

A Survey on High Performance Computing and its Forms

Course Survey Paper

Gauri Shashank Moghe
gaurishashank.moghe01@student.csulb.edu
Deep Kapadia
deep.kapadia01@student.csulb.edu
Priya Shah
PriyaSunilkumar.Shah01@student.csulb.edu
Raj Patel
raj.patel01@student.csulb.edu

CECS 530 - Advanced Computer Architecture Computer Engineering and Computer Science Department CSULB Fall-2023

Instructor:

Dr. Maryam S. Hosseini

December 10, 2023

Abstract

The relentless surge in data generation across diverse fields have propelled a pressing demand for efficient storage and processing capabilities. This paper addresses this imperative need for high-performance computing (HPC), focusing on its pivotal role in handling computationally complex data. In fields such as numerical simulations and predictions, weather forecasting, climate modeling, aerospace, and automotive engineering, the colossal volume of information necessitates advanced computational infrastructure for meaningful analysis and decision-making. In the ever-evolving realm of high-performance computing (HPC), diverse forms and paradigms have emerged to address the escalating demands of computational power across various domains. This survey paper navigates the multifaceted landscape of HPC, providing an in-depth analysis of different computing paradigms, each tailored to meet specific computational challenges. The paper unfolds the historical evolution, architectural intricacies, and practical applications of prominent HPC forms include supercomputing, cloud computing, cluster computing, and grid computing. Through systematically exploring each paradigm, the survey elucidates the underlying principles, technological advancements, and critical methodologies shaping the current state of HPC. Comparative analyses highlight these forms' strengths, weaknesses, and complementary aspects, offering valuable insights for researchers, practitioners, and decision-makers seeking optimal solutions for diverse computational tasks. The survey also addresses emerging trends, challenges, and future directions, illuminating the trajectory of HPC as it redefines the boundaries of computational capabilities. This paper serves as a comprehensive reference, providing a road map for understanding, navigating, and harnessing the power of different forms of high-performance computing in the modern computational landscape.

Contents

Abstract						
1						
2						
	2.1	Backg	round	3		
		2.1.1	Historical development	3		
		2.1.2	Importance of the evolution of HPC	5		
	2.2	Relate	ed work	6		
	2.3	Metho	odology	7		
3	Discussion and Findings					
	3.1	Cloud	Computing	9		
		3.1.1	Architectural Framework	10		
		3.1.2	Security Considerations	10		
		3.1.3	Performance Evaluation	11		
	3.2	Grid (Computing	11		
		3.2.1	Grid Architecture	12		
		3.2.2	Scheduling strategies	12		
		3.2.3	Fault Tolerance	12		
		3.2.4	Performance Evaluation	12		
	3.3	Cluste	er Computing	12		
		3.3.1	Role of Containers/Clusters	13		
		3.3.2	Cluster Techniques	13		
		3.3.3	Orchestration Systems	13		
		3.3.4	Fault Tolerance	14		
	3.4	Super	Computing	15		
		3.4.1	Parallel Processing	16		

Contents							
		3.4.2	Evaluation Methods	17			
		3.4.3	Benchmarks and Simulation	17			
	3.5 Benefits of using HPC Systems			17			
	3.6	Limita	ations and gaps in HPC systems	18			
	3.7	Future	e scope	19			
4	4 Conclusion			20			
Bi	Bibliography						

Introduction

In our data-driven era, the uninterrupted surge in information production from various sources, especially in fields like numerical simulations, weather forecasting, climate modeling, and aerospace and automotive engineering, has ushered in an era where efficient storage and processing of computationally complex data are imperative. This burgeoning demand for computational prowess has propelled High-Performance Computing (HPC) into the forefront, serving as the linchpin for tackling the challenges posed by vast datasets and intricate simulations.

High-performance computing, commonly called HPC, stands as a "paradigm in computing that involves the strategic utilization of supercomputers and parallel processing methods to perform intrinsic and demanding computational tasks". The essence of HPC lies in its ability to harness the collective computational power of multiple processing units, enabling the rapid execution of calculations that would be insurmountable for traditional computing systems.

The indispensability of HPC systems in contemporary scientific and industrial landscapes cannot be overstated. As articulated by [1], the influx of data from diverse domains necessitates sophisticated decision support systems, underlining the pivotal role of HPC in numerical simulations, predictions, and fields such as weather forecasting, climate modeling, and aerospace and automotive engineering. The potential to efficiently collect, process, and analyze vast datasets is the lifeblood of advancements in these domains. The landscape of HPC is diverse, encompassing various forms tailored to specific computational needs. From the foundational cluster computing elucidated by [1] to the cutting-edge quantum computing explored by [2] and the container-based orchestration systems meticulously categorized by [3], HPC manifests in myriad forms. Each form addresses unique challenges and brings forth distinct advantages, contributing to the versatility of HPC solutions.

While HPC stands as a cornerstone in computational progress, it is essential to acknowledge certain disadvantages. The work by [4] highlights reliability and anomaly detection concerns in HPC systems. The sophisticated nature of these systems necessitates intricate fault-mitigation strategies, and challenges persist in achieving optimal performance, especially as we move towards exascale com-

1. Introduction 2

puting. The evolutionary trajectory of HPC systems is *intricately woven into the fabric of technological advancements*. From the pioneering days of supercomputers in the 1940s, as chronicled in [1], to the contemporary era where quantum computing and container-based technologies redefine the paradigm, HPC systems continue to evolve. The ongoing exploration into GPU architectures [5] and innovative clustering techniques [6] underscores the dynamic nature of HPC evolution.

This paper provides a panoramic view of the imperative role HPC plays in the face of escalating data complexities. The exploration encompasses the definition of HPC, its critical importance in diverse domains, the diverse forms it takes, associated challenges, and the evolutionary journey of HPC systems. As we delve into subsequent sections, drawing insights from seminal works, we aim to unravel the nuanced facets of HPC, bridging the theoretical realms explored in research with the practical implications for the computational landscapes of today and tomorrow.

Literature Survey

2.1 Background

This review seeks to concisely show the historical progression of High-Performance Computing (HPC) from the 1940s to the current day [7][5]. It covers significant stages such as the early computer period, the Cray era, the cluster era with Beowulf clusters, and the current GPU and hybrid era [8]. The background also discusses the integration of artificial intelligence and the importance of various computing paradigms, such as supercomputing, cluster computing, grid computing, and cloud computing. The goal is to give a concise yet thorough overview, laying the groundwork for further investigating technologies and methodologies within each kind of HPC.

2.1.1 Historical development

The history of High-Performance Computing (HPC) is divided into various eras, each marking a key milestone in computing advancement. Early computers such as UNIVAC and ENIAC laid the basis in the 1940s and 1960s [7]. The 1960s and 1990s saw the advent of supercomputing, with machines such as the CDC 6600 and Cray-1 setting speed records. "Seymour Cray's" designs came to be associated with high-speed calculation. Beowulf clusters rose to prominence during the 1990s-2010 cluster period, utilizing commodity hardware for cost-effective HPC [8]. Grid computing arose concurrently, linking distant computers into a virtual supercomputer. The GPU and hybrid era, which began in 2010, delivered enhanced processing capability, most notably through Graphics Processing Units (GPUs) and specialized hardware. The potential for quantum computing to transform HPC drew attention, while artificial intelligence integration became a key theme. Throughout these eras, supercomputing, cluster computing, grid computing, and cloud computing have played critical roles. Supercomputers such as the CDC 6600 and Cray-1 have shown impressive capabilities. Beowulf clusters and grid computing solved scalability and cooperation issues. Cloud computing's function evolved, becoming a popular, low-cost choice in supercomputing. This concise historical review highlights the dynamic history of HPC, from early computers to current breakthroughs across many computing paradigms.

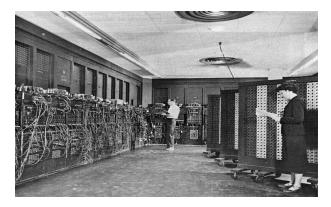


Figure 2.1: ENIVAC



Figure 2.2: UNIVAC

- 1. Machine Learning Methodologies: Overview of machine learning approaches, emphasizing the difficulties in applying classical supervised learning to High-Performance Computing (HPC) anomaly detection due to a lack of labeled data [4][6].
- 2. **Semi-supervised learning is being introduced**: The recognized need for semi-supervised learning, which uses a combination of labeled and unlabeled input for training, particularly in dynamic contexts such as HPC systems [4].
- 3. The Role of Autoencoders: Autoencoders, a specific neural network architecture, are introduced. Autoencoders are wellsuited for anomaly detection in HPC because of their dimensionality reduction and feature learning capabilities [6].

- 4. **Anomaly Detection Using Autoencoders**: Explanation of how autoencoders work in anomaly detection, including learning to rebuild standard patterns and finding abnormalities based on more significant reconstruction errors [3].
- 5. Anomaly Detection in HPC in Real Time: The goal of the conceptual framework is to provide a methodology for real-time anomaly identification in HPC systems utilizing a semi-supervised learning approach using autoencoders while accounting for the dynamic character of HPC settings [9].



Figure 2.3: GPU AND HYBRID ERA

2.1.2 Importance of the evolution of HPC

Fundamental concepts such as High-Performance Computing (HPC), Anomaly Detection, Machine Learning (ML), semi-supervised learning, Autoencoder, Dimensionality Reduction, and Feature Learning [1][4] are introduced. HPC is the use of modern computer technology to solve complicated problems quickly. Anomaly detection detects departures from the norm, which is critical for assuring the stability of HPC systems. When labeled data is insufficient, ML, particularly semi-supervised learning, assists in anomaly detection [1]. Autoencoders, a neural network, help with this process by learning efficient data representations. Dimensionality reduction reduces computer complexity, whereas feature learning identifies significant patterns from unstructured data [4]. These ideas form the conceptual framework, emphasizing the interaction of HPC, anomaly detection, and machine learning in the research setting. The relevance of the research by addressing the increasing requirement for effective processing of large-scale data in modern computer settings [6] is underlined. The exponential rise of Internetgenerated information, including structured and unstructured data, demands modern processing and analysis technology. Cloud computing develops as a critical alternative for massive dataset storage and administration, leveraging parallel and map reduction approaches [6]. Despite the availability of these approaches, there are still obstacles in processing large amounts of data. Hadoop's distributed file system solves previous challenges, enabling high-performance computing in cloud environments [6]. The combination of a single cluster with the Analytic Hierarchy Process (AHP) improves fault tolerance, decreases processing time, and eliminates communication issues, all of which contribute to an efficient solution for the challenges provided by big data in cloud computing.

2.2 Related work

High-performance computing (HPC) has been the subject of numerous studies in the past, which have shown its wide range of uses and configurations. A noteworthy study by Zhang et al. [5] examines HPC architectures in great detail, emphasizing the developments and essential characteristics that have influenced the field of high-performance computing. Many studies have focused on the importance of containers in high-performance computing (HPC), highlighting how vital a role they play in resource deployment [3][6][10]. Notably, Bachiega et al. [10] have elucidated the elements and advantages of containerization in the context of HPC by thoroughly analyzing the difficulties associated with assessing performance in container-based systems.

Cluster computing and orchestration systems have been the focus of efforts to improve resource management and scheduling [9][11][12]. Jacek Burys and colleagues [12] have developed a parallel iteration of the Canopy Clustering algorithm using MPI to overcome synchronization issues in extended clustering. Research has been conducted on evaluating performance in high-performance computing (HPC), and several publications have proposed innovative methods [1][4][3][12]. In a survey on performance evaluation with containers, Naylor G. Bachiega et al. [10] stress the significance of this evaluation and offer insightful information on the approaches and difficulties involved in assessing HPC performance.

High-Performance Computing (HPC) security issues have been the subject of numerous investigations, focusing on "container security" [10]. An overview of current research and challenges about containers can be found in Naylor G. Bachiega et al.'s study [10]. Their research also explores security elements and aspects of container HPC environments. Many aspects of HPC present obstacles and suggest future directions, such as complex architecture, security concerns, and scalability constraints. [3][6][10][12]. Combining the problems and solutions suggested by these sources leads to a thorough understanding of the current and future directions in HPC technological advancement. Investigating HPC architectures has given us a thorough understanding of the origins and evolution of these systems [5]. Zhang et al.'s research is essential for understanding the high-performance computing architecture landscape [5]. Taken as a whole, these articles provide a helpful collection of relevant literature that serves as a foundation for our detailed analysis of several high-performance computing models.

Every citation adds distinctive viewpoints, methods, and ideas that broaden our comprehension of HPC in its manifestations.

2.3 Methodology

Recent years have seen a substantial evolution in high performance computing (HPC), a topic that has been thoroughly reviewed by many intellectuals [8][7][2]. An extensive analysis covering HPC designs, containerization, cluster computing, and security considerations is necessary, given the explosion of large-scale computing systems and the increasing need for efficient processing. The ever-changing HPC ecosystem and growing dependence on advanced computing paradigms drive this survey [7][8][2]. For computational technologies to evolve in the future, it will be essential to understand the problems, developments, and possible solutions in high-performance computing.

This survey was developed to give a thorough overview of high-performance computing by addressing important topics such as cluster computing, containerization, HPC designs, performance evaluation, security considerations, and new hardships [8][2]. Searched Springer, IEEE, ACM, and Scopus to create an extensive collection. The inclusion criteria outlined in [10] are prudently utilized to screen works that correspond with the survey's purview, ensuring the assimilation of literature that substantially advances our knowledge of HPC, its architectures, and associated technologies. The foundation for the analysis is formed by the insightful assistance made by each publication on topics including cluster computing, containerization, HPC architectures, and performance evaluation. Thematic analysis organizes and combines literature data to clarify significant articles and patterns. This methodology allows an organized investigation of several aspects of high-performance computing[3][10].

An extensive analysis of HPC architectures is given [5]. Getting the evolution of high-performance computing requires an awareness of the architectural terrain. Analyzing the evolution of HPC designs [5] provides an analysis of the historical context of the development and progress of these systems, in addition to the revolutionary transformations that have shaped the direction of high-performance computing. In HPC, containers have become a critical tool that facilitates resource usage and deployment [6][10]. Their function in enhancing efficiency and scalability is examined, offering senses into current containerization techniques. Performance improvements and potential roadblocks are highlighted in an examination of the advantages and difficulties of containerization in high-performance computing [3][6][10]. Comprehending these facets is vital for making knowledgeable choices regarding HPC implementation. Cluster computing techniques are elucidated, drawing from studies that delve into efficient resource management and task scheduling. The section provides insights into the myriad methodologies operated in cluster environments. Orchestration systems [11][9][12] are vital in

managing containerized applications within clusters. An exploration of prominent orchestration frameworks contributes to a thorough understanding of cluster computing paradigms.

A critical examination of performance evaluation approaches in HPC is offered, drawing on insights from research that propose novel assessment strategies [1]. This section aims to outline practical methods for assessing computing performance. Simulation and benchmarking in HPC performance assessments are discussed, providing insight into strategies for evaluating system scalability and efficiency [1]. A sophisticated viewpoint on quantitative evaluation techniques is provided in this section. HPC security considerations are discussed, highlighting container security [10]. Comprehending the subtleties of containerized environment security while considering high-performance computing is necessary.

An overview of the difficulties found in numerous areas of high-performance computing is provided, including architectural challenges, security issues, and scaling problems [9][11]. It is essential to acknowledge these barriers to guide future developments. Discussions of suggested innovations and solutions tackling HPC problems provide insights into possible directions for development[4]. This section adds to the discussion on how to direct the development of high-performance computing. The survey's principal findings are combined into a comprehensive summary that offers an outline of cluster computing, containerization, HPC architectures, performance assessment, and security issues [5][9][3]. Finishing remarks highlight essential areas for future research and technological developments while providing reviews on the state and revolution of HPC today [5][9][3].

Discussion and Findings

This section presents a comprehensive analysis of the survey results, providing a detailed examination of the key takeaways from the research. The papers included in the survey are assessed critically, emphasizing both their advantages and disadvantages. The survey also recognizes the limitations and gaps found. Along with outlining possible directions for future study.

3.1 Cloud Computing

The paradigm-shifting advancement in information technology known as cloud computing has completely changed how people access and use computer resources. This revolutionary method eliminates the need for enterprises to make significant infrastructure investments by offering on-demand internet access to a shared pool of computing resources. The key features of cloud computing are scalability, flexibility, and affordability, which allow customers to take advantage of computing power without being constrained by conventional on-premise infrastructure [8]. The core idea behind cloud computing is providing a range of services via the internet.

The terms Infrastructure as a Service (IaaS), Platform as a Service (PaaS), and Software as a Service (SaaS) have become more often used in cloud computing. SaaS offers software applications on a subscription basis, PaaS provides a platform that allows application development and deployment without worrying about the underlying infrastructure, and IaaS offers virtualized computing resources over the internet [7]. Even with cloud computing's benefits, there are still specific issues and things to consider. Organizations must consider possible vendor lock-in, regulatory compliance, security and privacy concerns, and more. Furthermore, depending on an organization's unique needs and limitations, selecting cloud service models and deployment patterns (public, private, hybrid, or multi-cloud) requires careful study [2].

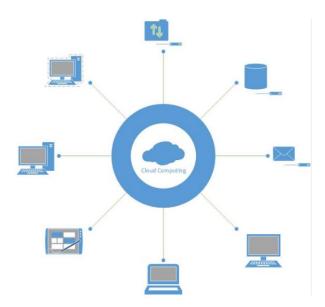


Figure 3.1: Cloud Computing [13]

3.1.1 Architectural Framework

- Researched the internal structure of cloud computing, with a focus on frontend and backend components. [8]
- Studied the combination of service-oriented architecture (SOA) and event-driven architecture (EDA) in cloud architecture. [7]
- Explored the impact of containers (Docker, Podman, Singularity, and Charliecloud) on HPC environments, considering startup times and I/O performance. [2]

3.1.2 Security Considerations

- Examined security tools in cloud computing, concentrating on backend security for resources, systems, files, and infrastructure. [8]
- Managed security aspects in cloud technologies, emphasizing measures for secure data processing and management. [7]
- Investigated the security significance of HPC containers, with a focus on the role of Singularity and Charliecloud in providing secure execution environments. [2]

3.1.3 Performance Evaluation

- Considered the performance of cloud computing vendors using High-Performance Linpack. [8]
- Analyzed high-performance parallel computing in cloud technologies, emphasizing scalability and intelligent monitoring. [7]
- Conducted empirical analysis on the performance of HPC containers (Docker, Podman, Singularity, Charliecloud) on the Lustre file system. [2]

3.2 Grid Computing

Grid computing is a cutting-edge computing paradigm that connects dispersed computer resources worldwide to establish a strong network for increased processing power and data storage. This technology is divided into two categories: computing grid, which focuses on processing power, and data grid, which addresses storage demands.

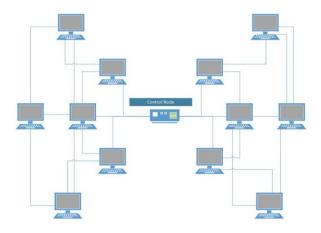


Figure 3.2: Grid Computing [13]

The comparative research tries to sort out the complexities of grid computing by bridging the common ground and distinctive variations among these scholarly publications, presenting an in-depth understanding of its growing surroundings and its implications for future improvements.

As described in Paper [5] grid computing emerges as an effective technique for aggregating globally distributed assets for better computing capabilities. Critical issues such as job scheduling, resource balancing, and network security are managed via the intra-grid scope. Grid computing benefits include unparalleled resource access, ideal resource usage, superior reliability through shared processing, and support for concurrent computing and scalability. Nevertheless, downsides

include its primarily batch-mode applicability and the requirement for parallel processing compatibility. Paper [9] broadens the scope of grid computing by defining its applications and distinguishing it from cloud, cluster, and distributed computing. It highlights the global significance of grid computing in many fields, ranging from radio astronomy to collaborative research in science. Specific examples in India, such as the Health Grid Project and GARUDA, demonstrate how grid computing may assist multiple sectors. The study emphasizes grid computing's dynamic nature, highlighting its critical role in tackling complex challenges across various industries.

3.2.1 Grid Architecture

- Global arrangement of interrelated circulated computer resources. [5]
- Horizontal scalability to cope with rising visitors and resource demands. [9]

3.2.2 Scheduling strategies

- Using complex algorithms to enhance task allocation. [5]
- Balances different factors such as duration, price, and accessibility to resources. [9]

3.2.3 Fault Tolerance

- Using resilient strategies to ensure uninterrupted operation. [5]
- Based on previous data and patterns, it predicts foreseeable failures [9]

3.2.4 Performance Evaluation

- comprehensive monitoring of efficiency, throughput, and resource use measures. [5]
- Based Conventional tests are used to monitor and compare system efficiency. [9]

3.3 Cluster Computing

Cluster computing is a group of tightly or loosely connected computers. The connected cluster acts as a single entity because the operations in these systems are executed together, thus creating the idea of a single system. The cluster of interconnected computers is linked through fast local area networks (LANs).

Cluster computing helps to solve computational problems. These interlinked systems, called nodes or cluster nodes, behave as a single system to perform tasks that require substantial computational power. They communicate and work with each other to coordinate and distribute the workload.

This discussion concentrates on cluster computing, with a comparative analysis of the contributions made by authors in papers [3], [6], [12] and [3]. Each paper sheds light on distinguishable elements of cluster computing, ranging from the role of containers to orchestration systems and performance evaluation.

3.3.1 Role of Containers/Clusters

- The paper discusses the crucial role of containers in large-scale clusters, emphasizing container orchestration systems, their taxonomy, challenges, and future directions[3].
- Containers can enhance resource isolation, portability, scalability, and efficiency [6].
- Employs Canopy Clustering, MPI parallelization, leveraging canopies to reduce complexity and optimize clustering efficiency[12].
- The paper highlights containers for efficiency, scalability, rapid deployment, and OS-level virtualization[10].

3.3.2 Cluster Techniques

- The paper explores cluster techniques, scheduling, rescheduling, and multitenancy in orchestration systems [3].
- Techniques involve Hadoop for distributed processing, signal clustering, AHP, and validation[6].
- The paper stresses Docker Swarm, and Kubernetes for efficient container deployment and management [12].
- The paper highlights DockerSwarm and Kubernetes for efficient deployment, workload management, and resource optimization[10].

3.3.3 Orchestration Systems

- Kubernetes, Swarm, Mesos, Aurora, Marathon, YARN, Borg, Omega, Apollo, and Fuxi, providing insights into their characteristics and classifications [3].
- Hadoop [6].

- Orchestration systems like Kubernetes or Docker Swarm are commonly associated with containerized applications[12].
- Docker Swarm, Kubernetes, Mesos, and Cloudify [10].

3.3.4 Fault Tolerance

- Highlights of frameworks such as Kubernetes are designed to address fault tolerance and availability challenges [3].
- It is achieved through data replication, data locality, job recovery mechanisms, task redundancy, and monitoring components like Job Tracker and Task Tracker[6].
- The paper needs to mention fault tolerance measures in the context of the parallelized Canopy Clustering using MPI, and specific details on fault tolerance need to be outlined[12].
- The paper does not explicitly discuss fault tolerance [10].

The paper [3] emphasizes the disparities between commercial and open-source container orchestration systems, underlining advanced features in commercial counterparts. It delves into scheduling flexibility, resource oversubscription challenges, and the absence of preemption support in most open-source systems. There is a suggestion for multitenancy support to be standardized while optimizing prospects and the need for rescheduling in long-running workloads. The significance of cloud-aware placement algorithms and challenges in acquiring more robust isolation mechanisms for multitenancy is highlighted. The findings underscore future directions, emphasizing the need for more sophisticated QoS management and fault-tolerant approaches. Meanwhile, paper [6] exhibits the proposed method's efficiency in large-scale data processing using Hadoop models in a cloud environment. Signal clustering and analytical hierarchy processes improve data organization. Comparative analysis shows outstanding performance compared to traditional methods. Application to smart city parameters highlights adaptability. The paper proposes a future extension with a fuzzy-based cluster computing algorithm—paper [12] efficiently parallel Canopy Clustering using MPI, delivering three optimized variants. Asynchronous Canopy MPI displays exceptional performance in execution time and scalability across datasets, showcasing its prospect for distributed parallel clustering. Finally, the paper [10] highlights Docker's pivotal role in virtualization, stressing efficiency and scalability. Despite the significance of container performance evaluation, there needs to be more focus on existing research. Widespread technologies like Docker, OpenVZ, and LXC are examined, with Docker reported for instantaneous administration and high performance. Challenges include security and resource management. Container clusters, orchestrated by tools like Kubernetes, improve scalability. The paper calls for further evaluation techniques typical to container models, identifying their influence on project design and deployment decisions.

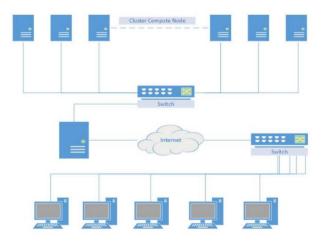


Figure 3.3: Cluster Computing [13]

3.4 Super Computing

Supercomputing is the epitome of computing prowess, meticulously crafted to manage intricate and demanding tasks beyond conventional computing systems' capacities. These cutting-edge devices are crucial in advancing scientific research, conducting simulations, analyzing data, and facilitating groundbreaking developments across diverse domains.

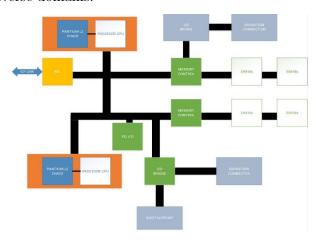


Figure 3.4: Super Computing [13]

By comparing Papers [11], [1], and [4], it's navigated into the realm of supercomputing. The analysis highlights the distinctive contributions put forth by

the authors in each paper, focusing on various facets of supercomputing. These encompass parallel processing, assessment techniques, and approaches to benchmarking.

The research paper [11] conducted an extensive literature review to pinpoint critical factors that improve supercomputing training in higher education. It highlights the significance of course organization, curriculum adaptations, problem-solving training, and teacher qualifications in enhancing learners' skills. Moreover, the paper delves into addressing limitations in supercomputing training and proposes practical solutions to overcome these challenges. By synthesizing pertinent literature, the study provides valuable insights into how high-performance computers positively impact educators and researchers in the higher education landscape. The research paper [1] employs a survey-based approach to categorize different forms of high-performance computing, including cluster computing, supercomputing, grid computing, and cloud computing. Notably, Cloud Computing emerges as one of the most widely adopted forms due to its cost efficiency and secure development. The paper offers practical insights into the pros and cons of each form, providing a nuanced understanding of their applications and implications. The survey-based evaluation contributes to a comprehensive understanding of the diverse landscape of high-performance computing technologies. The research paper [4] introduces a groundbreaking semi-supervised approach for anomaly detection in high-performance computing systems using autoencoders. The study evaluates the proposed approach to the real-life HPC system, D.A.V.I.D.E., demonstrating a notable 12 percent increase in accuracy compared to existing techniques. Two distinct autoencoder models are discussed: one for each supercomputer node and another with a single, generalized model, shedding light on the trade-off between accuracy and ease of deployment. The paper concludes with practical guidelines for implementing the proposed approach, emphasizing its efficacy in real-world scenarios where automated anomaly detection is crucial for ensuring the reliability and availability of HPC systems.

3.4.1 Parallel Processing

- Investigates the use of supercomputers in higher education, with an emphasis on training enhancements and restrictions fixes. [11]
- Examines several forms of high-performance computing, highlighting the cost-effectiveness and safe growth of cloud computing. [1]
- Semi-supervised autoencoder strategy that provides useful implementation guidance and outperforms existing methods for anomaly identification in HPC systems. [4]

3.4.2 Evaluation Methods

- Study conducts a comprehensive review of 34 papers to evaluate supercomputing training parameters. [11]
- Compares autoencoder models on an HPC cluster to evaluate its methodology, attaining a 12 percent improvement in detection accuracy. [1]
- provides implementation guidelines and compares a semi-supervised autoencoder algorithm with alternatives on a real HPC system. [4]

3.4.3 Benchmarks and Simulation

- There is lacks of information on simulations or benchmarks. [11]
- Evaluates its HPC methodology by means of simulations, contrasting the performance of the model on computing nodes. [1]
- Emphasizes comparisons over explicit benchmarks or simulations by using real HPC system data for useful evaluation. [4]

3.5 Benefits of using HPC Systems

High-Performance Computing (HPC) is becoming widely used in many different industries because of its many benefits. The following are the benefits of using HPCs.

- Parallel Processing: Several tasks can be carried out concurrently thanks to HPC systems' ability to support parallel processing.
- Greater computing Power: HPC systems produce higher computing power by utilizing cutting-edge hardware components.
- Scalability: By adding more processing units or nodes, HPC designs are made to grow easily.
- Scientific Research and Discovery: Through the facilitation of modeling, data analysis, and simulations, HPC speeds up scientific research.
- Applications for Weather Forecasting and Climate Modeling: HPC is essential to these fields of study.
- Optimization of Industrial Processes: Manufacturing and engineering processes are optimized in industries with the use of HPC.

- **Big Data Analytics:** Real-time management and analysis of massive data quantities requires the use of HPC systems.
- Financial Modeling and Simulation: The financial industry makes substantial use of HPC to model and simulate intricate market movements.
- Drug Discovery and Healthcare: By using large-scale simulations and data analysis, HPC speeds up medical research and drug discovery.
- National Security and Defense: Defense applications including threat analysis, simulations, and cryptography depend heavily on HPC.

3.6 Limitations and gaps in HPC systems

Understanding the difficulties associated with High-Performance Computing (HPC) is necessary for a thorough analysis. These describe the various kinds of gaps and difficulties that are common in the field of high-performance computing.

- **High Cost**: Purchasing, maintaining, and running HPC systems need a substantial cash outlay.
- **Programming Complexity**: Creating programs for parallel processing can be difficult and calls for certain programming knowledge.
- Energy Consumption: HPC systems require a significant quantity of electricity, particularly large-scale ones.
- Scalability Challenges: Although high-performance computing (HPC) systems are built to scale, some applications may find it difficult to achieve ideal scalability.
- Limited Software Compatibility: Some current software might need major alterations or not be well-suited for parallel processing.
- Data Transfer Bottlenecks: In HPC systems, bottlenecks can occur when huge amounts of data are transferred between processors or nodes.
- Hardware Reliability: The hardware dependability of an HPC system is essential to its continuous operation.
- Limited Applicability for Some Tasks: Not every kind of computational activity is a good fit for high performance computing (HPC).
- Security Concerns: It can be difficult to maintain security in extensive HPC systems.
- Customization and Maintenance: It takes constant work to adapt HPC systems to particular requirements and preserve peak performance.

3.7 Future scope

HPC systems hold immense potential to bring about revolutionary transformations in processing large amounts of data, offering numerous benefits such as improved speed and efficiency in performance. Consequently, HPC is witnessing several promising research directions actively pursued. Some of these directions include:

- HPC Hybrid Cloud Solutions: After examining the best practices for combining on-premises and cloud-based HPC resources and analyzing the cost and performance of infrastructure of hybrid cloud cost and performance implications for diverse scientific and computational applications, best workload migration and data management practices in hybrid settings can be a possible future scope.
- 2. Integration of AI and Machine Learning:Sophisticated AI and machine learning approaches to improve the performance of HPC applications can be a possible future scope. The use of AI in HPC systems for workload optimization, resource allocation, and predictive maintenance can be further researched.
- 3. Quantum Computing and High-Performance Computing: The influence of quantum computing on HPC methods and applications can be further researched. For specific HPC problems and investigation of hybrid techniques that mix conventional and quantum computing can be done. Analyzing the scalability and applicability of quantum computing to complicated scientific and computational challenges emerges as a promising research topic.
- 4. **High-Efficiency HPC**: To lessen environmental effects, investigate energy-efficient designs and strategies for HPC systems. Create tools and frameworks for monitoring and optimizing energy usage in high-performance computing clusters. The tradeoffs with performance and energy efficiency in various HPC situations can be found.

Conclusion

The current status of HPC research, which encompasses multiple models such as cloud computing, grid computing, cluster computing, and supercomputing, signifies a dynamic world with numerous problems and possibilities. As technology advances, the future of HPC research will focus on many critical areas. Hybrid cloud solutions that incorporate on-premises and cloud-based resources are expected to grow in popularity. AI and machine learning integration will be important in optimizing HPC operations, boosting performance, and enabling more competent resource management. The confluence of quantum computing with high-performance computing (HPC) provides an intriguing opportunity to research fresh methodologies and strategies for dealing with complicated problems. Security and compliance will continue to be significant concerns, especially in cloud-based HPC, emphasizing the need for solid protocols and encryption solutions. Energy efficiency will continue to be crucial, with researchers attempting to lessen the adverse environmental effects of HPC systems. Containerization and orchestration technologies will advance to decrease performance overhead, and federated HPC resources will enable global collaborative research. Education and training programs will be vital in educating the workforce for the rapidly increasing HPC landscape, ensuring that practitioners have the necessary skills to realize the potential of advanced computing technologies fully.

Bibliography

- [1] P. Z. Xin Liu, Yong Liu, "Quantum simulator with high-performance on a new sunway supercomputer," *ACM*, p. 13, 2021.
- [2] R. I. S. K. A. R. B. Subil Abraham, Arnab K. Paul, "On the use of containers in high performance computing environments," p. 10, 2020.
- [3] R. B. Maria A. Rodriguez, "Container-based cluster orchestration systems: A taxonomy and future directions," *Wiley*, p. 22, 2019.
- [4] M. L. M. M. L. B. Andrea Borghesi, Andrea Bartolini, "A semisupervised autoencoder-based approach for anomaly detection in high performance computing systems," p. 11, 2019.
- [5] U. Angadi and A. G. Karegowda, "A survey on grid computing and its applications." p. 5, 2018. [Online]. Available: https://api.semanticscholar. org/CorpusID:88495402
- [6] Y. Z. S. Z. Wenzhun Huang, Haoxiang Wang, "A novel cluster computing technique based on signal clustering and analytic hierarchy model using hadoop," p. 8, 2019.
- [7] P. B. A. K. K. N. G. S. R. Kannadasan, N. Prabakaran, "High performance parallel computing with cloud technologies," p. 7, 2018.
- [8] M. Mohammadi and T. Bazhirov, "Comparative benchmarking of cloud computing vendors with high-performance linpack," p. 5, 2018.
- [9] K. T. Sungkar Armu, "A review of grid computing." p. 6, 2020.
- [10] S. M. B. S. d. R. S. d. S. Naylor G. Bachiega, Paulo S. L. Souza, "Container-based performance evaluation: A survey and challenges," *IC2E*, p. 6, 2018.
- [11] J. M.-D. M. C. V. M. Álvaro Fernández, Camino Fernández, "Supercomputers to improve the performance in higher education: A review of the literature," p. 12, 2019.
- [12] T. H. Jacek Burys, Ahsan Javed Awan, "Large-scale clustering using mpi-based canopy," p. 8, 2018.
- [13] R. K. B. R. S. S. A. I. K. E. A. Priati Assiroj, H L H S Warnars, "The form of high-performance computing: A survey," *IOP Conference Series: Materials* Science and Engineering, p. 10, 2019.