

A Mobility-aware Cross-edge Computation Offloading Framework for Partitionable Applications

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Outline

1 Introduction

- A Brief Introduction to Mobile Edge Computing (MEC)
- What is the Problem?

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2 Cross-edge Computation Offloading

- System Model
- Problem Formulation
- Proposed Framework and Algorithms
- Experimental Evaluation

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What is Mobile Edge Computing?

Mobile Edge Computing

Mobile Edge Computing (MEC) is a new computation paradigm:

- 1 deployed at **the network edge**
- 2 use **widespread** wireless access network (Small-cell Base Station)
- 3 provide **service and computing resource**



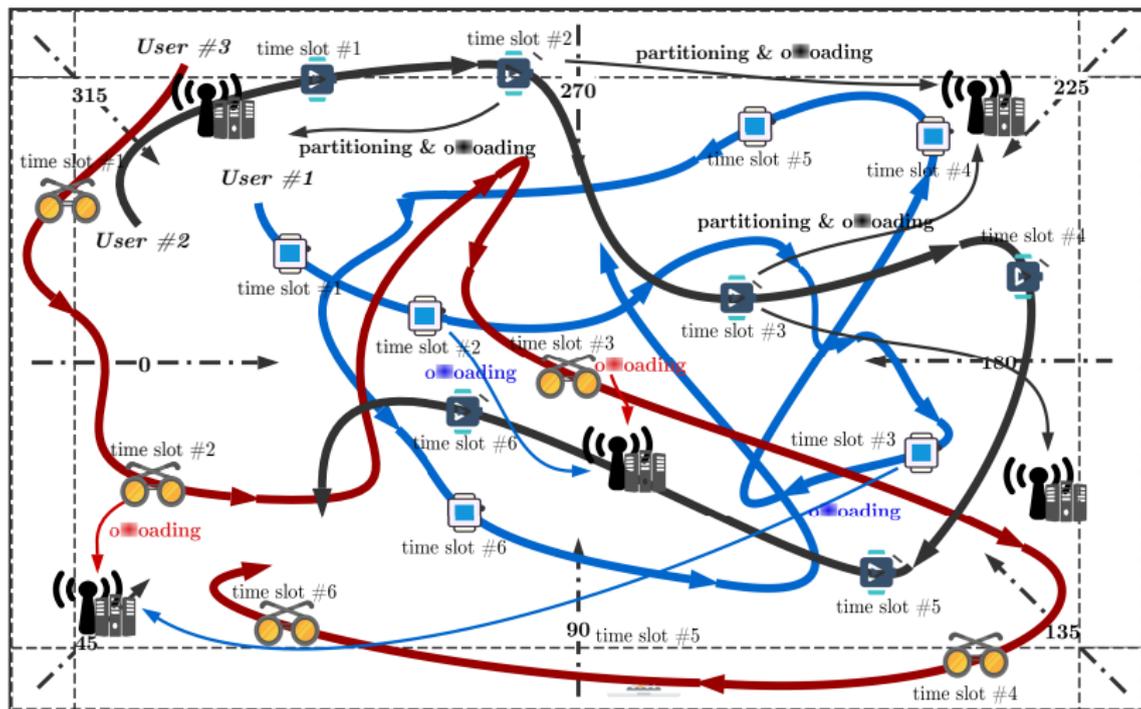
What's its properties?

Edge site

An edge site is a micro data center with applications deployed, attached to a small-cell base station (SBS).

- 1 Heterogeneous edge sites
- 2 User mobility
- 3 Edge site selection and user profile handover
- 4 Overlapped signal coverage of SBSs (Cross-edge Collaboration!)
- 5 Partitionable applications (data stream)
- 6 Insufficient battery energy of mobile devices
- 7

Motivation Scenario



Problem Definition

For **partitionable** applications,
how to make the **offloading** strategy with the
minimum overall cost achieved?

Composition of **overall cost**

- 1 execution delay
- 2 task dropping penalty
- 3 collaboration cost

Energy Harvesting (EH) technology is adopted.

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System Model

1 Local execution latency evaluation

- 1 execution latency: $\tau_i^{lc} \triangleq \eta_i^l / f_i$
- 2 energy consumption: $\epsilon_i^l \triangleq \kappa_i \cdot \eta_i^l f_i^2$

2 Offloading latency evaluation

- 1 transmission delay: $\tau_{i,j}^{tx}(t) \triangleq \frac{\mu_i^r}{\sum_{j \in \mathcal{M}_i(t)} I_{i,j}(t)} \cdot \frac{1}{R_{i,j}(t)}$
- 2 execution delay: $\tau_{i,j}^{rc}(t) \triangleq \frac{\eta_i^r}{f_j \cdot \sum_{j \in \mathcal{M}_i(t)} I_{i,j}(t)}$
- 3 collaboration cost: $\varphi \cdot \sum_{j \in \mathcal{M}_i(t)} I_{i,j}(t)$
- 4 constraint:

$$\tau_d \geq \max_{j \in \mathcal{M}_i(t)} \{ \tau_{i,j}^{tx}(t) + \tau_{i,j}^{rc}(t) \} + \tau_i^{lc} + \varphi \cdot \sum_{j \in \mathcal{M}_i(t)} I_{i,j}(t)$$

3 Battery energy level evaluation

- 1 evolution function:

$$\psi_i(t+1) = \psi_i(t) - \sum_{j \in \mathcal{M}_i(t)} \epsilon_{i,j}^{tx}(t) \cdot I_{i,j}(t) - \epsilon_i^l + \alpha_i(t)$$
- 2 constraint: $\epsilon_i^l + \sum_{j \in \mathcal{M}_i(t)} \epsilon_{i,j}^{tx}(t) I_{i,j}(t) \leq \psi_i(t)$

Problem Formulation

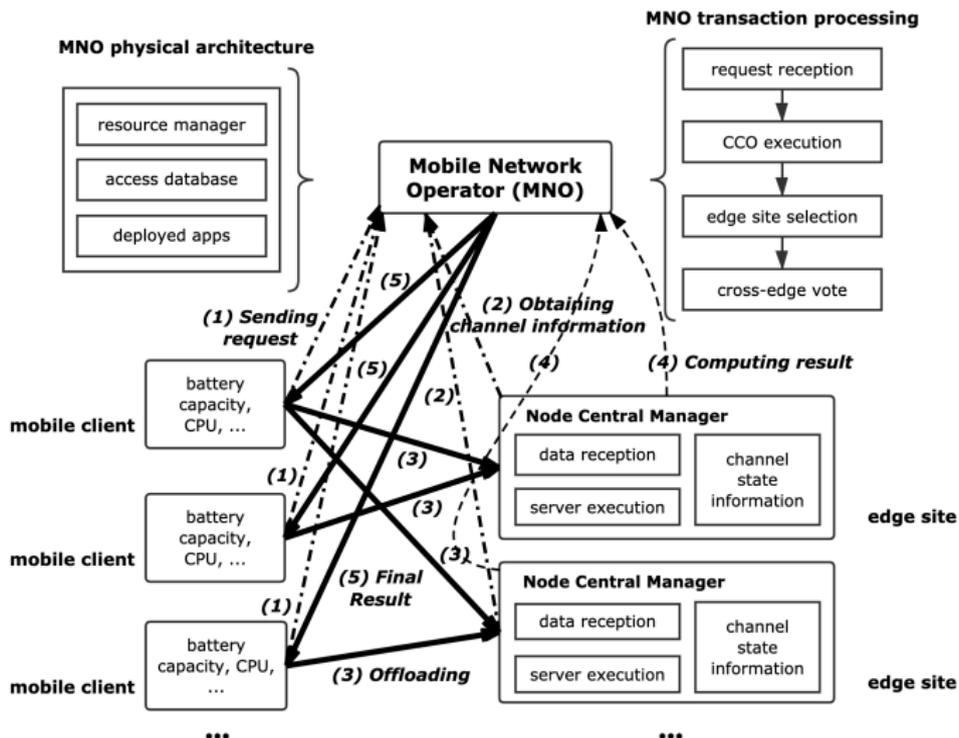
A non-convex optimization problem

$$\mathcal{P}_1 : \min_{\forall i, \mathbf{I}_i(t), \alpha_i(t)} \lim_{T \rightarrow \infty} \frac{1}{T} \sum_{t=0}^{T-1} \mathbb{E} \left[\sum_{i \in \mathcal{N}} \mathcal{C}(\mathbf{I}_i(t)) \right],$$

with several constraints.

$$\begin{aligned} \mathcal{C}(\mathbf{I}_i(t)) &\triangleq \max_{j \in \mathcal{M}_i(t) : I_{i,j'}(t)=1} \{ \tau_{i,j}^{tx}(t) + \tau_{i,j}^{rc}(t) \} \\ &+ \tau_i^{lc} + \varphi \cdot \sum_{j \in \mathcal{M}_i(t)} I_{i,j}(t) + \rho_i \cdot D_i(t) \end{aligned}$$

Proposed Framework



Proposed Algorithms

The CCO algorithm

- 1 Lyapunov optimization (drift-plus-penalty)

$$\mathcal{P}_2 : \min_{\forall i, \mathbf{I}_i(t), \alpha_i(t)} \Delta_V^{up}(\Theta(t)),$$

with several constraints.

$$\begin{aligned} \Delta_V^{up}(\Theta(t)) &\triangleq \sum_{i=1}^N \psi'_i(t) \left[\alpha_i(t) - \epsilon_i^l - \sum_{j=1}^M \epsilon_{i,j}^{tx}(t) I_{i,j}(t) \right] \\ &+ V \sum_{i=1}^N \mathcal{C}(\mathbf{I}_i(t)) + C \end{aligned}$$

Proposed Algorithms

The CCO algorithm

Algorithm 1 Cross-edge Computation Offloading (CCO)

- 1: At the beginning of the t th time slot, obtain i.i.d. random events $\mathbf{A}(t), \mathbf{E}^h(t) \triangleq [E_1^h(t), \dots, E_N^h(t)]$ and channel state information.
 - 2: $\forall i \in \mathcal{N}$, decide $\mathbf{I}_i^*(t), \alpha_i^*(t)$ by solving the deterministic problem \mathcal{P}_2 .
 - 3: $\forall i \in \mathcal{N}$, update the battery energy level $\psi_i(t)$.
 - 4: $t \leftarrow t + 1$.
-

- ① optimal energy harvesting: $\alpha_i^*(t)$
- ② optimal edge site selection: $\mathbf{I}_i^*(t)$

Benchmark Policies

- ① Random Selection (RS)
- ② Greedy Selection on Communication (GSC1)
- ③ Greedy Selection on Computation (GSC2)

Parameter settings

Parameter	Value	Parameter	Value
τ_d	2 ms	ϱ_i	2 ms
φ	0.02 ms	ρ_i	0.6
μ_i^l	100 bits	μ_i^r	3000 bits
f_i	1.5 GHz	f_j	32 GHz
κ_i	10^{-28}	ψ_i^{safe}	40 mJ
N_j^{max}	5	ω	$1.5 / \sum_{i \in \mathcal{N}_j(t)} I_{i,j}(t)$ GHz
ϖ_0	10^{-13} W	p_i^{tx}	1 W
g_0	10^{-4}	$E_{i,h}^{max}$	4.8×10^{-4} J

Optimality and stability

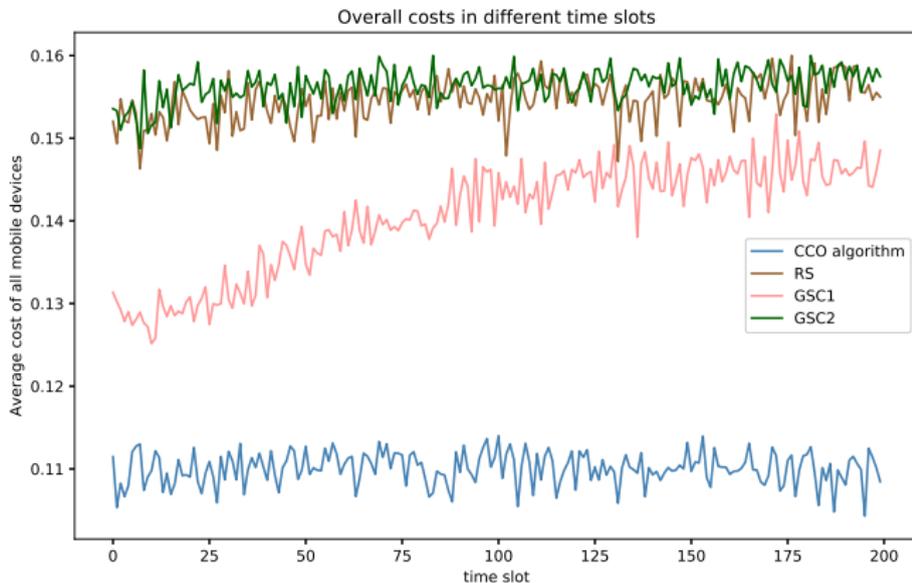


Figure: Average cost of mobile devices.

Optimality and stability

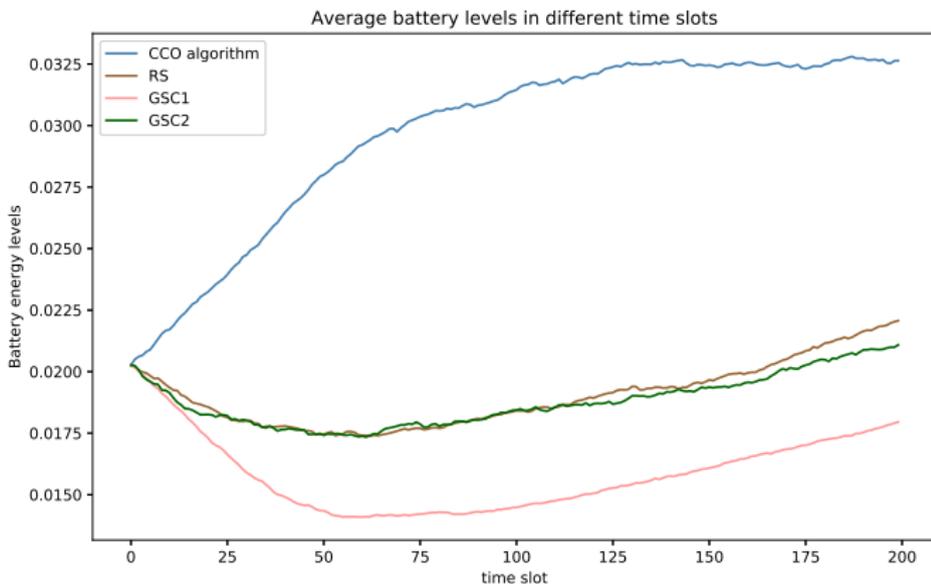


Figure: Average battery energy level of mobile devices.