



中国科学技术大学
University of Science and Technology of China



Vehicle Rebalancing in a Shared Micromobility System with Rider Crowdsourcing

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■ Education

- Pursuing his Doctor of Philosophy by Dr Pan Kai, Hong Kong Polytechnic University
- Master degree, KTH Royal Institute of Technology, Sweden (2020)
- Bachelor degree, South China University of Technology, China (2018)

■ Publication

- International Transactions in Operational Research
- IEEE Transactions on Industrial Informatics
- MSOM

■ Research Areas

- Stochastic Optimisation
- Integer Programming
- Robust and Data-Driven Optimisation



Wang Yulan (*Professor, The Hong Kong Polytechnic University*)

■ Education

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- Master degree, Bachelor degree, Shanghai Jiao Tong University

■ Publication (66 paper, 2642 citations)

- Management Science
- Operations Research
- Manufacturing & Service Operations Management
- Production and Operations Management

■ Research Areas

- Operations and supply chain management
- Socially sustainable operations
- Behavioral operations
- Operations-marketing-IS (information systems) interface



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- PHD, Georgia Institute of Technology, 2005
- Master degree, National University of Singapore, 1998
- Bachelor degree, National Central University, 1996

■ Publication (53 paper, 1000 citations)

- Management Science
- Operations Research
- Manufacturing & Service Operations Management
- Production and Operations Management

■ Research Areas

- e-Commerce and Marketplace Analytics
- Online Platforms
- Warehousing and Fulfillment
- Sustainable Urban Logistics
- Flexible Workforce and Resource Management



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■ Education

- PHD, University of Florida, 2012-2016
- Master degree, University of Florida, 2012-2014
- Bachelor degree, Zhejiang University, 2006-2010

■ Publication (UTD24 (9), IEEE (8), IISE Trans./EJOR (4))

- Manufacturing and Service Operations Management
- INFORMS Journal on Computing
- IEEE
- POM
- EJOR

■ Research Areas

- Stochastic and Discrete Optimization
- Robust and Data-Driven Optimization
- Energy Market, Smart City Operations
- Marketing, Supply Chain
- Transportation, Information Systems



Max Shen Zuo-Jun (Vice-President, the University of Hong Kong)

■ Education

- PHD, Northwestern University
- Master degree, Tsinghua University
- Bachelor degree, Shandong University

■ Publication (17099 citations)

- Associate editor for Operations Research
- Associate editor for Management Science
- Associate editor for MSOM

■ Research Areas

- Integrated **Supply Chain** Design and Management
- Data Driven Logistics and Supply Chain Optimization
- Design and Analysis of Optimization Algorithms
- Energy Systems Optimization
- **Transportation System Planning**



1. Background



- A two-stage stochastic mixed-integer program:
 - make the initial vehicle allocation decisions;
 - subsequent make relocation decisions to maximize the expected profit.



- A micromobility operator:
 - **3PL;**
 - **Crowdsourced Riders.**



2. Related literature

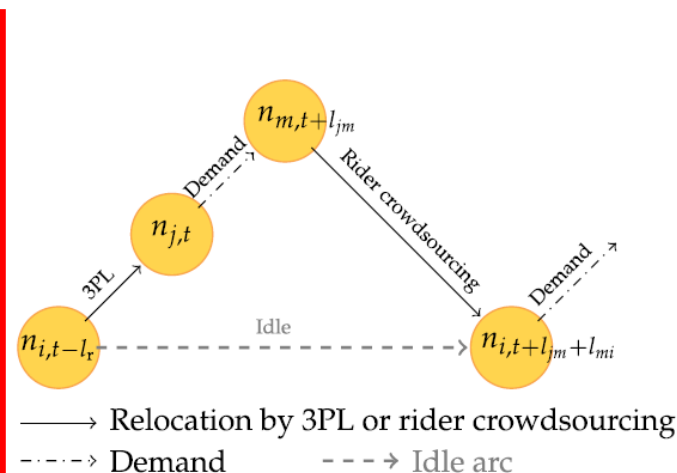
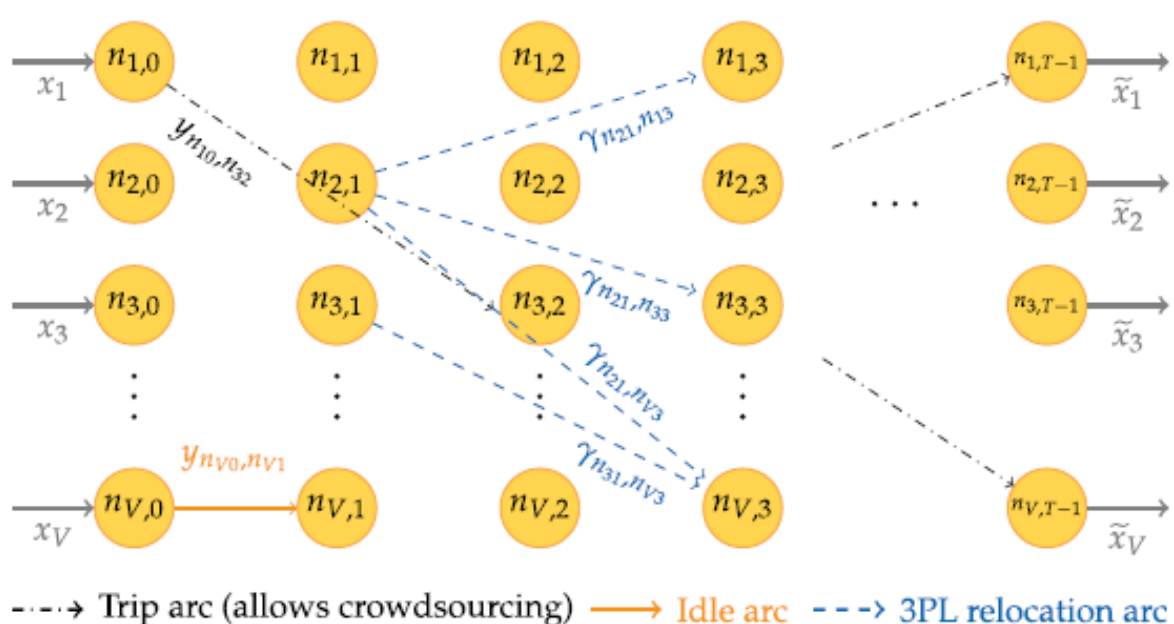


Related area	Research content	Published research	Research Gap
OM in shared mobility	vehicle-sharing systems: rent or hail cars from individual	He et al. (2017); Chang et al. (2017); Lu et al. (2018)	Cars vs bike/ e-bike
	Shared micromobility	Kabra et al. (2020); Shu et al. (2013)	Optimize the bicycle flows; initial allocation is given
Fleet operations of shared mobility	inventory redistribution or transshipment (cars)	Benjaafar et al. 2022; Miller-Hooks (2011); He et al. (2020)	Focus on shared cars
	inventory redistribution or transshipment (bicycle)	Dell'Amico et al. (2014); Freund et al. (2020); Li and Liu (2021)	Ignore the capacity design; two relocation strategies
Crowdsourcing	Crowdsourcing in last mile delivery operations	Qi et al. 2018; Fatehi and Wagner 2022; Kafle et al. 2017	two relocation strategies under different temporal demand patterns

3. Model



Construct a time-space network with multiple service regions:

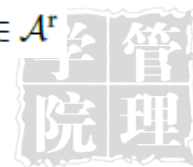


\mathcal{V} set of service regions $\mathcal{V} = \{1, 2, \dots, V\}$
 \mathcal{T} operational horizon $\mathcal{T} = \{0, 1, \dots, T-1\}$

x_i number of vehicle allocated to region $i \in \mathcal{V}$
 \tilde{x}_i number of vehicles in the ending period $T-1$ in region $i \in \mathcal{V}$

y_a realized flow on arc $a = (n_{it}, n_{j,t+l_{ij}}) \in \mathcal{A}^i \cup \mathcal{A}^t$

γ_a number of vehicles relocated by the 3PL from region i in period t to region j for each arc $a = (n_{it}, n_{j,t+l_r}) \in \mathcal{A}^r$



3. Model



The operator optimizes the vehicle allocation decisions

$$\begin{aligned} \Gamma = \min_{\mathbf{x}} \sum_{j \in \mathcal{V}} c_j x_j + \Theta(\mathbf{x}) \\ \text{s.t. } \mathbf{x} \in \mathcal{X} = \left\{ \mathbf{x} \in \mathbb{Z}_+^{|\mathcal{V}|} \mid x_j \leq B_j, j \in \mathcal{V}, \sum_{j \in \mathcal{V}} x_j \leq N \right\}, (\mathcal{M}) \end{aligned}$$

where $\Theta(\mathbf{x})$ is realized in the **second** stage as the operator optimizes the **vehicle relocation** using rider **crowdsourcing** and the **3PL**.

$$\begin{aligned} \Theta(\mathbf{x}) = \min_{Y^k \in \mathcal{Y}(\mathbf{x}, \lambda^k), k \in \mathcal{K}} \left\{ \sum_{k \in \mathcal{K}} p^k \left[\sum_{a \in \mathcal{A}^t} \left(C_p \eta_a^k + \phi_a^k - R l_a (y_a^k - \Lambda_a^k) \right) \right. \right. \\ \left. \left. + \sum_{t \in T(l_r)} C_r z_t^k \right] \right\}. \quad (\mathcal{P}) \end{aligned}$$

Given the **initial vehicle allocation** \mathbf{x} , the expected net cost $\Theta(\mathbf{x})$ in the second stage can be determined by solving the **network flow optimization problem**.



4. Comments



■ Positive comments:

- ✓ Construct a time-space network (fit the problem)
- ✓ Captures the main features of the problem (initial allocation; vehicle relocation; budget; travel time; penalty cost, etc.)
- ✓ Decomposition and acceleration strategies (based on time and scenarios)
- ✓ Good data and good figures

■ Negative comments:

- ✓ The method of dealing with demand uncertainty is crude (number of the scenarios is too large)

$$\Theta(x) = \min_{Y^k \in \mathcal{Y}(x, \lambda^k), k \in \mathcal{K}} \left\{ \sum_{k \in \mathcal{K}} p^k \left[\sum_{a \in A^t} (C_p \eta_a^k + \phi_a^k - R l_a (y_a^k - \Lambda_a^k)) + \sum_{t \in T(l_r)} C_r z_t^k \right] \right\}. \quad (\mathcal{P})$$





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Thank You!

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