QOCO: A QoE-Oriented Computation Offloading Algorithm based on Deep Reinforcement Learning for Mobile Edge Computing

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Enhancing mobile device (MD) networks for a smarter world

Mobile Edge Computing (MEC)

- Proximity to Computational Resources
- Computation Task Dispatch for MDs
- Reduced Task Processing Delay
- Extended MD Battery Life
- Enhanced Quality of Experience (QoE)

Challenges in Computation Offloading

- Dynamic Network Conditions
- Heterogeneous Device Capabilities
- Latency vs. Energy Trade-offs
- QoE Prioritization

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Contribution

Research Focus: Computation Task Offloading Problem in MEC

- Task Processing Deadlines
- Energy Constraints

Key Contributions:

- Formulated the computation task offloading problem as a finite and discret Markov Decision Process (MDP) to maximize the expected long term QoE for each MDs.
- Proposed the QoE-oriented computation offloading (QOCO) algorithm based on Deep Reinforcement Learning (DRL) to empower each MD to make offloading decisions independently.
- Conducting comprehensive experiments to evaluate the performance of QOCO compared with several benchmark methods.

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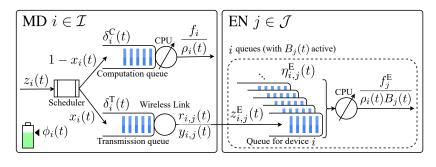
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System Model

We consider

- lacksquare the set of MD $\mathcal{I} = \{1, 2, ..., I\}$
- the set of EN $\mathcal{J} = \{1, 2, ..., J\}$
- lacksquare set of time slots $\mathcal{T} = \{1, 2, \dots, T\}$
- lacksquare each MD $i \in \mathcal{I}$ are connected to each EN $j \in \mathcal{J}$ with it's wireless interface.
- each time slot consider as a diuration of time.

System Model



$z_i(t)$	arrived task in MD	$\delta_i^C(t)$	Compution Queue in MD
$z_i^E(t)$	arrived offloaded task in EN	$\delta_i^C(t)$	Compution Queue in MD
$x_i(t)$	Offloading Decision	$\delta_i^C(t)$	Transsmition Queue in MD
$y_{i,j}(t)$	Offloading Target	$\eta_{i,j}^E(t)$	MD Compution Queue in EN
f_i	Computation capacity of MD	$B_j(t)$	Active Queue in EN
f_j^E	Computation capacity of EN	$\phi_i(t)$	MD battery level
$r_{i,j}(t)$	Transmission capacity of MD	$\rho_i(t)$	Required CPU cycles of task

Communication Model

 $lackbox{ } \delta_i^T(t)$: MD i Transmission Queue Waiting Time

$$\delta_i^T(t) = \left[\max_{t' \in \{0, 1, \dots, t-1\}} l_i^T(t') - t + 1 \right]^+ \tag{1}$$

 $lackbox{ } l_i^T(t)$: Task Transmited/Dropped Time Slot

$$l_i^T(t) = \min\left\{t + \delta_i^T(t) + \lceil D_i^T(t) \rceil - 1, t + \Delta_i(t) - 1\right\}$$
 (2)

■ $D_i^T(t)$: Task Transmission Time

$$D_i^T(t) = \sum_{\mathcal{J}} y_{i,j}(t) \frac{\lambda_i(t)}{r_{i,j}(t)\tau}$$
 (3)

 $lackbox{\bf E}_i^T(t)$: Task Transmission Energy Consumption

$$E_i^T(t) = D_i^T(t)p_i^T(t)\tau \tag{4}$$



Computation Model

Local Execution:

lacksquare $\delta_i^C(t)$: MD i Computation Queue Waiting Time

$$\delta_i^C(t) = \left[\max_{t' \in \{0, 1, \dots, t-1\}} l_i^C(t') - t + 1 \right]^+ \tag{5}$$

 $lackbox{l}_i^C(t)$: Task Executed/Dropped Time Slot

$$l_i^C(t) = \min\left\{t + \delta_i^C(t) + \lceil D_i^C(t) \rceil - 1, t + \Delta_i(t) - 1\right\}$$
 (6)

■ $D_i^C(t)$: Task Execution Time

$$D_i^C(t) = \frac{\lambda_i(t)}{f_i \tau / \rho_i(t)} \tag{7}$$

■ $E_i^L(t)$: Task Execution Energy Consumption

$$E_i^L(t) = D_i^C(t)p_i^C\tau \tag{8}$$

Computation Model

Edge Execution:

 \bullet $\eta_{i,j}^E(t)$: MD i Queue Backlog at EN j

$$\eta_{i,j}^{E}(t) = \left[\eta_{i,j}^{E}(t-1) + \lambda_{i,j}^{E}(t) - \frac{f_{j}^{E}}{\rho_{i}(t)B_{j}(t)} - \omega_{i,j}(t) \right]^{+}$$
 (9)

 $lackbox{ }\hat{l}_{i,j}^E(t)$: Task Execution Start Time Slot at EN j

$$\hat{l}_{i,j}^{E}(t) = \max\{t, \max_{t' \in \{0,1,\dots,t-1\}} l_{i,j}^{E}(t') + 1\}$$
 (10)

$$\sum_{t'=\hat{l}_{i,j}^{E}(t)}^{l_{i,j}^{E}(t)} \frac{f_{j}^{E}}{\rho_{i}(t)B_{j}(t')} \ge \lambda_{i,j}^{E}(t) > \sum_{t'=\hat{l}_{i,j}^{E}(t)}^{l_{i,j}^{E}(t)-1} \frac{f_{j}^{E}}{\rho_{i}(t)B_{j}(t')}$$
(11)

Computation Model

Edge Execution:

lacksquare $D_{i,j}^E(t)$: Task Execution Time at EN j

$$D_{i,j}^{E}(t) = \frac{\lambda_{i,j}^{E}(t)\rho_{i}(t)}{f_{j}^{E}\tau/B_{j}(t)}$$
(12)

 $lackbox{\blacksquare} E_{i,j}^E(t)$: Task Execution Energy Consumption at EN j

$$E_{i,j}^{E}(t) = \frac{D_{i,j}^{E}(t)p_{j}^{E}\tau}{B_{j}(t)}$$
 (13)

■ $E_i^I(t)$: MD i Standby Energy Consumption

$$E_i^I(t) = D_{i,j}^E(t)p_i^I \tau \tag{14}$$

$$E_i^{\mathsf{O}}(t) = E_i^{\mathsf{T}}(t) + \sum_{\mathcal{I}} E_{i,j}^{\mathsf{E}}(t) + E_i^{\mathsf{I}}(t).$$
 (15)

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Markov Decision Process

State Space:

$$s_{i}(t) = \left(\lambda_{i}(t), \delta_{i}^{C}(t), \delta_{i}^{T}(t), \boldsymbol{\eta}_{i}^{E}(t-1)^{\circ}, \phi_{i}(t), \mathcal{H}(t)^{\circ}\right)$$

$$^{\circ}\boldsymbol{\eta}_{i}^{E}(t-1) = (\boldsymbol{\eta}_{i,j}^{E}(t-1), j \in \mathcal{J})$$

$$^{\diamond}h_{i,j}(t) = B_{j}(t-T^{s}+i-1) \longrightarrow T^{s} \times J$$

$$\longrightarrow \mathcal{S} = \Lambda \times T^{2} \times \mathcal{U} \times 3 \times I^{T^{s} \times J}$$

$$(16)$$

Action Space:

$$\boldsymbol{a}_i(t) = (x_i(t), \boldsymbol{y}_i(t)^*) \tag{17}$$

$$\mathbf{y}_i(t) = (y_{i,j}(t), j \in \mathcal{J})$$

Markov Decision Process

QoE Function:

[Delay]

$$\mathcal{D}_i(\boldsymbol{s}_i(t),\boldsymbol{a}_i(t)) = (1-x_i(t)) \Big(l_i^{\mathsf{C}}(t)-t+1\Big) +$$

$$x_{i}(t) \left(\sum_{\mathcal{J}} \sum_{t'=t}^{T} \mathbb{1} \left(z_{i,j}^{\mathsf{E}}(t') = z_{i}(t) \right) l_{i,j}^{\mathsf{E}}(t') - t + 1 \right)$$
 (18)

[Energy]

$$\mathcal{E}_i(\boldsymbol{s}_i(t), \boldsymbol{a}_i(t)) = (1 - x_i(t))E_i^{\mathsf{L}}(t) +$$

$$x_i(t) \left(\sum_{\mathcal{J}} \sum_{t'=t}^T \mathbb{1} \left(z_{i,j}^{\mathsf{E}}(t') = z_i(t) \right) E_i^{\mathsf{O}}(t) \right) \tag{19}$$

Markov Decision Process

[Cost]

$$C_i(s_i(t), a_i(t)) = \phi_i(t) \mathcal{D}_i(s_i(t), a_i(t)) + (1 - \phi_i(t)) \mathcal{E}_i(s_i(t), a_i(t))$$
(20)

[QoE]

$$\begin{aligned} \boldsymbol{q}_i(\boldsymbol{s}_i(t), \boldsymbol{a}_i(t)) &= \\ \begin{cases} \mathcal{R}^* - \mathcal{C}_i(\boldsymbol{s}_i(t), \boldsymbol{a}_i(t)) & \text{if task } z_i(t) \text{ processed,} \\ -\mathcal{E}_i(\boldsymbol{s}_i(t), \boldsymbol{a}_i(t)) & \text{if task } z_i(t) \text{ dropped,} \end{cases} \end{aligned} \tag{21}$$

 $*\mathcal{R}$: A Constant Reward for Completed Task

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