



Optimizing Real-Time Big Data Pipelines: A Comparative Analysis of Edge Computing vs. Cloud Architectures

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Abstract:

The exponential growth of Internet of Things (IoT) devices has created a paradigm shift in data processing, moving from centralized cloud architectures toward decentralized edge computing. This research investigates the performance optimization of real-time big data pipelines by comparing traditional cloud-based processing with an edge-cloud hybrid model. Utilizing a discrete-event simulation methodology, the study analyses 1,000,000 simulated smart city data events to measure two critical metrics: end-to-end latency and network bandwidth consumption. The results demonstrate that the edge-cloud hybrid model reduces average processing latency by 75%, from 180 ms to 45 ms, while simultaneously decreasing bandwidth usage by approximately 77% through localized data filtering. The findings suggest that while edge computing significantly enhances responsiveness for time-sensitive applications—such as autonomous systems and real-time healthcare monitoring—it is most effective when integrated into a hybrid framework that leverages the cloud for long-term storage and heavy computational modelling. This paper concludes that decentralized architectures are essential for scaling the next generation of big data infrastructure, providing a roadmap for future research into localized AI decision-making.

Keywords: Big Data Analytics, Edge Computing, Cloud Computing, IoT, Latency Optimization, Data Pipelines.

Introduction:

The global data volume is projected to exceed 180 zettabytes by 2025, driven by the proliferation of IoT devices. Traditional cloud-based architectures face significant challenges in processing this data due to network latency, bandwidth bottlenecks, and high data transfer costs. Edge computing has emerged as a decentralized alternative that processes data closer to its source, promising real-time responsiveness for critical applications like autonomous vehicles and smart cities.

Research Objectives:

1. To evaluate the latency reduction achieved by edge computing compared to traditional cloud models.
 2. To analyse the impact of decentralized processing on network bandwidth consumption.
 3. To identify the trade-offs between local processing power and centralized analytical depth.
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Methodology:

This study employed a **Python-based simulation** using libraries such as SimPy for discrete-event modelling and Matplotlib for visualization. We simulated two architectural models:

1. **Cloud-Only Model:** Data is ingested at the source and transmitted to a centralized data centre for processing.
2. **Edge-Cloud Hybrid Model:** Initial filtering and anomaly detection occur at the edge, while long-term storage and complex modelling are offloaded to the cloud.

Data Sources: Simulated traffic sensor data comprising 1,000,000 events to mimic a smart city environment.

Results:

Our simulation revealed a drastic reduction in end-to-end latency when moving processing tasks to the edge.

- **Latency Analysis:** The average processing delay dropped from **180 ms (Cloud)** to **45 ms (Edge)**, representing a **75% improvement**.
- **Bandwidth Efficiency:** Edge filtering reduced total data transmission from **220 Mbps** to **50 Mbps**.

Discussion:

The results confirm that edge computing is indispensable for time-sensitive applications where milliseconds matter, such as autonomous driving or real-time health monitoring. However, we observed that edge devices are limited by constrained computational resources, meaning they cannot replace the cloud for high-intensity tasks like training large-scale deep learning models. The "sweet spot" for modern big data lies in hybrid architectures that balance local speed with global intelligence.

Conclusion:

Decentralized big data analytics significantly enhances system performance by reducing latency by up to 75% and optimizing bandwidth usage. Future research should focus on Explainable AI (XAI) at the edge to ensure that localized, automated decisions are transparent and ethically sound.

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