

# Decentralized Scheduling for Offloading of Periodic Tasks in Mobile Edge Computing

Sladana Jošilo, György Dán

Department of Network and Systems Engineering  
School of Electrical Engineering and Computer Science  
KTH Royal Institute of Technology

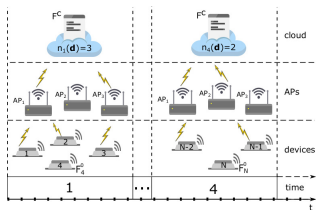
Zürich, May 16, 2018

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## Enabler for IoT Applications

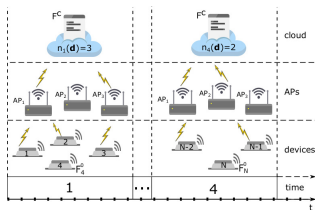
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  - computationally demanding processing
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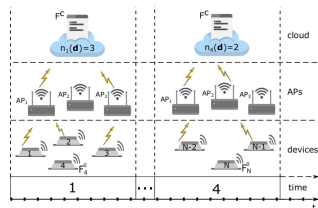
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- high bandwidth and computing resources close to the end users
- low latency
- real-time access to radio network information
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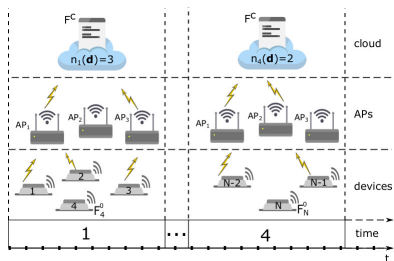
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## Important question

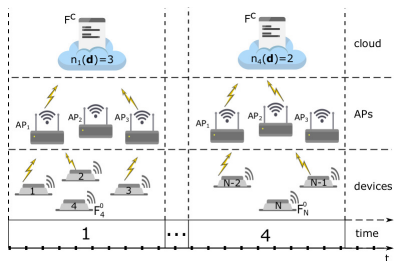
How to utilize mobile edge resources efficiently?

# Mobile Cloud Computing System with Periodic Tasks



- Cloud server
- Set of APs  $\mathcal{A} = \{1, 2, \dots, A\}$
- Set of WDs  $\mathcal{N} = \{1, 2, \dots, N\}$
- Set of time slots  $\mathcal{T} = \{1, 2, \dots, T\}$

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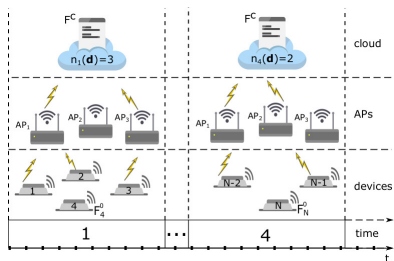


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## Computation Offloading

- Decision of WD  $i$   $d_i = \begin{cases} (t, 0), & \text{if performing the task locally in time slot } t \\ (t, a), & \text{if offloading the task using AP } a \text{ in time slot } t \end{cases}$
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- Number of WDs that offload via AP  $a$  in time slot  $t$  in a strategy profile  $\mathbf{d}$  is  $n_{(t,a)}(\mathbf{d})$
- Total number of WDs that offload in time slot  $t$  in a strategy profile  $\mathbf{d}$  is  $n_t(\mathbf{d})$



## Local Computing Cost

$$C_i^0 = \underbrace{\gamma_i^T T_i^0}_{\text{delay}} + \underbrace{\gamma_i^E E_i^0}_{\text{energy}}$$

## Cloud Offloading Cost

$$C_{i,(t,a)}^c(\mathbf{d}) = \underbrace{\gamma_i^T \left( \underbrace{T_{i,(t,a)}^{c,off}(n_{(t,a)}(\mathbf{d}))}_{\text{time to offload}} + \underbrace{T_{i,t}^{c,exe}(n_t(\mathbf{d}))}_{\text{time to execute}} \right)}_{\text{delay}} + \underbrace{\gamma_i^E E_{i,(t,a)}^c(n_{(t,a)}(\mathbf{d}))}_{\text{energy to offload}}$$

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## Coordinating MEC resources for offloading of periodic tasks

- Interactions between WDs modeled as a strategic game  $\Gamma = \langle \mathcal{N}, (\mathcal{D}_i)_i, (C_i)_i \rangle$

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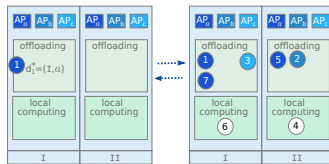
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- Is there a task allocation in which all selfish WDs are satisfied?
- Can it be computed using a decentralized algorithm?
- What is the complexity of the algorithm?
- How good are the system performance?

## Coordinated Myopic Alternating Best (MB)

- WDs enter the game one at a time and implement BR over all time slots

- Induction phase - starting from an empty system, WDs enter the game one at a time and play BR

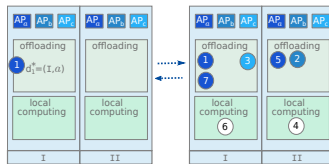


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  - WDs are not allowed to replace the previous deviators
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## Random Slot (RS) Equilibrium

- WDs enter the game one at a time and choose a time slot at random
- WDs implement BR within the chosen time slot

# Main Results

## Computability

- MB algorithm computes a NE of game  $\Gamma$  in  $O(N^2 \times T \times A)$  steps



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## Price of Anarchy (PoA) Bounds

- Upper bound on the PoA for the computation offloading game:

- $N \leq T$ :  $PoA = 1$

- $N > T$ :  $PoA \leq \frac{\sum_{i \in \mathcal{N}} C_i^0}{\sum_{i \in \mathcal{N}} \min\{C_i^0, \bar{C}_{i,1}^c, \dots, \bar{C}_{i,A}^c\}}$

- $\bar{C}_{i,a}^c$  is the offloading cost of WD  $i$  via AP  $a$  when she offloads alone

# Performance Evaluation - Simulations

## Evaluation scenario

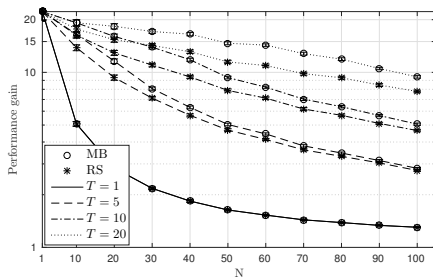
- $A = 4$  APs,  $F^c = 100$  Gcycles and  $F_i^0 \sim \mathcal{U}(0.5, 1)$  Gcycles
- Tasks: data size  $\sim \mathcal{U}(0.42, 2)$  Mb , complexity  $\sim \mathcal{U}(0.1, 0.8)$  Gcycles

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## System Cost Performance



- *Performance gain* defined w.r.t. a strategy profile in which all WDs perform computation locally
- *Performance gain* of the MB algorithm is higher than that of the RS algorithm for  $T > 1 \implies$  coordination is important

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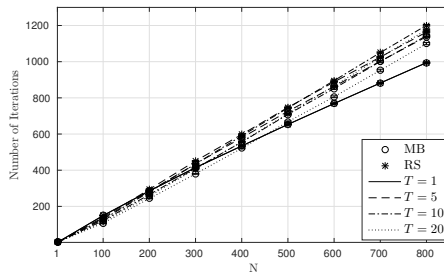
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## Computational complexity



- Number of iterations scales approximately linearly with the number of WDs for both algorithms

# Summary and Future Work

- Provided a game theoretical treatment of computation offloading for periodic tasks
- Proved the existence of equilibrium allocations
- Provided a polynomial time decentralized algorithm for computing an equilibrium

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- Interesting extensions
    - WDs with heterogeneous periodicities
    - resource allocation from the perspective of mobile cloud providers
    - D2D collaborative computation offloading



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