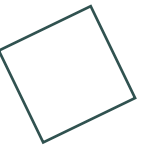


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Potential Game based Channel Allocation for Vehicular Edge Computing





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03 Potential Game Model

04 Algorithm Design

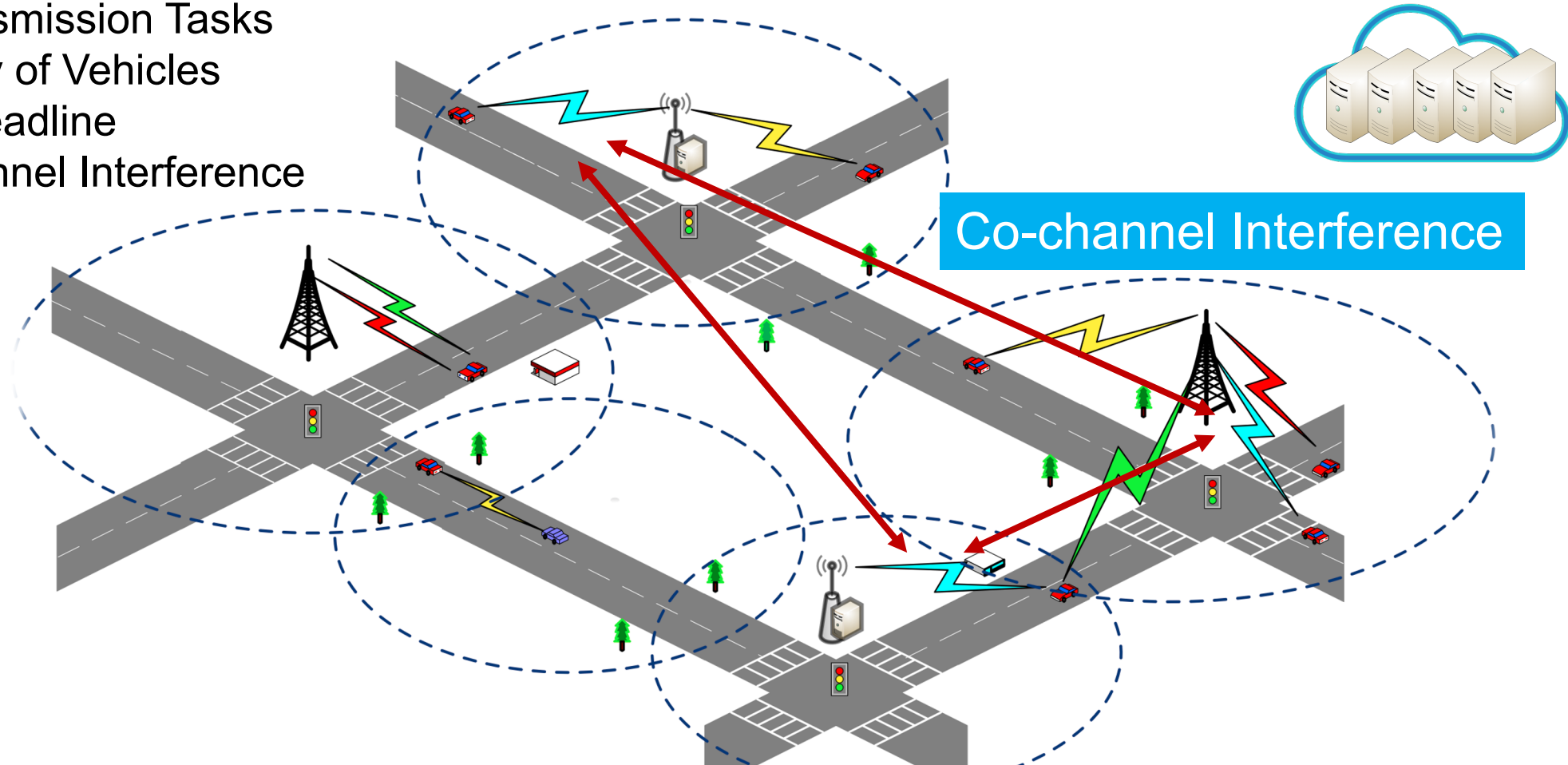
05 Performance Evaluation



► Introduction Vehicular Edge Computing Architecture

Data Transmission Tasks

- Mobility of Vehicles
- Task Deadline
- Co-channel Interference



Objective: Allocating sub-channels for different data transmission tasks and maximizing the number of completed tasks



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► Problem Formulation **Mobility-aware Communication Model**

- Signal to Interference and Noise Ratio, SINR

$$\mu_{nm}^k(t_i) = \frac{|h_{nm}|^2 \cdot \Psi \cdot d_{nm}(t_i)^{-\lambda} \cdot p_n}{\sum_{x=1, x \neq n}^{|E(t_i)|} \sum_{y=1}^{|B(t_i)|} |h_{xm}|^2 \Psi d_{xm}(t_i)^{-\lambda} p_x \mathbb{I}_{xy}^k(t_i)} + N_0$$

Signal

Co-channel Interference

- Transmission Rate

$$\delta_{nm}^k(t_i) = \omega_n \log_2 \left(1 + \mu_{nm}^k(t_i) \right)$$

- Transmission Data Size

$$z_{nm}^k = \int_{t_i}^{\min(t_{mi}^e, t_i + t_{mi}^c)} \delta_{nm}^k(t) dt$$

► Problem Formulation Channel Allocation Problem

■ Allocation Strategy

$$A(t_i) \in \{0,1\}^{|E(t_i)| \times |B(t_i)| \times |Q|} \quad a_{nm}^k(t_i) = 1 \quad \mathbb{A}(t_i) = \prod_{j=i}^{|T|} A(t_j) = A(t_i) \times A(t_{i+1}) \times \cdots \times A(t_{|T|})$$

■ Objective Function

$$\sum_{i=1}^{|T|} \max_{\mathbb{A}(t_i)} \sum_{m=1}^{|B(t_i)|} \mathbb{I}_{z_{mi} \leq \sum_{n=1}^{|E(t_i)|} \sum_{k=1}^{|Q_n(t_i)|} \int_{t_i}^{t_i + \beta \sum_{j=i}^{|T|} a_{nm}^k(t_j)} \delta_{nm}^k(t) dt} \quad \text{Task Completion Indicator}$$

$$s. t. \begin{cases} C1: a_{nm}^k(t_i) \in \{0,1\}, \forall e_n \in E(t_i), \forall b_m \in B(t_i), \forall q_k \in Q_n(t_i), \forall t_i \in T \\ C2: a_{nm}^k(t_i) = 0, \text{ if } d_{nm}(t_i) > c_n \\ C3: a_{nm}^k(t_i) = 0, \text{ if } q_k \notin Q_n(t_i) \\ C4: \sum_{m=1}^{|B_i|} a_{nm}^k(t_i) \leq 1, \forall q_k \in Q_n(t_i) \\ C5: \sum_{j=i}^{|T|} a_{nm}^k(t_j) \leq \left\lfloor \frac{\min(t_m(t_i) - t_i, t_m^c)}{\beta} \right\rfloor \end{cases}$$

C2 Vehicles can only be allocated channels within the coverage of edge nodes
 C3 The channel allocated by the edge node must be the currently available channel
 C4 Each channel of the edge node is assigned to at most one task
 C5 The total length of the channel occupation time slice needs to be less than the task deadline and the channel maintenance time

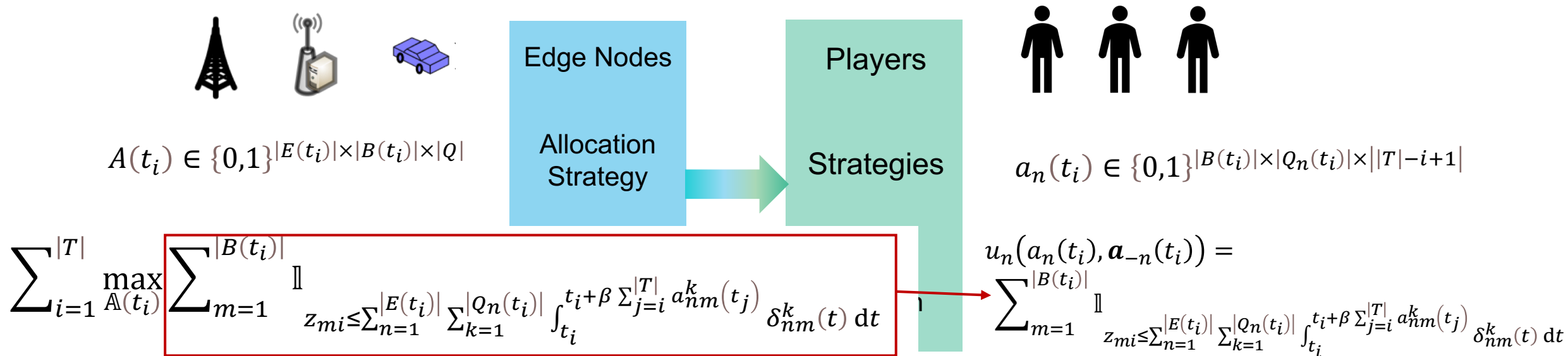


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► Potential Game Model Distributed Game Transformation



- Potential Function

$$\phi_n(a_n(t_i), \mathbf{a}_{-n}(t_i)) = \sum_{m=1}^{|B(t_i)|} \mathbb{1}_{z_{mi} - \theta_{m, E(t_i)/\{n\}} \leq \sum_{k=1}^{|Q_n(t_i)|} \int_{t_i}^{t_i+\beta \sum_{j=i}^{|T|} a_{nm}^k(t_j)} \delta_{nm}^k(t) dt}$$
- Exact Potential Game

$$u_n(a_n(t_i)^*, \mathbf{a}_{-n}(t_i)) - u_n(a_n(t_i), \mathbf{a}_{-n}(t_i)) = \phi_n(a_n(t_i)^*, \mathbf{a}_{-n}(t_i)) - \phi_i(a_n(t_i), \mathbf{a}_{-n}(t_i))$$

Theorem: Given the potential function of the game, the channel allocation game is an exact potential game, which possesses at least one Nash Equilibrium



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► Algorithm Design Incentive-based Probability Update and Strategy Selection (IPUS²)

Initialization



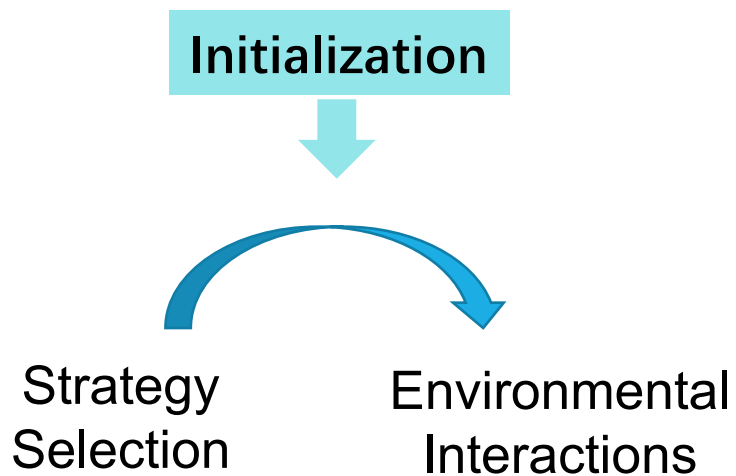
■ Strategy Selection Probability

$$x_{n,0}(a_n(t_i)) = 1/|\mathcal{A}_n(t_i)|$$

	a_1	a_2
b_1	0.5	1
b_2	1	2

Iteration	0	1	2	3	4
Strategy	/	$a_1 b_1$	$a_2 b_1$	$a_1 b_1$	$a_2 b_2$
$s(a_1)$	0	0	0	-0.5	0
$s(a_2)$	0	0	1	0	2
$s(b_1)$	0	0	1	-0.5	0
$s(b_2)$	0	0	0	0	2
$x_A(a_1)$	0.5	0.5	0.5	0.375	0.375
$x_A(a_2)$	0.5	0.5	0.75	0.75	1
$x_B(b_1)$	0.5	0.5	0.75	0.6625	0.6625
$x_B(b_2)$	0.5	0.5	0.5	0.5	1

► Algorithm Design Incentive-based Probability Update and Strategy Selection (IPUS²)



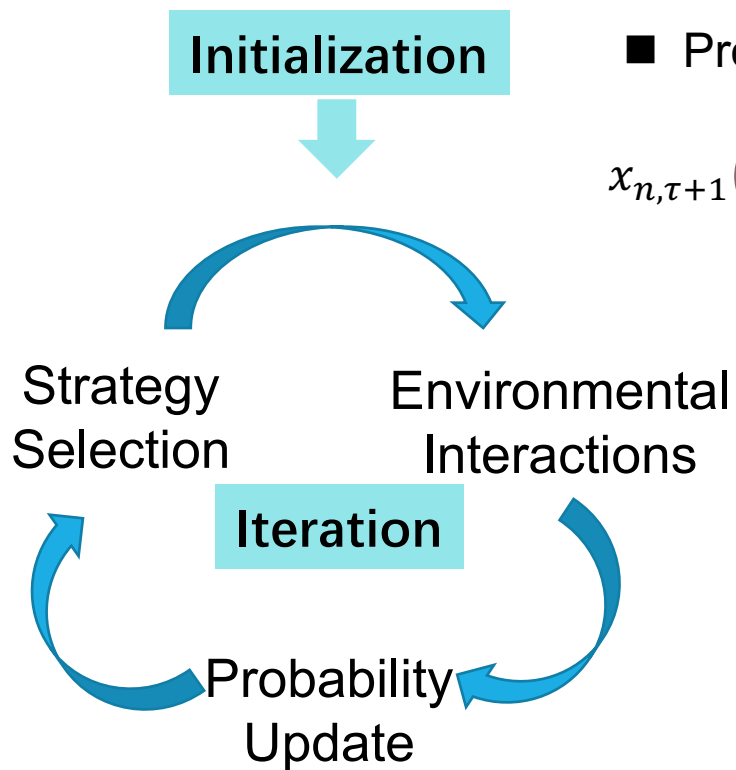
■ Probability Update Value

$$s_n(a_n(t_i)) = \frac{\xi_n(a_n(t_i)) - \max(\mathbb{E}_n(\mathcal{A}_n(t_i)))}{\max(\mathbb{E}_n(\mathcal{A}_n(t_i)))}$$

	a_1	a_2
b_1	0.5	1
b_2	1	2

Iteration	0	1	2	3	4
Strategy	/	$a_1 b_1$	$a_2 b_1$	$a_1 b_1$	$a_2 b_2$
$s(a_1)$	0	$(0.5-0.5)/0.5$	0	-0.5	0
$s(a_2)$	0	0	1	0	2
$s(b_1)$	0	$(0.5-0.5)/0.5$	1	-0.5	0
$s(b_2)$	0	0	0	0	2
$x_A(a_1)$	0.5	0.5	0.5	0.375	0.375
$x_A(a_2)$	0.5	0.5	0.75	0.75	1
$x_B(b_1)$	0.5	0.5	0.75	0.6625	0.6625
$x_B(b_2)$	0.5	0.5	0.5	0.5	1

► Algorithm Design Incentive-based Probability Update and Strategy Selection (IPUS²)



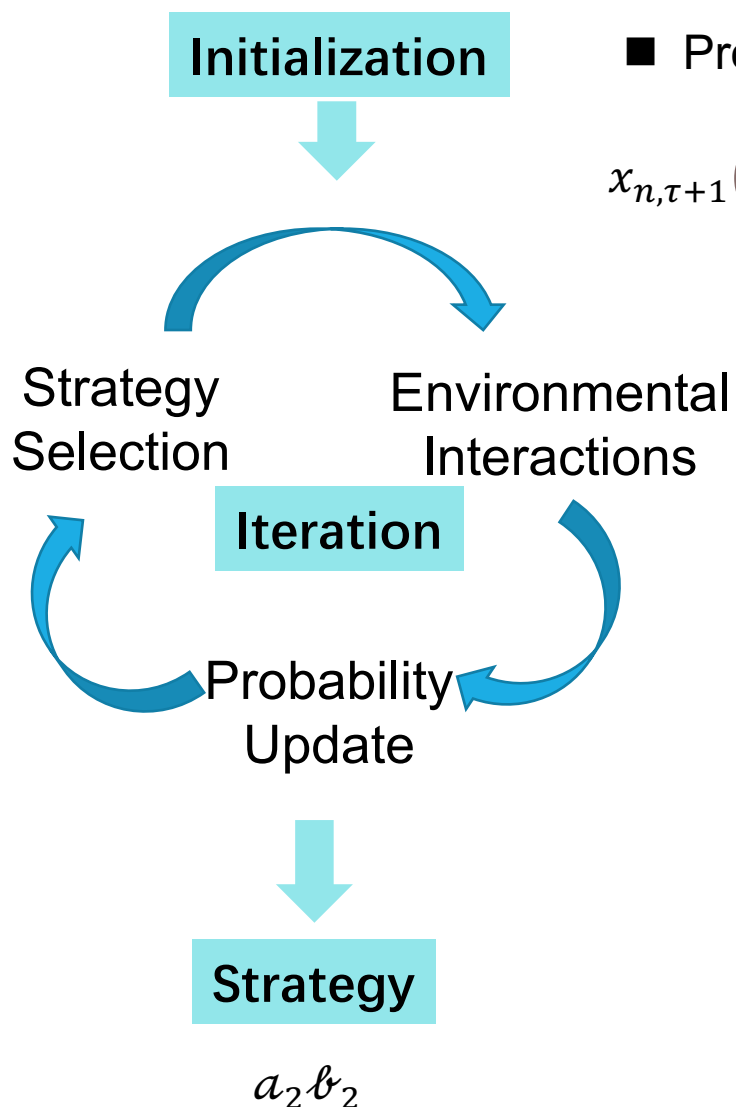
■ Probability Update Function

$$x_{n,\tau+1}(a_n(t_i)) = \begin{cases} x_{n,\tau}(a_n(t_i)) + \eta s_n(a_n(t_i)) (1 - x_{n,\tau}(a_n(t_i))) & s_n(a_n(t_i)) \geq 0 \\ x_{n,\tau}(a_n(t_i)) + \eta s_n(a_n(t_i)) x_{n,\tau}(a_n(t_i)) & s_n(a_n(t_i)) < 0 \end{cases}$$

	a_1	a_2
b_1	0.5	1
b_2	1	2

Iteration	0	1	2	3	4
Strategy	/	$a_1 b_1$	$a_2 b_1$	$a_1 b_1$	$a_2 b_2$
$s(a_1)$	0	0	0	-0.5	0
$s(a_2)$	0	0	1	0	2
$s(b_1)$	0	0	1	-0.5	0
$s(b_2)$	0	0	0	0	2
$x_A(a_1)$	0.5	0.5	0.5	0.375	0.375
$x_A(a_2)$	0.5	0.5	0.75	0.75	1
$x_B(b_1)$	0.5	0.5	0.75	0.6625	0.6625
$x_B(b_2)$	0.5	0.5	0.5	0.5	1

► Algorithm Design Incentive-based Probability Update and Strategy Selection (IPUS²)



■ Probability Update Function

$$x_{n,\tau+1}(a_n(t_i)) = \begin{cases} x_{n,\tau}(a_n(t_i)) + \eta s_n(a_n(t_i)) (1 - x_{n,\tau}(a_n(t_i))) & s_n(a_n(t_i)) \geq 0 \\ x_{n,\tau}(a_n(t_i)) + \eta s_n(a_n(t_i)) x_{n,\tau}(a_n(t_i)) & s_n(a_n(t_i)) < 0 \end{cases}$$

	a_1	a_2
b_1	0.5	1
b_2	1	2

Iteration	0	1	2	3	4
Strategy	/	$a_1 b_1$	$a_2 b_1$	$a_1 b_1$	$a_2 b_2$
$s(a_1)$	0	0	0	-0.5	0
$s(a_2)$	0	0	1	0	2
$s(b_1)$	0	0	1	-0.5	0
$s(b_2)$	0	0	0	0	2
$x_A(a_1)$	0.5	0.5	0.5	0.375	0.375
$x_A(a_2)$	0.5	0.5	0.75	0.75	1
$x_B(b_1)$	0.5	0.5	0.75	0.6625	0.6625
$x_B(b_2)$	0.5	0.5	0.5	0.5	1



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► Performance Evaluation Experiment Settings

■ Experiment Dataset

- Taxi Trajectories in 3x3 km Area of Chengdu City
- 2 Base Stations, 9 RSUs, 5 Vehicular Edge Nodes

■ Price of Anarchy, PoA

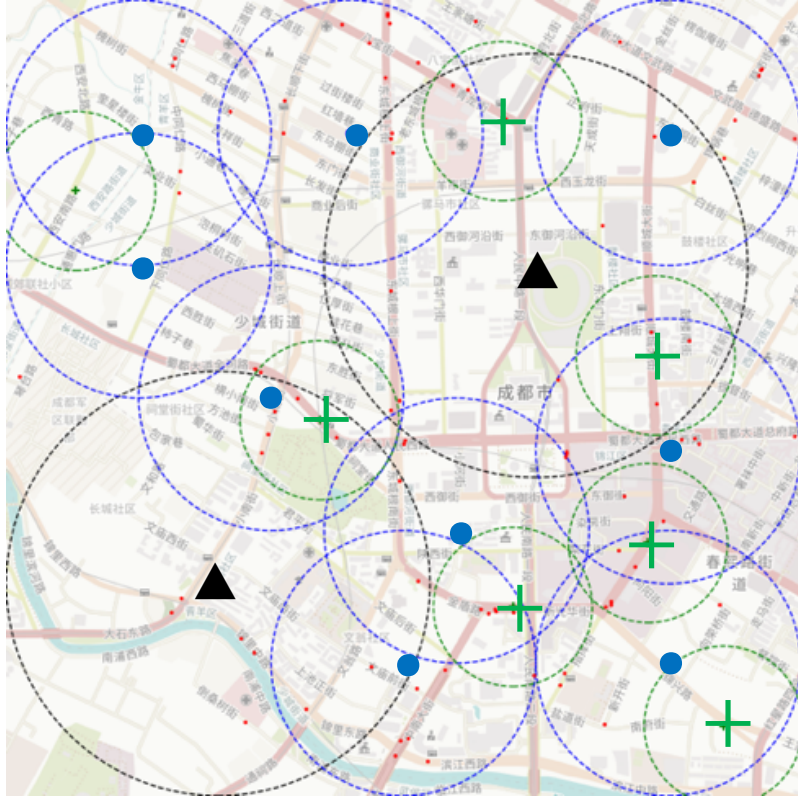
$$\text{PoA} = \frac{\max_{a \in \mathbb{A}} w(a)}{\min_{a^* \in \mathbb{A}^{\text{NE}}} w(a^*)}$$

■ Task Completion Ratio

$$\mathcal{U}(t_i) = \frac{\sum_{m=1}^{|B(t_i)|} \mathbb{I}_{z_{mi} \leq \sum_{n=1}^{|E(t_i)|} \sum_{k=1}^{|Q_n(t_i)|} \int_{t_i}^{t_i + \beta \sum_{j=i}^{|T|} a_{nm}^k(t_j)} \delta_{nm}^k(t) dt}}{|B(t_i)|}$$

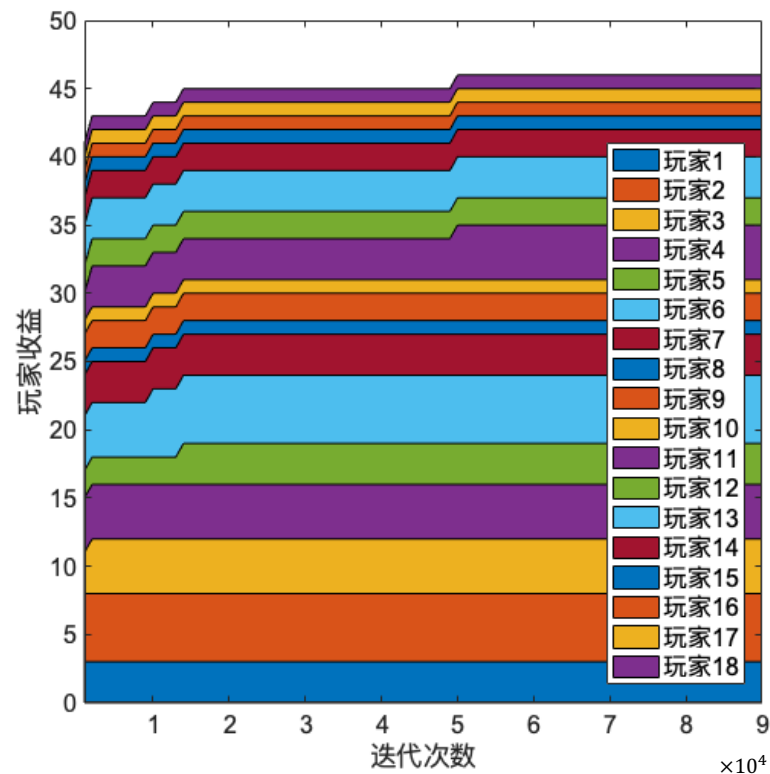
■ Channel Utilization Efficiency

$$\mathcal{E}(t_i) = \sum_{m=1}^{|B(t_i)|} \frac{\sum_{n=1}^{|E(t_i)|} \sum_{k=1}^{|Q_n(t_i)|} \int_{t_i}^{t_i + \beta \sum_{j=i}^{|T|} a_{nm}^k(t_j)} \delta_{nm}^k(t) dt}{\sum_{k=1}^{|Q_n(t_i)|} \sum_{j=i}^{|T|} a_{nm}^k(t_j)} / |B(t_i)|$$

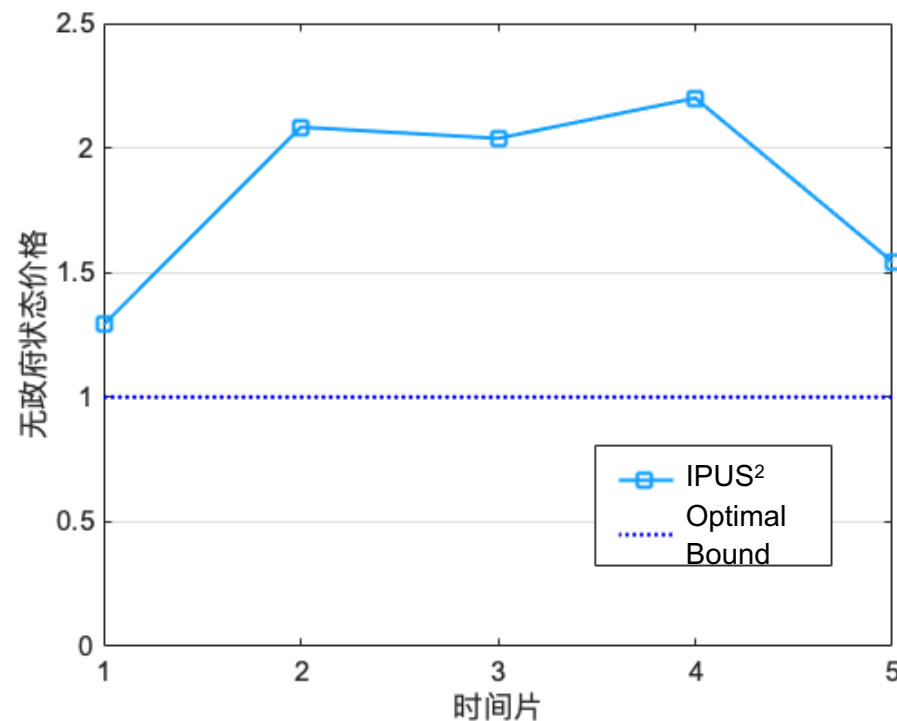


▲ Base Station ● RSU + Vehicular Edge Node

► Performance Evaluation Simulation Results



The Convergence of The IPUS²



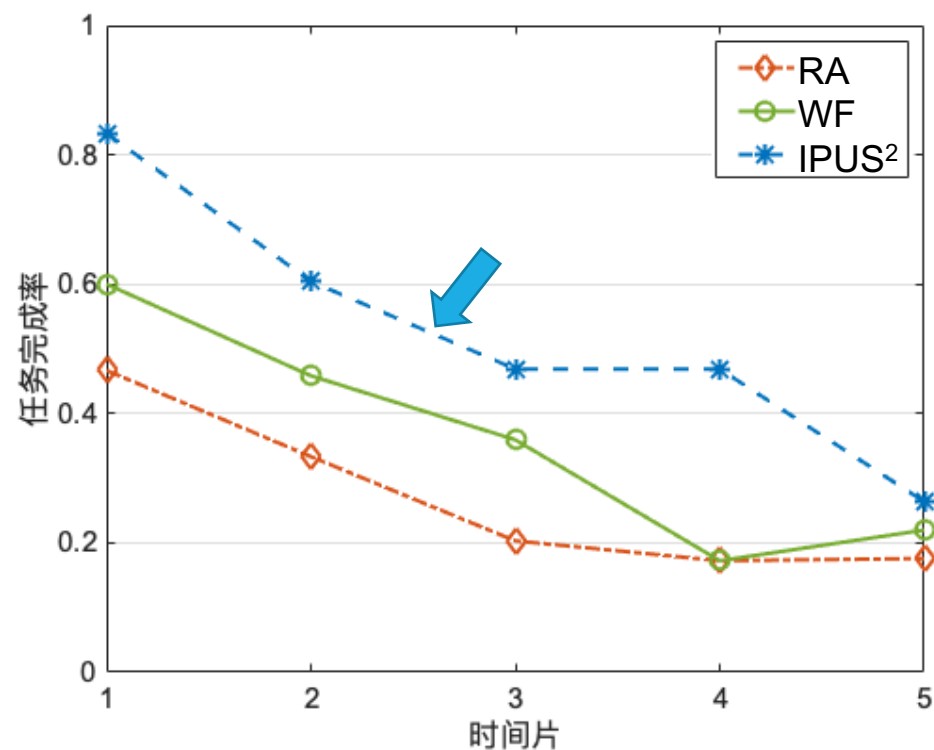
The Effectiveness of The Nash Equilibrium

► Performance Evaluation Simulation Results

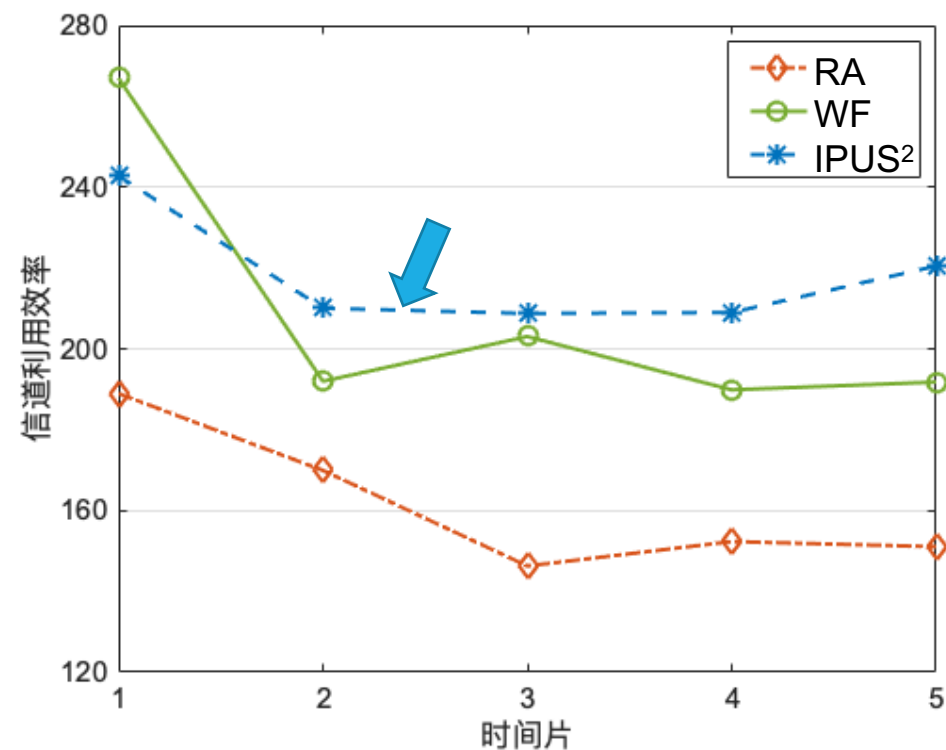
The Comparative Algorithms

- Random Allocation, RA

- Water Filling, WF



Task Completion Ratio



Channel Utilization Efficiency

► Conclusion

- We **formulated the problem** of channel allocation in vehicular edge computing, and transformed the global optimization problem into a **distributed channel allocation potential game**
- We proposed an Incentive-based Probability Update and Strategy Selection (**IPUS²**) algorithm, and verified the **convergence** of the IPUS² and the **effectiveness** of the Nash Equilibrium
- We give an experiment with **real vehicle trajectories**, and the results showed **IPUS² outperforms** existing representative algorithms

Thank You