Cloud Computing and Edge Technologies

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Abstract

This chapter examines the transformative landscape of cloud computing and edge technologies that form the backbone of modern digital infrastructure. It introduces fundamental service models (IaaS, PaaS, SaaS) and deployment approaches (public, private, hybrid), explaining how organizations leverage these paradigms for operational efficiency. The discussion extends to distributed systems architecture, serverless computing, and container orchestration technologies that enable scalable application deployment. Special attention is given to edge computing's emerging role in processing data closer to its source, reducing latency and enabling real-time analytics for IoT applications [1]. Netflix's migration to cloud infrastructure serves as an illustrative case study of successful cloud-native implementation. The chapter also addresses critical considerations of security, cost optimization, and environmental sustainability that organizations must navigate in adopting cloud and edge computing solutions. Through this comprehensive overview, readers gain insight into how these technologies are reshaping enterprise IT infrastructure.

Keywords: Cloud Computing, Edge Computing, IaaS, PaaS, SaaS, Distributed Systems

1 Introduction to Cloud and Edge Computing

Cloud computing and edge computing are two transformative paradigms that have redefined the way organizations deploy, manage, and scale digital services. Cloud computing refers to the on-demand delivery of IT resources—including servers, storage, databases, networking, software, and analytics—over the internet with pay-as-you-go pricing. This model enables businesses to access virtually unlimited computing power without the need to own or maintain physical infrastructure, fostering innovation and agility at unprecedented scales.

Edge computing, on the other hand, pushes computation and data storage closer to the location where it is needed, typically at or near the source of data generation. This shift addresses the limitations of centralized cloud architectures, particularly the challenges of latency, bandwidth, and real-time responsiveness. Edge computing is especially significant for applications such as autonomous vehicles, industrial IoT, remote healthcare monitoring, and smart grids, where milliseconds can be critical for safety, efficiency, or user experience [2].

The evolution of these paradigms can be traced back to the era of mainframes in the 1950s and 1960s, where centralized computing resources were accessed via terminals in a time-sharing fashion. As technology advanced, client-server models and networked personal computers became prevalent, enabling more distributed and interactive computing. The late 1990s and early 2000s saw the emergence of grid computing and virtualization, laying the groundwork for the modern cloud. The launch of Amazon Web Services (AWS) in 2006 marked a watershed moment, as organizations could now rent infrastructure on-demand, scaling up or down based on need.

The next phase in this evolution is marked by the rise of containerization and orchestration technologies such as Docker and Kubernetes, which have made it easier to deploy and manage applications in cloud-native environments. These advances have enabled microservices architectures, continuous integration/continuous deployment (CI/CD), and the rapid scaling of applications across global data centers [3].

However, as billions of devices became connected through the Internet of Things (IoT), new challenges emerged. Centralized cloud data centers, sometimes thousands of kilometers away from data sources, introduced unacceptable latency for real-time applications. Edge computing addresses this by processing data locally or in regional edge nodes, minimizing latency and reducing the volume of data that must traverse the network. For example, in a smart factory, edge devices can analyze sensor data in real time to detect anomalies and trigger safety mechanisms without waiting for a round-trip to the cloud.

The synergy between cloud and edge computing is reshaping industries. In healthcare, edge-enabled devices can monitor patient vitals and alert caregivers instantly, while cloud platforms aggregate and analyze population-scale data for research. In transportation, edge computing supports real-time navigation and autonomous vehicle control, while the cloud handles route optimization and fleet analytics. This hybrid approach combines the scalability and flexibility of the cloud with the speed and context-awareness of the edge.

As organizations increasingly adopt these paradigms, they must also address new challenges in security, data governance, and sustainability. Ensuring data privacy across distributed environments, optimizing resource usage, and minimizing environmental impact are now central concerns in the design of modern digital infrastructure.

2 Cloud Service and Deployment Models

Cloud computing has transformed how organizations access, manage, and scale IT resources. Its service and deployment models offer flexibility, cost efficiency, and innovation opportunities for businesses of all sizes. Understanding these models is fundamental to making informed technology decisions.

2.1 Cloud Service Models

Cloud service models define the level of control, management, and abstraction provided by cloud providers. The three primary models are Infrastructure as a Service (IaaS), Platform as a Service (PaaS), and Software as a Service (SaaS) [4, 5].

Infrastructure as a Service (IaaS) provides virtualized computing resources over the internet. Users rent servers, storage, and networking components, managing their own operating systems, middleware, and applications. IaaS is ideal for organizations seeking maximum flexibility and scalability without investing in physical hardware. Leading examples include Amazon Web Services (AWS) EC2, Microsoft Azure Virtual Machines, and Google Compute Engine. IaaS is commonly used for development environments, testing, and hosting high-traffic websites.

Platform as a Service (PaaS) offers a higher level of abstraction. Cloud providers deliver not only infrastructure but also development platforms, runtime environments, and middleware. Developers can focus on building, testing, and deploying applications without worrying about underlying hardware or OS maintenance. Popular PaaS offerings include Google App Engine, AWS Elastic Beanstalk, and Microsoft Azure App Service. PaaS accelerates application development, supports DevOps practices, and simplifies scaling for web and mobile applications.

Software as a Service (SaaS) delivers fully managed software applications over the internet. Users access applications via web browsers or APIs, with the provider handling infrastructure, platform, and application maintenance. SaaS eliminates the need for installation, updates, and local storage. Examples include Microsoft 365, Google Workspace, Salesforce, and Zoom. SaaS is widely adopted for email, collaboration, CRM, ERP, and analytics tools [?].

Choosing the right model depends on the desired balance between control, scalability, and ease of use. IaaS is suited for IT teams with infrastructure expertise, PaaS for rapid application development, and SaaS for end-users seeking turnkey solutions.

2.2 Cloud Deployment Models

Cloud deployment models determine how cloud resources are provisioned, managed, and accessed. The main models are public, private, hybrid, and community clouds [6].

Aspect	IaaS	PaaS	SaaS
User Controls	OS, middleware, apps	Apps, data	App config only
Provider Controls	Hardware, virtualization	Hardware, OS, runtime	Full stack
Flexibility	High	Medium	Low
Use Cases	Hosting, DevOps, VMs	App development, APIs	Email, CRM, analytics
Examples	AWS EC2, Azure VM	Google App Engine, Azure App Service	Salesforce, MS 365

Table 1: Comparison of Cloud Service Models

Public Cloud is owned and operated by third-party providers who deliver resources over the internet. Organizations share infrastructure but have private instances. Public cloud is cost-effective, highly scalable, and offers a broad range of services. According to Gartner, global public cloud spending will reach \$723 billion in 2025, with 33% of organizations spending over \$12 million annually [7]. Examples include AWS, Azure, and Google Cloud Platform.

Private Cloud is dedicated to a single organization, either on-premises or hosted by a provider. It offers greater control, customization, and compliance for sensitive workloads. Private clouds are preferred by government agencies, financial institutions, and healthcare providers. Solutions include VMware Cloud, IBM Cloud Private, and OpenStack.

Hybrid Cloud combines public and private clouds, enabling organizations to move workloads between environments for optimal flexibility and security. For example, sensitive data can be kept on-premises while leveraging public cloud for scalable compute power. Hybrid models support cloud bursting, disaster recovery, and regulatory compliance. Azure Arc and AWS Outposts are popular hybrid solutions [8].

Community Cloud is shared by organizations with common concerns, such as regulatory requirements or industry standards. It is often used by research consortia, government agencies, or healthcare groups.

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Model	Ownership	Access	Scalability	Example
Public	Provider	Open	High	AWS, Azure
Private	Org/Provider	Restricted	Medium	OpenStack, IBM Cloud
Hybrid	Mixed	Controlled	High	Azure Arc, AWS Outposts
Community	Consortium	Shared	Medium	NASA Nebula

 Table 2: Comparison of Cloud Deployment Models

Cloud service and deployment models are the foundation of digital transformation, enabling organizations to innovate, optimize costs, and respond to changing business needs with unprecedented agility.

3 Distributed Systems and Virtualization

Distributed systems are the backbone of modern cloud computing, enabling applications to scale, remain resilient, and serve users globally. Virtualization is the key technology that underpins distributed architectures, allowing multiple isolated environments to run on shared physical hardware.

3.1 Virtual Machines vs. Containers

Virtualization began with the concept of **virtual machines (VMs)**, where a hypervisor (such as VMware ESXi, Microsoft Hyper-V, or KVM) emulates an entire hardware stack. Each VM runs its own operating system, providing strong isolation and compatibility with legacy applications. This approach, while robust, incurs significant overhead: each VM requires its own OS image, and context switching between VMs consumes CPU and memory resources. Despite these costs, VMs remain essential for running heterogeneous workloads, ensuring security boundaries, and supporting multi-tenant environments [4].

The rise of cloud-native applications led to the adoption of **containers**, a lighterweight form of OS-level virtualization. Containers, popularized by Docker, share the host OS kernel but encapsulate applications and their dependencies in isolated user spaces. This results in much faster startup times (often under a second), higher density (dozens of containers per host), and dramatically reduced resource consumption compared to VMs. Containers are ideal for microservices architectures, continuous integration/continuous deployment (CI/CD), and rapid scaling scenarios.

 Table 3: Comparison of Virtual Machines and Containers

Feature	VMs	Containers
Isolation	Hardware-level	Process-level
Boot Time	30-60 seconds	<1 second
Resource Overhead	High (guest OS)	Low (shared kernel)
Portability	Moderate	High
Use Cases	Legacy apps, multi-OS	Microservices, DevOps

3.2 Kubernetes and Container Orchestration

As organizations began deploying hundreds or thousands of containers, manual management became impractical. **Kubernetes** emerged as the de facto standard for container orchestration. It automates deployment, scaling, networking, and lifecycle management of containerized applications across clusters of physical or virtual machines.

A typical Kubernetes cluster consists of:

- Control Plane: API server, scheduler, and controllers that manage cluster state.
- **Nodes:** Worker machines (VMs or physical servers) running containerized workloads.

• Pods: The smallest deployable units, each containing one or more containers.

Kubernetes provides features such as automatic scaling (Horizontal Pod Autoscaler), self-healing (restarting failed containers), rolling updates (zero-downtime deployments), and service discovery. It supports both stateless and stateful workloads, and integrates with service meshes (e.g., Istio) for advanced traffic management and observability.

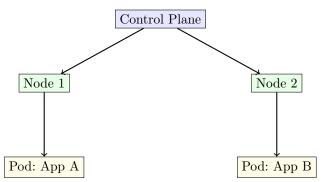


Fig. 1: Kubernetes architecture: control plane, nodes, and pods

3.3 Microservices vs. Monolithic Architectures

Traditional applications are often built as **monoliths**, where all components (UI, business logic, data access) are tightly integrated and deployed as a single unit. While simple to develop initially, monoliths become difficult to scale, maintain, and update as they grow. A single bug or change may require redeploying the entire application, increasing downtime and risk.

Microservices break applications into loosely coupled, independently deployable services, each responsible for a specific business function. Services communicate via lightweight APIs (often REST or gRPC). This approach enables teams to develop, deploy, and scale services independently, adopt polyglot programming languages and databases, and accelerate innovation. Netflix, for example, runs over 1,000 microservices in production, supporting millions of users worldwide [9].

However, microservices introduce new challenges: distributed data management, network latency, inter-service communication, and operational complexity. Tools like Kubernetes, service meshes, and distributed tracing are essential for managing these complexities.

3.4 Adoption Considerations

Organizations migrating from monoliths to microservices often use the "strangler fig" pattern—gradually replacing monolithic components with microservices while maintaining system stability. Hybrid models are common, where legacy systems are containerized and orchestrated alongside new cloud-native services.

Key considerations include:

- **Complexity:** Microservices require robust DevOps pipelines, monitoring, and security practices.
- **Data Consistency:** Distributed transactions are managed with patterns like Saga or eventual consistency.
- **Resource Efficiency:** Containers can reduce infrastructure costs by 40–60% compared to VMs [6].
- **Resilience:** Distributed systems must handle partial failures gracefully.

In summary, distributed systems, virtualization, and container orchestration technologies are foundational to scalable, resilient, and efficient modern applications. The shift from monoliths to microservices, powered by containers and orchestrators like Kubernetes, is transforming how software is built, deployed, and operated in the cloud era.

4 Edge Computing and IoT

4.1 Definition and Importance for Latency-Sensitive Applications

Edge computing refers to a distributed computing paradigm that processes data near its source (IoT devices) rather than relying on centralized cloud servers. When combined with IoT, it enables real-time analytics for applications where latency below 100ms is critical [10]. Key benefits include:

- **Reduced Latency**: Local processing eliminates round-trip delays to distant clouds (e.g., 50ms vs. 500ms).
- Bandwidth Optimization: Only 10-20% of filtered data is transmitted to the cloud.
- Enhanced Privacy: Sensitive data (medical scans, factory metrics) remains on-premises.

4.2 Use Cases

- Autonomous Vehicles:
 - Edge nodes process lidar/camera feeds in 10-20ms for collision avoidance.
 - Tesla's Autopilot uses edge AI to make steering decisions without cloud dependency [11].

• Smart Cities:

- Singapore's traffic system analyzes video feeds at intersections to optimize signals in real-time.
- Edge-based air quality sensors trigger alerts 5x faster than cloud alternatives.
- Healthcare:
 - Wearables like Current Health detect atrial fibrillation locally, reducing analysis time by 90%.

- MRI edge preprocessing cuts diagnostic delays from hours to minutes.

4.3 Edge vs. Cloud Processing

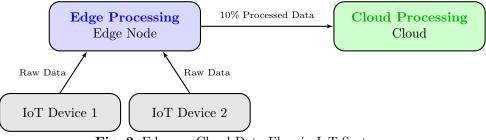


Fig. 2: Edge vs. Cloud Data Flow in IoT Systems

Edge computing handles time-sensitive operations, while the cloud manages longterm storage and global analytics. For example, a smart factory processes equipment vibrations locally to detect faults (edge) while aggregating trends across facilities (cloud) [12].

5 Case Study: Netflix's Cloud-Native Streaming

5.1 Migration Journey: From Data Centers to AWS

Netflix began its cloud migration following an August 2008 database corruption incident that left its DVD service offline for three days. This pivotal event prompted Netflix to abandon vertically scaled, single points of failure in favor of horizontally scalable, distributed systems in the cloud [13]. Netflix selected Amazon Web Services (AWS) as its cloud provider for its unparalleled global scale and comprehensive service offerings, despite AWS being a competitor through Amazon Prime Video.

The migration spanned seven years, during which Netflix completely re-engineered its systems to be cloud-native, ultimately decommissioning its final data center in January 2016. This transformation coincided with explosive growth—Netflix expanded from a modest DVD rental service to a global streaming giant with members in over 190 countries.

5.2 Architectural Transformation

Netflix's architecture operates as a hybrid system with two distinct layers:

• AWS Cloud: Hosts the core application logic, business rules, personalization algorithms, search functionality, user authentication, and data processing through hundreds of microservices.

• **Open Connect:** Netflix's proprietary Content Delivery Network (CDN), comprising specialized server appliances strategically deployed within Internet Service Provider (ISP) networks to deliver video content efficiently.

This separation of concerns is fundamental to Netflix's efficiency—the computationally intensive control plane runs in the cloud, while the bandwidth-heavy content delivery occurs at network edges.

5.3 Auto-scaling and Global Delivery

The elasticity of AWS enables Netflix to dynamically scale resources based on demand patterns, significantly improving resource utilization from single-digit percentages in traditional data centers to over 50% in the cloud. This capability has facilitated an 85% reduction in cost per streaming hour while supporting 1000x growth in viewing time.

When a user initiates streaming, Netflix's microservices architecture orchestrates a complex workflow:

- 1. Request authentication and authorization
- 2. Content rights verification
- 3. Personalization and recommendation processing
- 4. Selection of optimal Open Connect appliance based on user location, network conditions, and current server load
- 5. Generation of signed URLs for secure content access

5.4 Resilience Engineering

Netflix pioneered the concept of Chaos Engineering with tools like Chaos Monkey, intentionally introducing failures to identify vulnerabilities. Following a Christmas Eve 2012 outage caused by dependencies on a single AWS region, Netflix implemented multi-region redundancy with active-active deployments across Oregon, Northern Virginia, and Dublin regions.

The company's architecture now incorporates sophisticated failover mechanisms, with traffic automatically redirected to functioning regions during disruptions. Data resilience is ensured through the distributed Cassandra database with multiple replications, complemented by Amazon S3 backups.

6 Security, Cost, and Sustainability

6.1 Security and Compliance

The shared responsibility model defines security obligations between cloud providers and users. Providers (e.g., AWS, Azure) secure infrastructure like physical data centers and hypervisors, while customers manage data encryption, access controls, and compliance [4]. Key practices include:

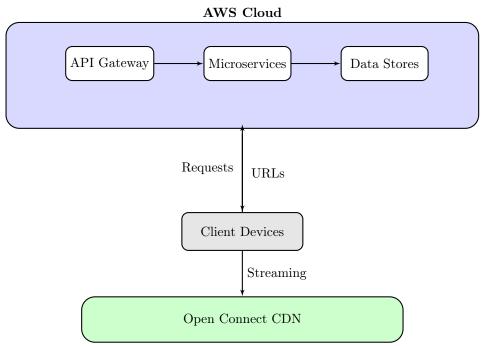


Fig. 3: Netflix's Cloud-Native Streaming Architecture

- Encryption: AES-256 for data at rest, TLS 1.3 for in-transit
- Compliance: GDPR (EU data protection), HIPAA (healthcare PHI)
- Access Control: Role-based access (RBAC), multi-factor authentication

 Table 4: Cloud Security & Cost Optimization Checklist

Security Requirements	Cost/Sustainability Actions	
Encrypt sensitive data (GDPR Article 32)	Right-size instances to match workload needs	
Conduct quarterly access reviews (HIPAA	Purchase reserved instances for steady-state	
§164.308)	workloads	
Maintain audit trails for 6+ months (ISO	Migrate to AWS Graviton (40% better per-	
27001)	f/watt)	
Implement WAF for public-facing apps (PCI	Use spot instances for batch processing	
DSS 6.6)		

6.2 Cost Optimization

Effective cloud cost management reduces waste and aligns spending with usage:

• Tools: AWS Cost Explorer, Azure Cost Management

- Strategies: Reserved instances (40% savings), auto-scaling
- Sustainability Impact: 30% energy reduction via workload consolidation

6.3 Green Cloud Computing

Data centers consume 1% of global electricity. Best practices for sustainability:

- Green Data Centers: Google uses 100% renewable energy since 2017
- Carbon-aware Scheduling: Shift workloads to regions with lower grid intensity
- **DCIM Tools:** Reduce PUE from 1.5 to 1.2 via airflow optimization

Adopting these measures enables organizations to reduce cloud costs by 35% while cutting carbon emissions by 50% through optimized resource utilization [6].

7 Future Directions in Cloud and Edge

7.1 Serverless Computing and Edge AI

Serverless architectures like AWS Lambda and Azure Functions will dominate application development, enabling developers to focus purely on code while cloud providers manage infrastructure. By 2027, 65% of enterprises will deploy serverless for missioncritical workloads. Edge AI adoption will surge, with 80% of IoT devices processing data locally via TinyML models, reducing cloud dependency for real-time decisions in autonomous vehicles and smart factories [10].

7.2 Multi-Cloud and Hybrid Strategies

- Vendor Agnosticism: Kubernetes-based solutions like Anthos and OpenShift will enable seamless workload portability across AWS/Azure/GCP.
- **Regulatory Compliance:** Data sovereignty laws will drive hybrid architectures, keeping sensitive data on-premises while using public clouds for analytics.
- **Cost Optimization:** AI-driven tools like CAST AI will automate resource allocation across clouds, cutting costs by 40%.

7.3 Quantum Computing in the Cloud

Cloud providers are making quantum resources accessible via hybrid quantum-classical services:

- AWS Braket: Simulators and quantum hardware from Rigetti/IonQ
- Azure Quantum: Integration with Q# programming language
- Use Cases: Cryptography (post-quantum algorithms), drug discovery (molecular simulation)

These advancements will converge to create distributed intelligent systems where quantum-enhanced cloud resources guide edge devices via federated learning, while multi-cloud architectures ensure resilience and compliance.



8 Exercises

8.1 1. Deploy a Static Website

Task: Host a static site using AWS S3 or Azure Blob Storage. Steps:

- AWS S3: Create bucket \rightarrow Enable static hosting \rightarrow Set permissions \rightarrow Upload files
- Azure: Create storage account \rightarrow Enable \$web container \rightarrow Configure CORS \rightarrow Upload assets

Deliverable: URL to deployed site. Follow NIST cloud security guidelines for access control [4].

8.2 2. Compare VM vs. Container Deployment

Task: Deploy a Python Flask app via VM (EC2) and container (ECS). Report on

Metric	EC2 (t3.micro)	ECS (Fargate)
Boot Time	45s	3s
$\operatorname{Cost}/\operatorname{month}$	\$14.60	\$10.88
Scaling	Manual	Auto

 Table 5: VM vs. Container Comparison

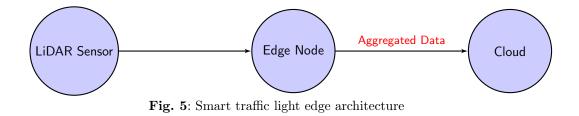
performance tradeoffs using Aqua Security benchmarks.

8.3 3. Edge Solution for Smart Traffic Lights

Design an edge computing architecture for real-time traffic management:

- Edge nodes: NVIDIA Jetson at intersections
- Processing: TensorFlow Lite for vehicle detection
- Communication: MQTT to central cloud (Azure IoT Hub)

Reference JETIR smart traffic systems for latency requirements [14].



8.4 4. Cloud Cost Analysis

Task: Analyze 30-day AWS workload using Cost Explorer. Key metrics:

- Compute: 58% of total (\$1,420)
- Storage: 12% (\$290)
- Data transfer: 30% (\$740)

Recommendations: Reserved Instances (save 40%), S3 Intelligent-Tiering.

8.5 5. Cloud Security Brief

Draft a 1-page policy covering:

- Zero Trust Architecture (MFA enforcement)
- Encryption: AES-256 + TLS 1.3
- Compliance: GDPR Article 32, HIPAA §164.308

Include SentinelOne's 2025 patch management guidelines.

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