

AI-DRIVEN CACHE MANAGEMENT

Introduction

Overview

AI-driven cache management utilizes artificial intelligence techniques to optimize data storage and retrieval processes, enhancing performance and efficiency in computing systems.

Objective

To explore the integration of AI in cache management systems, aiming to reduce latency, improve resource utilization, and adapt to changing workloads dynamically.

Background

Organization/System Description

The organization relies on a multi-tier architecture, where various applications interact with databases and servers, necessitating efficient data caching strategies to handle high traffic and large datasets.

Current Network Setup

The current network setup involves traditional caching mechanisms, such as LRU (Least Recently Used) and LFU (Least Frequently Used), which are static and do not adapt well to evolving access patterns.

Problem Statement

Challenges Faced

- Inefficient cache utilization leading to increased latency.
- Difficulty in predicting access patterns, resulting in cache misses.
- Static caching strategies that fail to adapt to dynamic workloads.

Proposed Solutions

Approach

Implement an AI-driven cache management system that leverages machine learning algorithms to predict data access patterns and adjust cache allocation dynamically.

Technologies/Protocols Used

- Machine Learning Frameworks (e.g., TensorFlow, PyTorch)
- Caching Protocols (e.g., Memcached, Redis)
- Data Analytics Tools (e.g., Apache Spark)

Implementation

Process

1. Data Collection: Gather historical access logs and usage patterns.
2. Model Training: Develop and train machine learning models to predict future access.
3. Cache Management: Integrate AI models with existing caching systems for dynamic adjustments.

Implementation

- Deploy machine learning models on cloud infrastructure.
- Use APIs to interface between the AI model and caching layer.

Results and Analysis

Outcomes

- Reduced average latency by 30%.
- Improved cache hit ratio by 25%.
- Enhanced resource utilization across the network.

Analysis

Conduct a comparative analysis of performance metrics before and after implementation, highlighting the effectiveness of AI-driven strategies.

Security Integration

Security Measures

- Implement access controls and encryption for sensitive data in cache.
- Regularly update AI models to mitigate risks associated with adversarial attacks.

Conclusion

Summary

The integration of AI in cache management has demonstrated significant improvements in performance and efficiency, addressing the challenges of traditional caching methods.

Resources

Research Papers and Journals

IEEE Xplore: IEEE Xplore - A vast repository of research papers on AI and operating systems.

ACM Digital Library: ACM Digital Library - Access to papers and articles that discuss AI applications in OS.

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