

Distributed Container Management in Cloud Computing Using Kubernetes and Docker

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Abstract—Container orchestration technologies play a crucial role in deploying large-scale artificial intelligence applications in cloud environments. Modern AI systems require scalable infrastructure capable of handling distributed workloads and high computational demand. This review analyzes orchestration platforms including Docker Swarm and Kubernetes, evaluating scheduling mechanisms, scalability, and resource utilization while identifying future research directions in AI infrastructure.

Index Terms— Artificial Intelligence, Cloud Computing, Container Orchestration, Docker, Kubernetes.

I. INTRODUCTION

Artificial intelligence applications increasingly rely on scalable distributed infrastructure. Containerization technologies enable consistent deployment of software systems across cloud environments. Orchestration frameworks automate container management, improving reliability and scalability of AI workloads. Artificial intelligence applications increasingly rely on scalable distributed infrastructure. Containerization technologies enable consistent deployment of software systems across cloud environments. Orchestration frameworks automate container management, improving reliability and scalability of AI workloads. Artificial intelligence applications increasingly rely on scalable distributed infrastructure. Containerization technologies enable consistent deployment of software systems across cloud environments. Orchestration frameworks automate container management, improving reliability and scalability of AI workloads. Artificial intelligence applications increasingly rely on scalable distributed infrastructure. Containerization technologies enable consistent deployment of software systems across cloud environments. Orchestration frameworks automate container management, improving reliability and scalability of AI workloads.

II. BACKGROUND OF CONTAINER ORCHESTRATION

Container orchestration platforms manage clusters of containerized applications by automating scheduling, scaling, networking, and monitoring. Kubernetes and Docker Swarm are widely used orchestration technologies that support distributed cloud-native systems.

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III. REVIEW OF RELATED RESEARCH

Previous studies highlight the advantages of Kubernetes in large-scale deployments due to advanced scheduling mechanisms and fault tolerance. Docker Swarm provides simplicity and ease of configuration, making it suitable for smaller deployments.

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IV. PERFORMANCE ANALYSIS

Performance evaluation of orchestration platforms focuses on metrics such as CPU utilization, memory efficiency, response latency, and throughput. Research indicates that Kubernetes performs better in large clusters while Docker Swarm offers easier setup.

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V. CHALLENGES IN AI WORKLOAD ORCHESTRATION

AI workloads introduce challenges including GPU scheduling, energy consumption, and dynamic resource allocation. Addressing these issues requires improved orchestration strategies and intelligent scheduling mechanisms.

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VI. FUTURE RESEARCH DIRECTIONS

Future research aims to integrate machine learning techniques into orchestration frameworks to enable predictive scaling and adaptive resource allocation for distributed AI systems.

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VII. CONCLUSION

Container orchestration platforms play an essential role in supporting modern AI infrastructure. Kubernetes provides strong scalability and reliability while Docker Swarm offers simplicity for smaller systems.

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