dynamic challenges of chromatin accessibility in cfDNA-seq data. By integrating ensemble learning and semi-supervised strategies, this study exemplifies the importance of sophisticated computational methods in genomic research, especially in areas like chromatin accessibility.

Genomic data in clinical and prognostic applications

Shifting the spotlight to clinical implications, "Development and validation of focal adhesion-related genes signature in gastric cancer" illustrates the power of genomic data in disease prognosis (Zhao et al.). This paper offers a prognostic signature based on focal adhesion-related genes, showcasing how genomic data can be instrumental in identifying critical prognostic genes for gastric cancer. This research bridges computational methodologies

and genomic insights, enhancing our understanding of cancer biology and providing valuable tools for cancer prognosis.

Exploring disease mechanisms through gene modification analysis

The issue concludes with "Comprehensive Analysis of Key m5C Modification-Related Genes in Type 2 Diabetes," which delves into the role of 5-methylcytosine (m5C) RNA methylation in type 2 diabetes (T2D) (Song et al.). Employing a variety of computational techniques, including LASSO regression and Gene Set Enrichment Analysis, the study sheds light on the molecular mechanisms of T2D. It highlights potential biomarkers and therapeutic targets, demonstrating the utility of genomic data in deciphering the complexities of disease processes.

Collectively, these articles represent the multifaceted applications of computational techniques in genomic data analysis. From optimizing data storage and processing to enhancing disease prognosis and understanding molecular mechanisms, these studies underscore the transformative impact of computational methods in the era of genomic big data. The innovations and insights showcased in this Research Topic are set to significantly shape future research and applications in genomics, bridging computational prowess with biological discovery.

Author contributions

JX: Conceptualization, Funding acquisition, Writing-original draft, Writing-review and editing. ZY: Conceptualization, Funding acquisition, Writing-original draft, Writing-review and editing. WS: Conceptualization, Funding acquisition, Writing-original draft, Writing-review and editing.

Funding

The author(s) declare financial support was received for the research, authorship, and/or publication of this article. This work is supported partially by National Natural Science Foundation of China (Grant No. 62202117), partially by the Tertiary Education Scientific Research Project of Guangzhou Municipal Education Bureau (No. 202235388), partially by the Special Foundation in Department of Higher Education of Guangdong (Grant No. 2022ZDX 2053), partially by the Guangzhou Basic and Applied Basic Research Foundation (No. 2023A04J0386).

Acknowledgments

We would like to thank Dr. Janet Ajibade for her helpful suggestions on organizing this Research Topic.

Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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