

Smart Resource Allocation for Mobile Edge Computing: A Deep Reinforcement Learning Approach

Pavitra Basappa Gudimani

6880555

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Computer Networks Group
Universität Paderborn

Agenda

- Introduction
- Problem Definition
- Mobile Edge Computing (MEC) architecture
- Software Defined Network (SDN) enabled MEC architecture
- Deep Reinforcement Learning based Resource Allocation (DRLRA) Algorithm
- Performance Evaluation
- Conclusion
- Reference



Introduction

- Huge amount of data generated by communication devices cause congestion at cloud for data processing and cause delay in service time.
- Solution - Mobile Edge Computing (MEC) architecture with Deep Reinforcement Learning (DRL).

Problem Definition

Average service time minimization

- Edge network routing delay.
- Data processing delay.

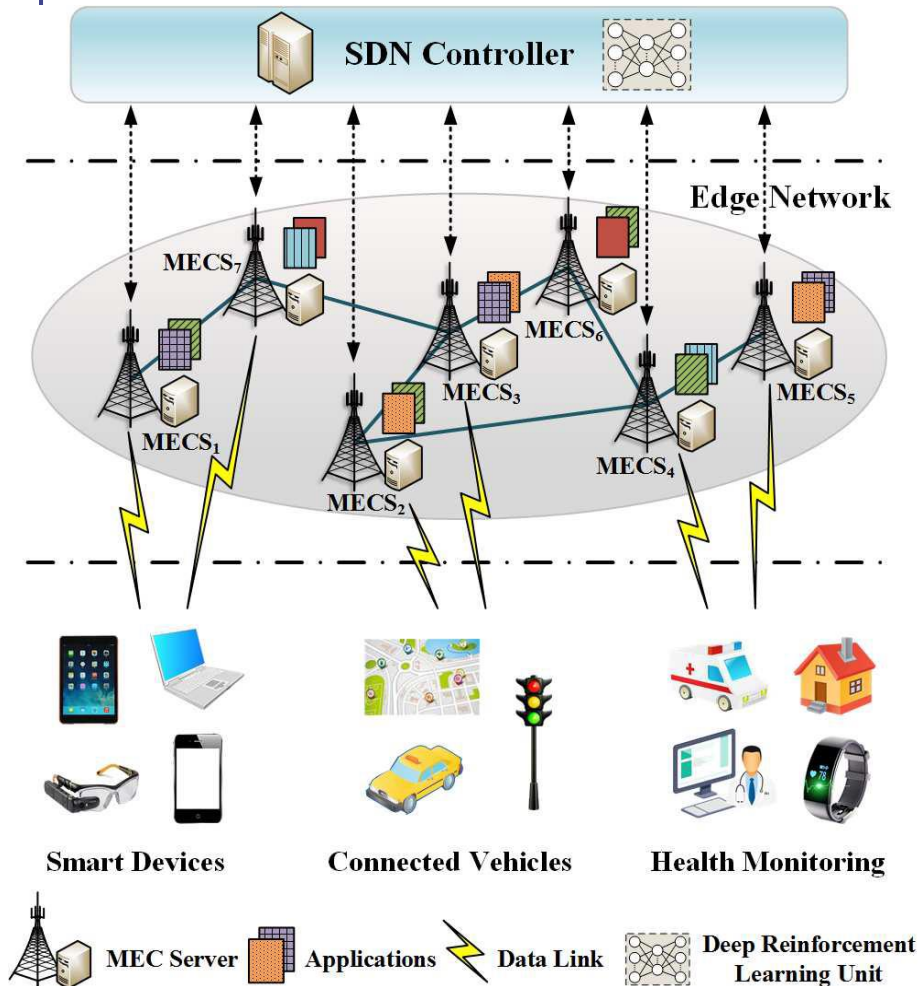
Resource allocation balancing

- Variance of network resource allocation.
- Variance of computing resource allocation.

Mobile Edge Computing (MEC) architecture

- MEC architecture reduces response delay by moving computing and storage resources near to mobile devices.
- Limitations of MEC
 - Deployment and maintenance is expensive.
 - Limited number of applications deployed.
 - MEC server goes down with burst requests.
- DRLRA algorithm to allocate resources adaptively under varying MEC environment.

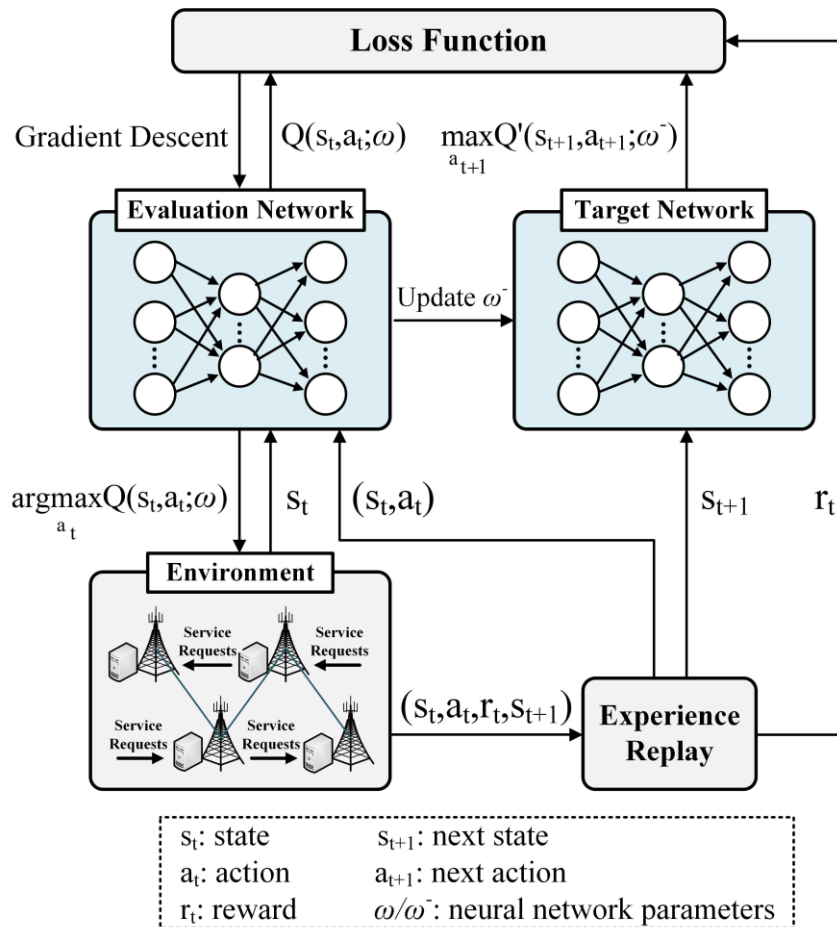
Software Defined Network (SDN) enabled MEC architecture



- Requests are routed to destined MEC server for processing.
- SDN technology maintains network infrastructure.
- DRLRA is deployed in SDN controller plane to offer intelligent routing decisions.

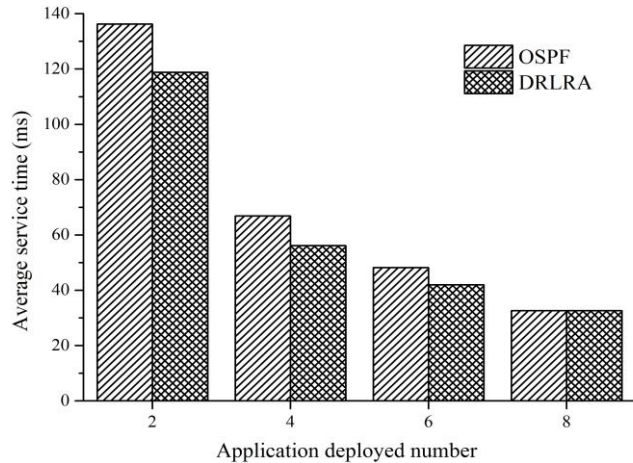
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Deep Reinforcement Learning Based Resource Allocation (DRLRA) Algorithm

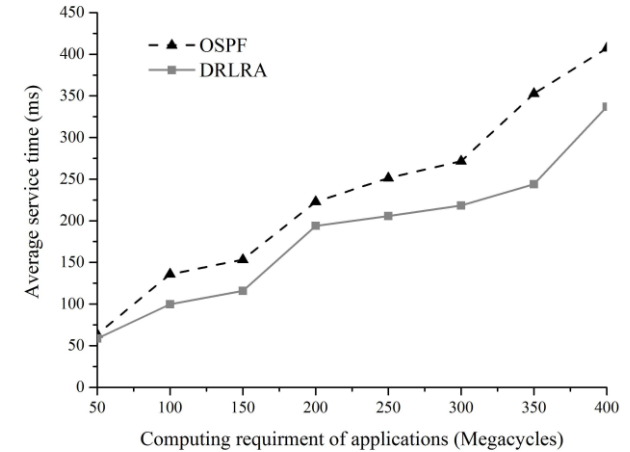


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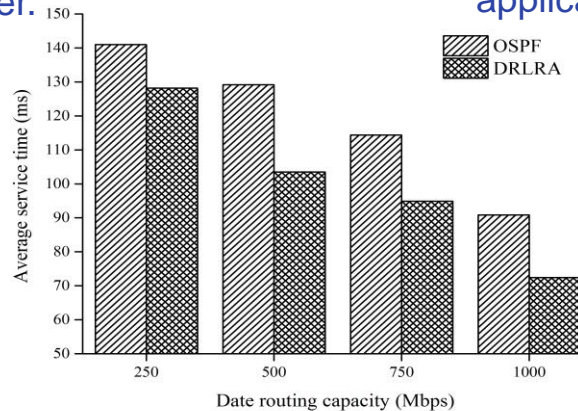
Performance Evaluation



Comparison of performance under different applications deployed number.



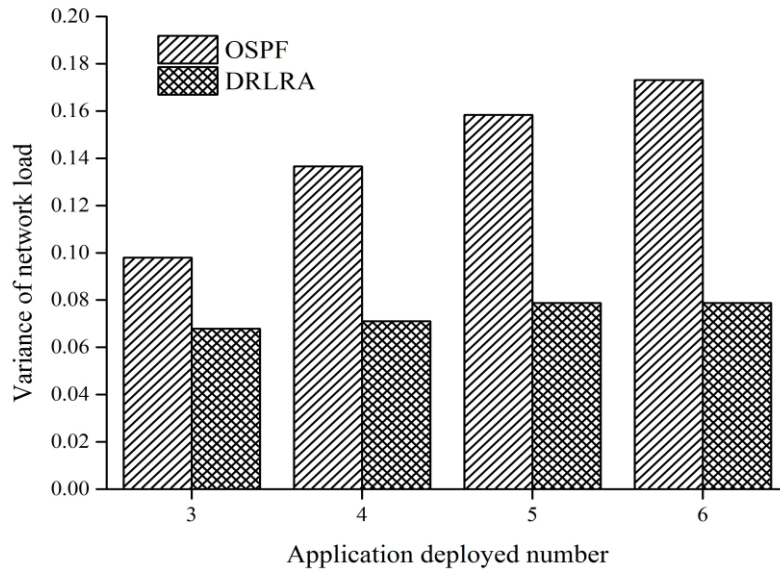
Impact of computing requirement of applications on average service time.



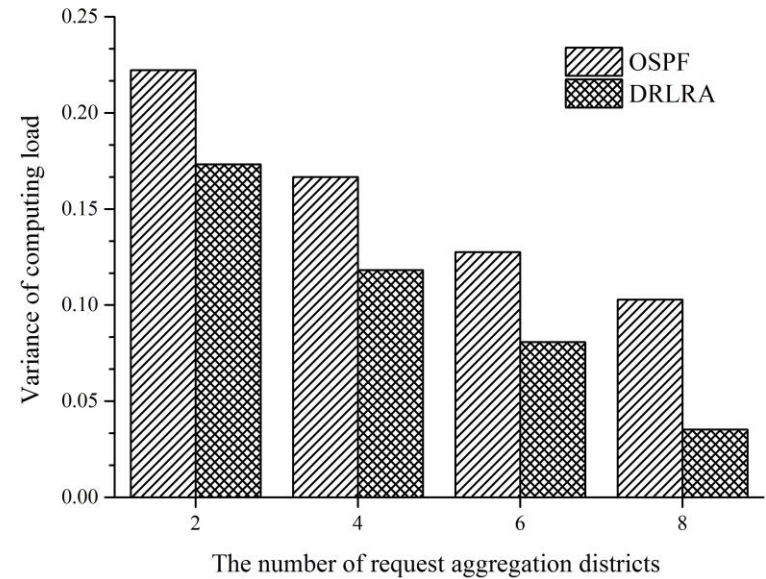
Performance comparison under different data routing capacities.

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Performance Evaluation



Comparison of variance of network load under different application deployed number.



Comparison of the variance of computing load under different request aggregation districts.

Source [1]

Conclusion

- The DRL's capacity of being adaptable to varying environment made it suitable for MEC server's burst environment.
- Reduced service time with DRLRA proved efficient than classical Open Shortest Path First (OSPF) algorithm.

Reference

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Thank You

