

A Metaheuristic Approach for 3L-SDVRP

Fei Liu, Qingling Zhu

Department of Computer Science City University of Hong Kong

March 11, 2021



Outline

- 1. EMO2021 Competition Huawei 3L-SDVRP
- 2. A Metaheuristic Approach for 3L-SDVRP
- 3. Conclusion

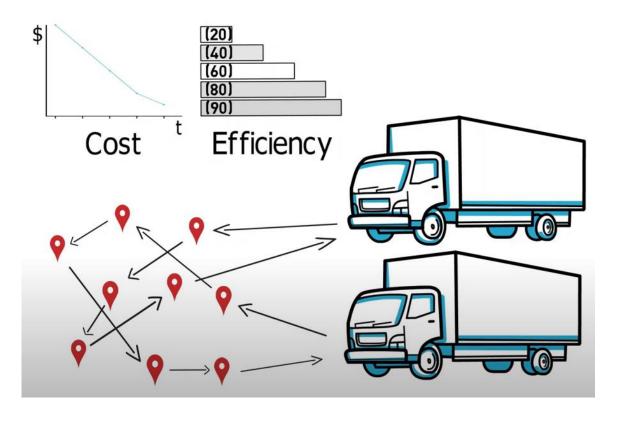


Outline

- 1. EMO2021 Competition Huawei 3L-SDVRP
- 2. A Metaheuristic Approach for 3L-SDVRP
- 3. Conclusion

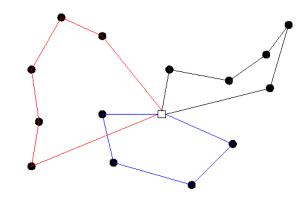


Vehicle Routing Problem (VRP)



• https://www.youtube.com/watch?v=OKMssWdC0I0

Vehicle Routing Problem (VRP)

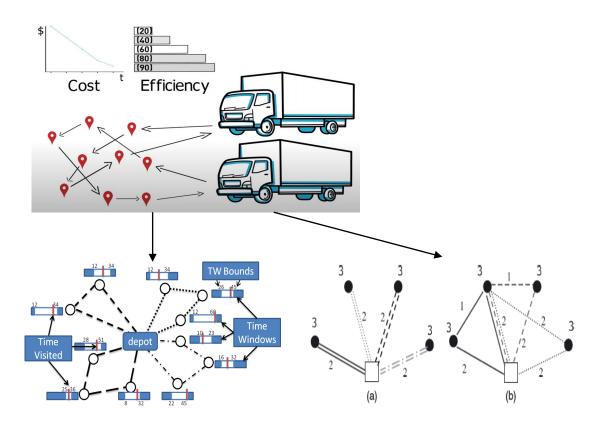


Four components

- > Network
- Sites to be visited (customers to serve, tasks to process, etc.)
- > Fleet of vehicles
- > Depot(s)



Variants of VRP



Objectives: Cost & Efficiency

Constraints: Limited capacities

Many depots

Customer demand

Time windows

Precedence and synchronization

• • •

Typical VRPs: Capacitated Vehicle Routing Problem (CVRP)

Vehicle Routing Problem with Backhauls (VRPB)

Vehicle Routing Problem with Split Deliveries (VRPSD)

Vehicle Routing Problem with Multiple Depot (VRPMD)

Vehicle Routing Problem with Time Windows (VRPTW)

• • •

• https://www.youtube.com/watch?v=OKMssWdC0I0

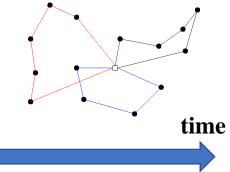


Methods for VRP

Dantzig and Ramser (1959) the first to introduce "Truck **Dispatching Problem**"

Clarke Wright (1964)and generalized this problem to a linear optimization problem "Vehicle Routing Problem"

Lenstra and Rinnooy Kan (1981) VRP is an NP-hard proved problem, exact algorithms are only efficient for small problems



1950

1970

1980

2000

5. Machine Learning:

Reinforcement Learning

Pointer network

1. Constructive Heuristics:

1960

Savings heuristic

Sweep algorithm

2. Improvement Heuristics:

K-opt

λ-interchange

3. Exact algorithms:

Branch and Bound

Cutting Plane

Network-flows

Dynamic Programming

4. Metaheuristics:

Tabu search Genetic Algorithm

Simulated Annealing GRASP

Local search methods

Partical Swarm Optimization

Ant Colony Algorithm

Perhaps the most famous heuristic of this category is the Clarke and Wright (1964) savings heuristic

The development of exact algorithms for the VRP took off in 1981 with the publication of two papers by Christofides

The development of modern heuristics for the VRP really started in the 1990s with the advent of metaheuristics.

^{6/23}



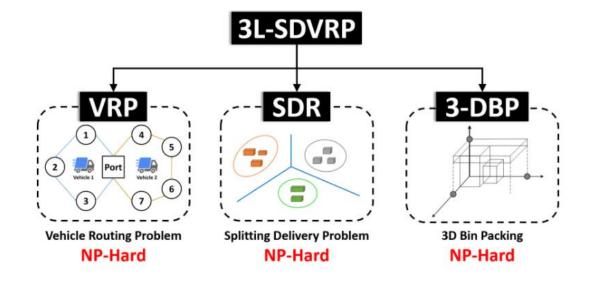
EMO2021 Competition – Huawei 3L-SDVRP

It involves multiple-pickup-points, single-delivery-point, many types of cargoes, multiple kinds of trucks with containers, and various routes to the port or airport.

The two objectives of the MOCO problem are to <u>maximize the loading rate</u> as well as <u>to minimize the traveling distance</u> under specific constraints and requirements.

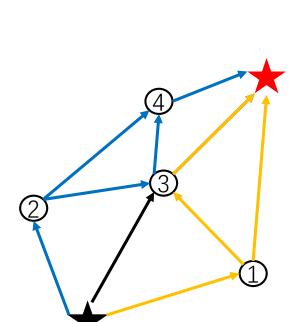


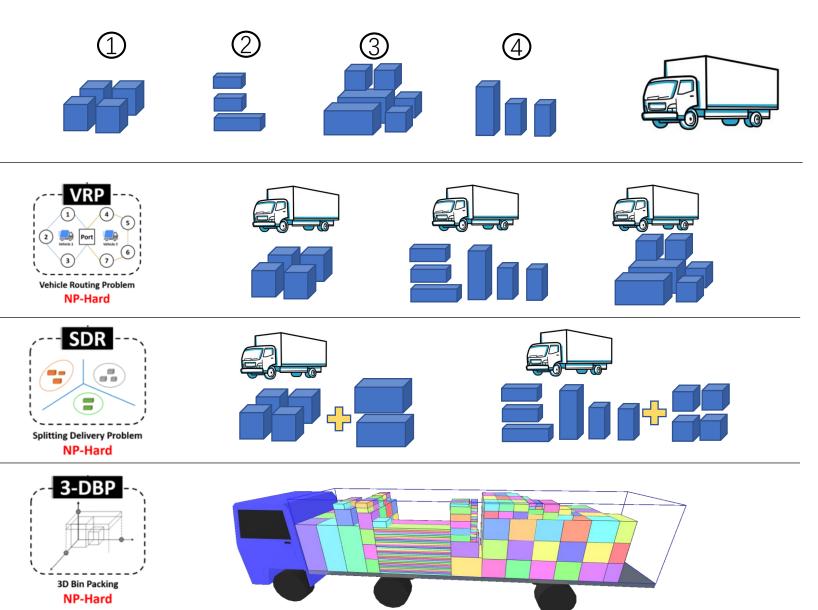






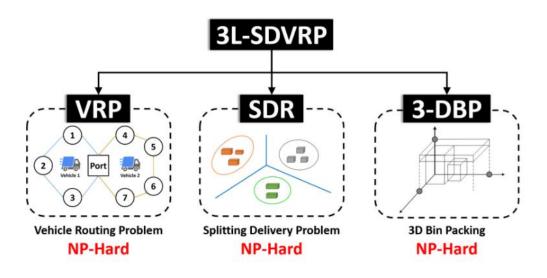
電腦科學系 Department of Computer Science







EMO2021 Competition – Huawei 3L-SDVRP



Objective functions

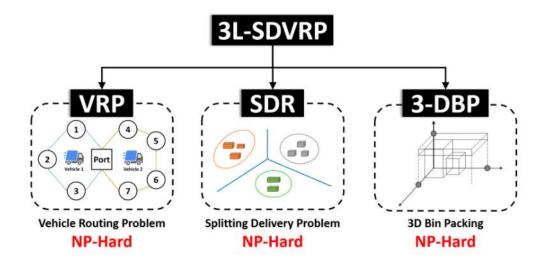
- $$\begin{split} \bullet & \ f_1 = 1 \sum_{i=1}^n loading_rate_i/n \,, \\ & \ \text{where } loading_rate_i = max(v_rate_i, w_rate_i) \,, \\ & \ v_rate_i = \text{the total volume of all the items in truck } i \,/ \text{ the size of the truck } i \,, \\ & \ w_rate_i = \text{the total weight of all the items in truck } i \,/ \text{ the weight capacity of the truck } i \,. \end{split}$$
- $f_2 = \sum_{i=1}^n truck_distance_i$ where $truck_distance_i$ is the distance traveled by the truck i .

Constraints

This problem includes two types of constraints: vehicle routing and 3-D loading,

- · valid route constraints: each route must start from the starting point and end at the delivery point
- bonded warehouse constraints: the bonded warehouse is a special pickup point that can only be
 visited by empty trucks. Hence, if a pickup point is marked as a bonded warehouse, it must be the first
 of any route that contains it
- single visit constraints: each truck can only visit each pickup point once
- none-overlapping constraints: items cannot overlap in any dimension
- none-splitting constraints: one item can only be loaded in one truck
- supporting constraints: the item must have sufficient supporting surface, i.e., the supporting area should be greater than 80% of the bottom area of the item
- weight capacity constraints: the total weight of the cargoes loaded on one truck cannot exceed the weight capacity of the truck
- · size constraints: all the items must be loaded inside the container
- one-time loading constraints: once a truck leaves a pickup point, the items loaded at the pickup point cannot be moved anymore





The three approaches can hardly solve 3D-SDVRP directly

Exact algorithms

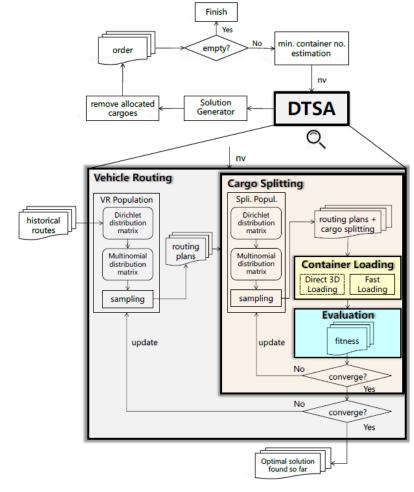
Heuristics

Machine Learning

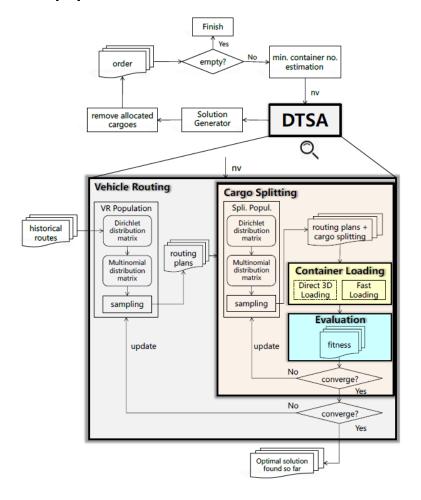


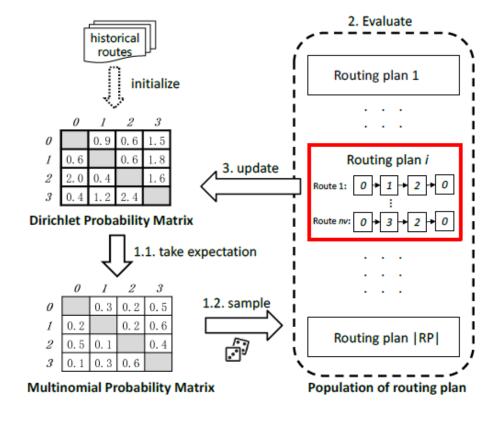
Applied Data Science Track Paper KDD 2018, August 19-23, 2018, London, United Kingdom A Data-Driven Three-Layer Algorithm for Split Delivery Vehicle **Routing Problem with 3D Container Loading Constraint** Mingxuan Yuan Di Chen Xijun Li Shanghai Jiao Tong University Noah's Ark Lab Noah's Ark Lab Noah's Ark Lab of Huawei Huawei Technologies Huawei Technologies di.chen@huawei.com lixijun@sjtu.edu.cn yuan.mingxuan@huawei.com Jianguo Yao Jia Zeng Shanghai Jiao Tong University Noah's Ark Lab jianguo.yao@sjtu.edu.cn Huawei Technologies Zeng.Jia@huawei.com

- 1. First propose 3L-SDVRP (Split Delivery Vehicle Routing Problem with 3D Loading Constraints)
- 2. Three-Layer method
- Out-layer: vehicle routing
- Mid-layer: cargo splitting
- Inner-layer: container loading

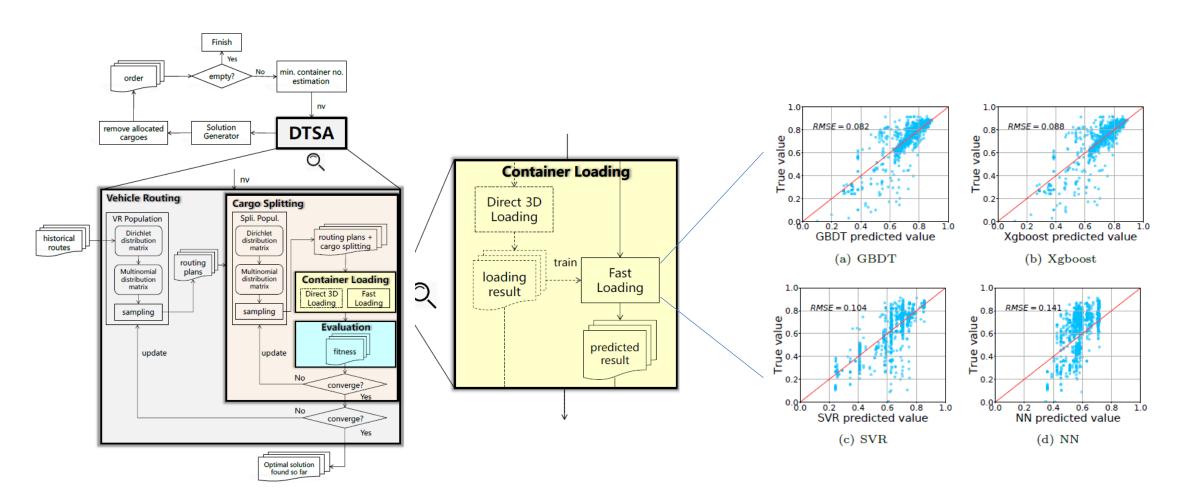




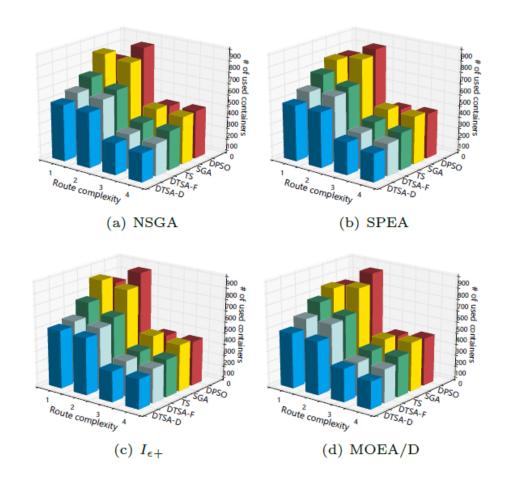


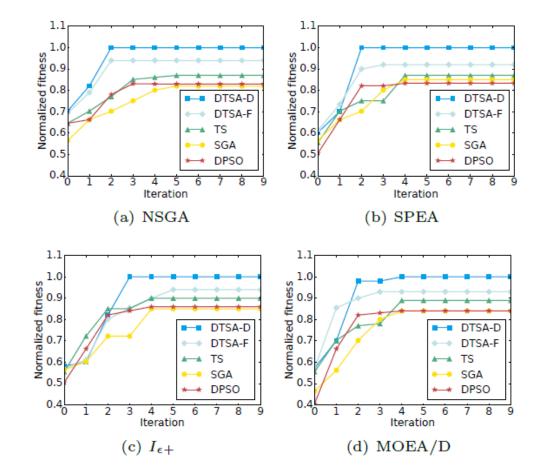














電腦科學系 Department of Computer Science

Approaches for 3L-SDVRP

European Journal of Operational Research 282 (2020) 545-558



Contents lists available at ScienceDirect

European Journal of Operational Research

journal homepage: www.elsevier.com/locate/ejor



Production, Manufacturing, Transportation and Logistics

The Split Delivery Vehicle Routing Problem with three-dimensional loading constraints



Andreas Bortfeldta, Junmin Yib,*

1. Introduce two variance of 3L-SDVRP

- 3L-SDVRP with *forced* splitting
 A delivery is only split if the demand of a customer cannot be transported by a single vehicle
- 3L-SDVRP with optional splitting
- 2. Packing first hybrid algorithm
- a local search algorithm for routing
- a genetic algorithm and several construction heuristics for packing.

Algorithm 1 Overview of the hybrid algorithm.

- SDVRLH2 (in: problem data, parameters, out: best solution s_{best}) // Packing step
- 2: **for** each customer *i* **do**
- generate patterns for customer i (1C-FLPs where necessary, one 1C-SP) by container loading GA
- 4: endfor
- 5: **for** selected customer pairs (ij) **do**
- generate 2C-SP pattern for customer pair (i, j) by construction heuristics
- 7: endfor

// Routing step

- 8: **for** each customer i
- 9: generate as many direct trips $0 \rightarrow i \rightarrow 0$ as 1C-FLPs for customer i do exist
- 10: endfor
- 11: if only forced splits allowed then
- 12: solve remaining CVRP by local search
- 13: else // optional splits allowed
- 14: solve remaining SDVRP by (modified) local search
- 15: endif

// Final step

16: prepare solution s_{best} consisting of best achieved routing plan and related 3D packing patterns

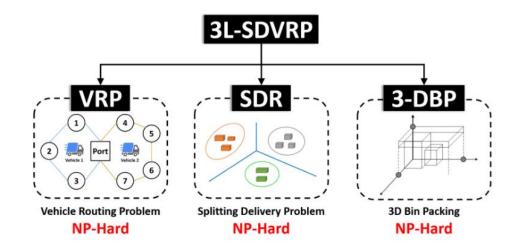
17: **end**.

^a Otto-von-Guericke-University Magdeburg, Germany

bXiamen University of Technology, China



電腦科學系 Department of Computer Science



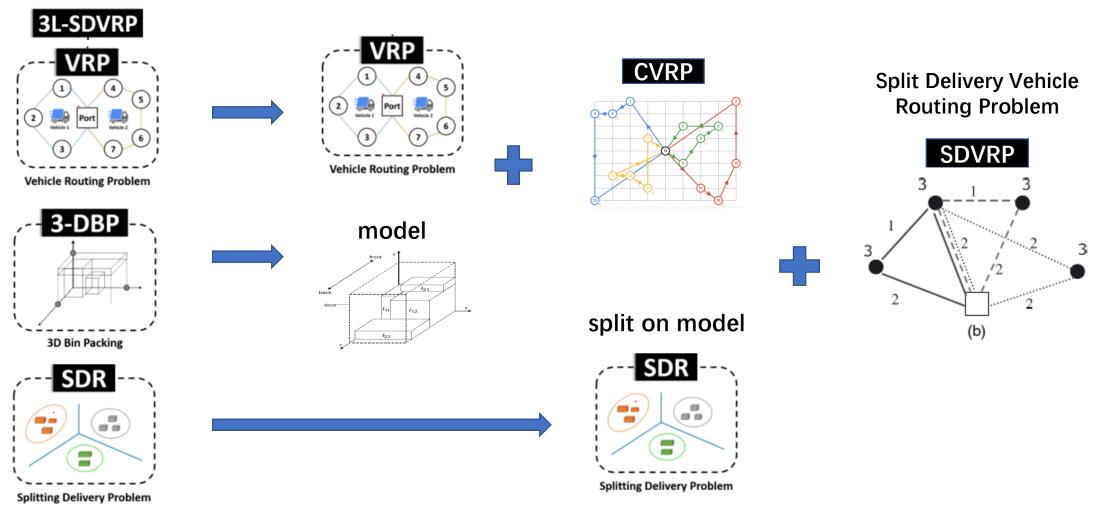
- 1. Proposed in 2018
- 2. Complex and time-consuming
- 3. There are three approaches:
 - Routing first
 - Packing first
 - Mixed routing & packing

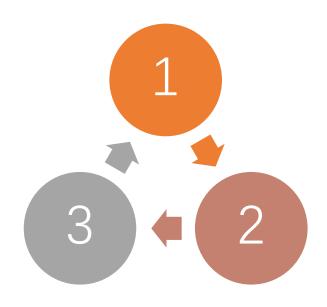


Outline

- 1. EMO2021 Competition Huawei 3L-SDVRP
- 2. A Metaheuristic Approach for 3L-SDVRP
- 3. Conclusion

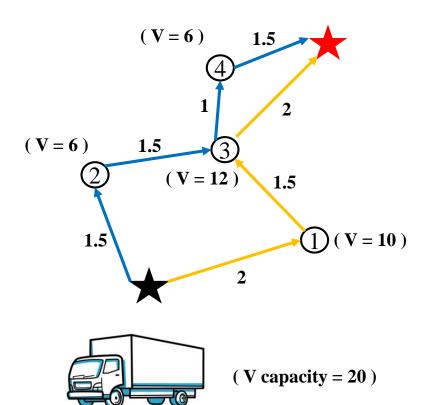






- 1. Objectives
- 2. Design space
- 3. Search method





Objectives:

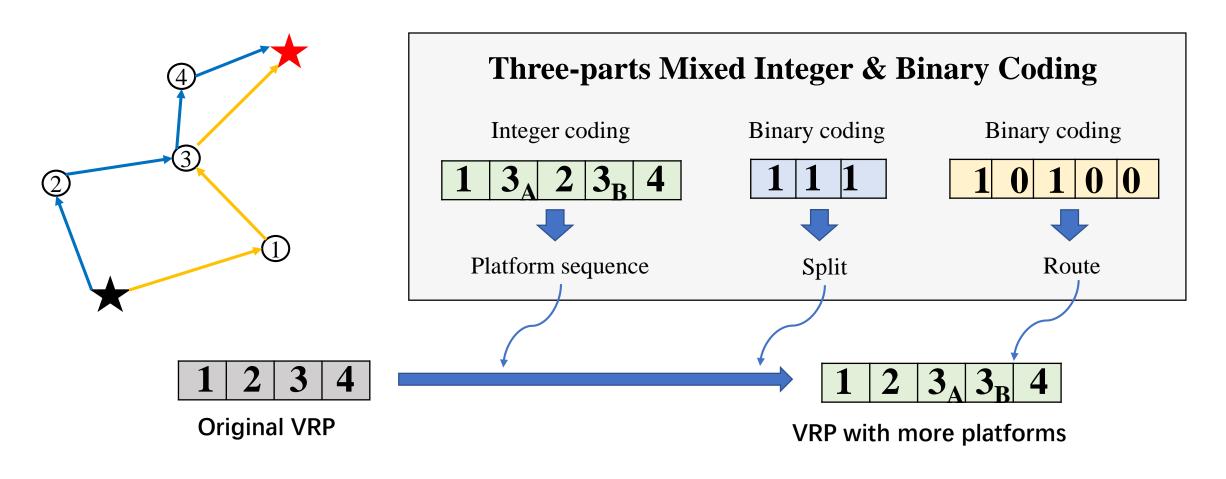
- $$\begin{split} \bullet & \ f_1 = 1 \sum_{i=1}^n loading_rate_i/n \ , \\ & \ \text{where } loading_rate_i = max(v_rate_i, w_rate_i) \ , \\ & \ v_rate_i = \text{the total volume of all the items in truck } i \ / \ \text{the size of the truck } i \ , \\ & \ w_rate_i = \text{the total weight of all the items in truck } i \ / \ \text{the weight capacity of the truck } i \ . \end{split}$$
- $f_2 = \sum_{i=1}^n truck_distance_i$ where $truck_distance_i$ is the distance traveled by the truck i .

Example:

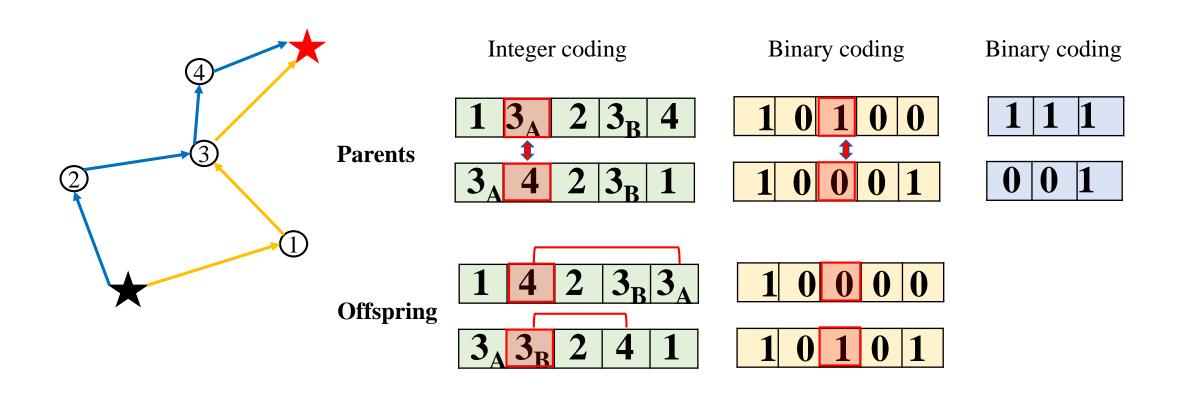
Objective 1: 1 - ((10+6)/20 + (6+6+6)/20)/2 = 0.15

Objective 2: (2+1.5+2) + (1.5+1.5+1+1.5) = 11

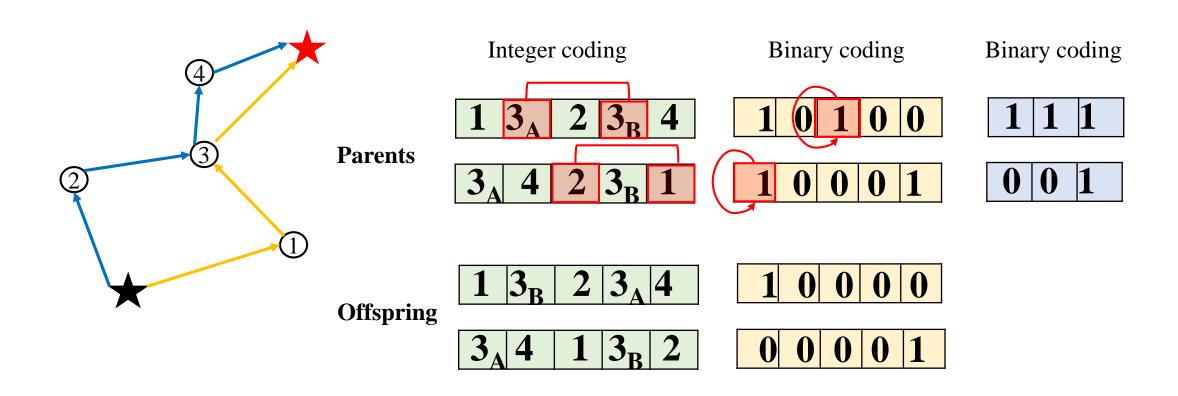












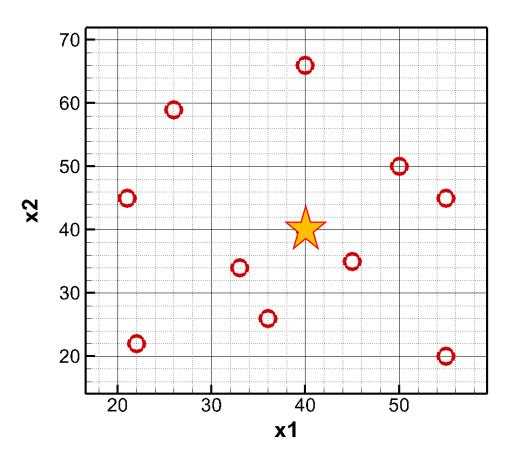


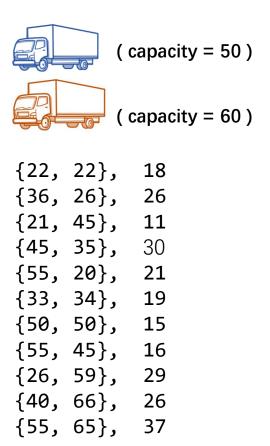
 $f_1(x)$

MOEA/D Integer coding Binary coding Binary coding Neighbour $|3_{\rm A}| |2| |3_{\rm R}| |4|$ 1 0 1 0 0 $f_2(x)$ Collaboration select from neighbour with ps Mating Swap mixed integer & binary Neighbour Mutation mixed integer & binary Replacement replace neighbour with pr



A simple SDVRP test instance



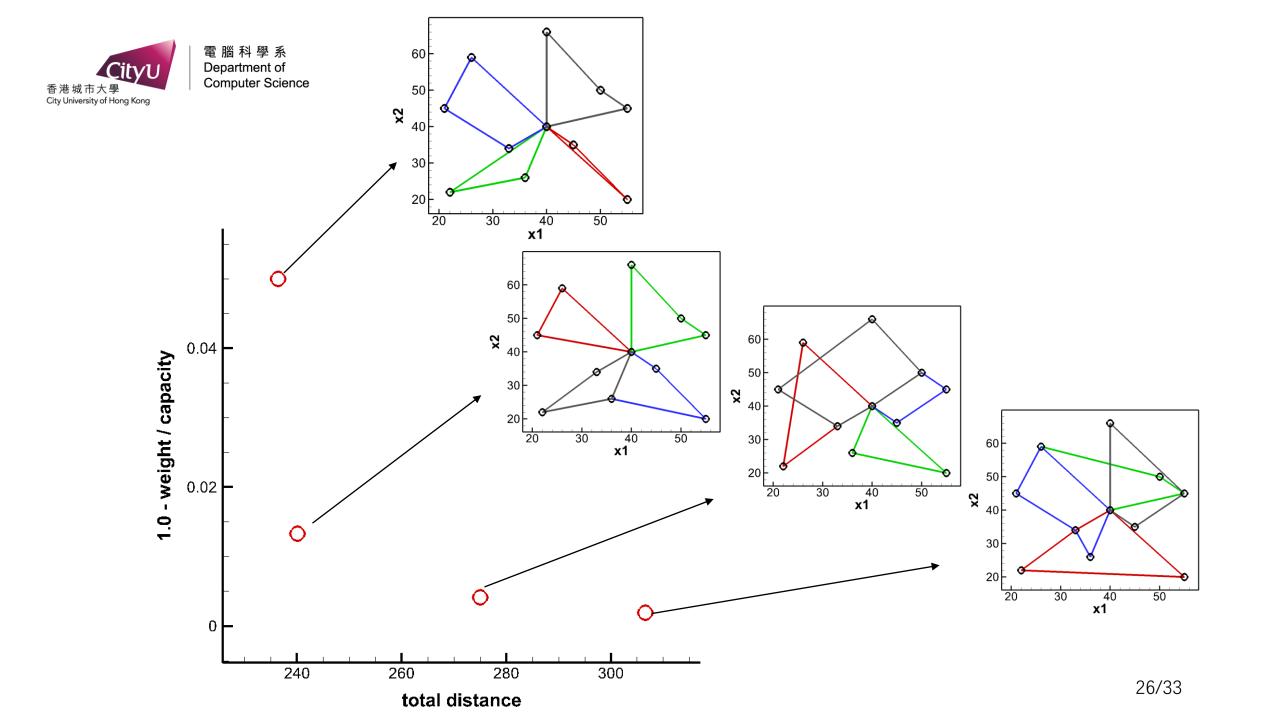


Two objectives:

$$f_1 = 1 - \sum_{i=1}^n loading_rate_i/n$$
 $f_2 = \sum_{i=1}^n truck_distance_i$

Constraints:

- Start & end nodes
- Max capacity
- 2 Splits





Advantages:

1. Adaptive& Flexible

2. Extensible

3. Efficient

4. Constraints

Adaptive to different size of VRP problems, different split number and vehicle type ...

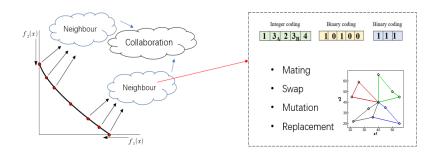
Combine with 3D bin packing method or used in other framework.

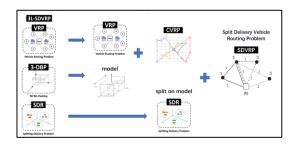
In general, more efficient than exactly method.

Constraints can be integrated easily.



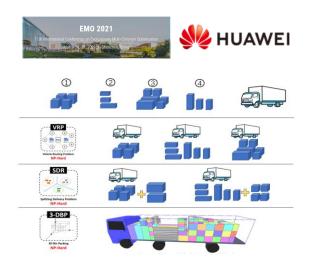
Proposed Multiobjective Metaheuristic





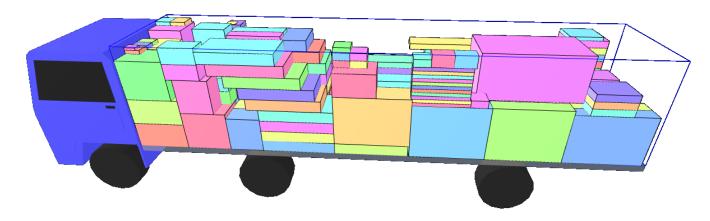
- Model for packing
- Constraints







A Multiobjective Metaheuristic Approach for 3L-SDVRP (model for 3D bin packing)



- V_boxs /V_truck
- W_boxs/W_truck
- N_platform
- N_type_of_box
- N_abnormal

• ...

Total volume of boxes / volume of truck

Total weight of boxes / capacity of truck

Number of platform

Number of types of boxes

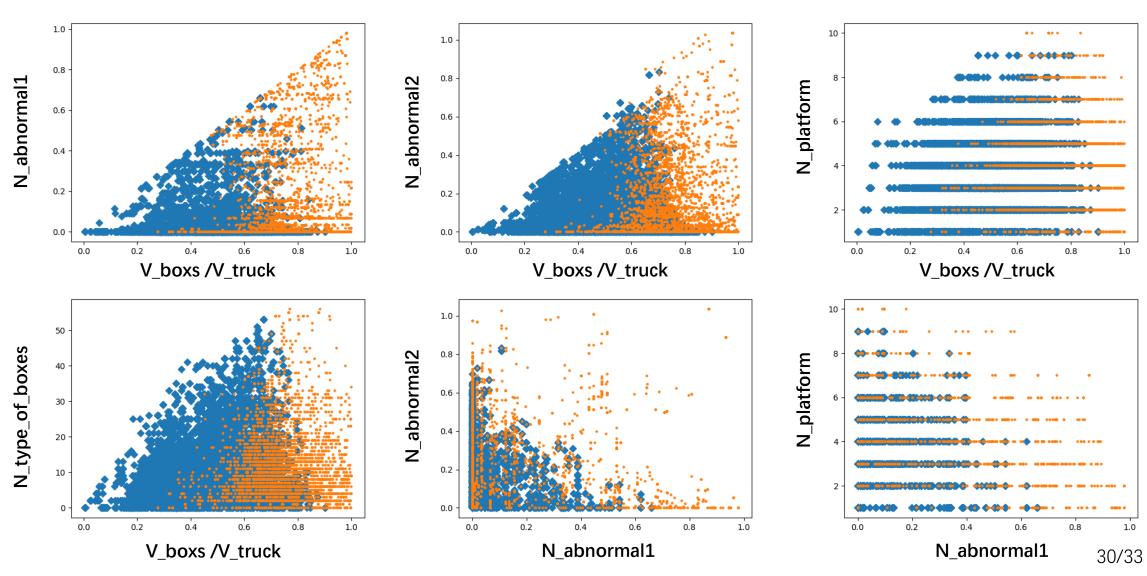
Number of abnormal boxes

...



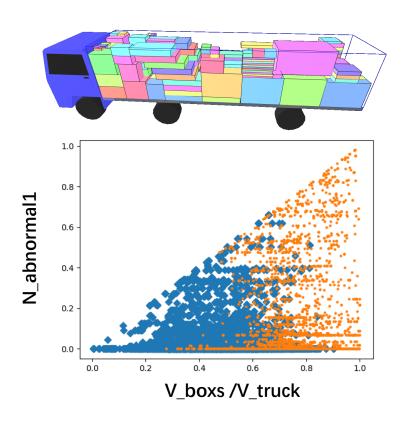
電腦科學系 Department of Computer Science

- 50000 data samples with label on if it is a successful packing
- 6 features
- V_box /V_truck is the most important feature





A Multiobjective Metaheuristic Approach for 3L-SDVRP (model for 3D bin packing)





- SVM, NN, linear or …
- Accuracy is not the most important thing?
- Regression or Classification
- How to define features



- Volume ratio is much more significant than other features
- Multi-level volume ratio is used as the model



A Multiobjective Metaheuristic Approach for 3L-SDVRP (constraints)

Constraints

This problem includes two types of constraints: vehicle routing and 3-D loading,

- · valid route constraints: each route must start from the starting point and end at the delivery point
- bonded warehouse constraints: the bonded warehouse is a special pickup point that can only be
 visited by empty trucks. Hence, if a pickup point is marked as a bonded warehouse, it must be the first
 of any route that contains it
- single visit constraints: each truck can only visit each pickup point once
- none-overlapping constraints: items cannot overlap in any dimension
- none-splitting constraints: one item can only be loaded in one truck
- supporting constraints: the item must have sufficient supporting surface, i.e., the supporting area should be greater than 80% of the bottom area of the item
- weight capacity constraints: the total weight of the cargoes loaded on one truck cannot exceed the weight capacity of the truck
- · size constraints: all the items must be loaded inside the container
- one-time loading constraints: once a truck leaves a pickup point, the items loaded at the pickup point cannot be moved anymore

Routing constraints

implicit

explicit

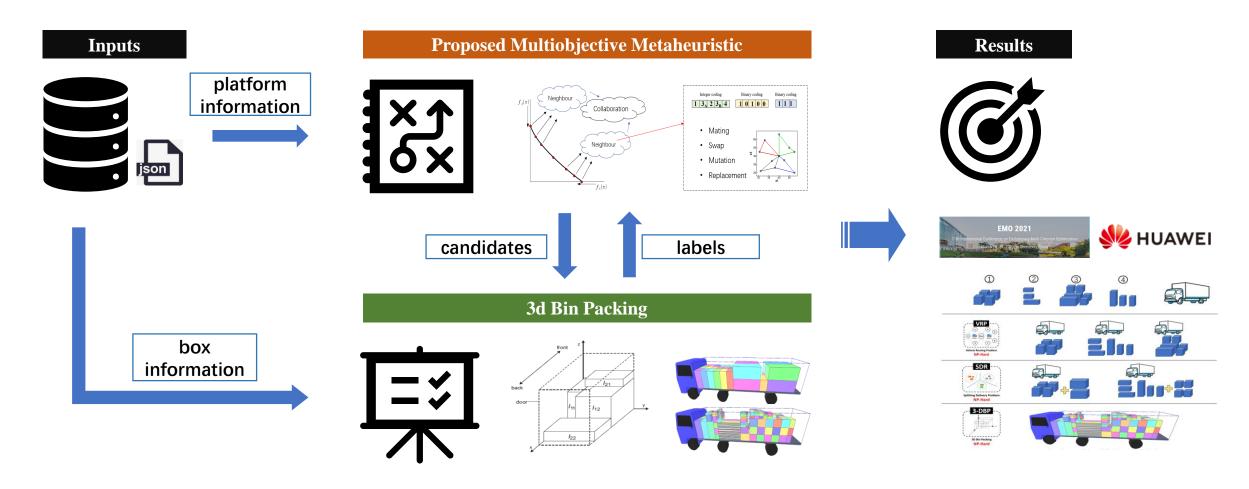
- Valid route constraints
- Single visit constraints
- Weight capacity constraints
- Bonded warehouse constraints

Packing constraints

- none-overlapping constraints
- None-splitting constraints
- Size constraints
- One-time loading constraints
- Supporting constraints

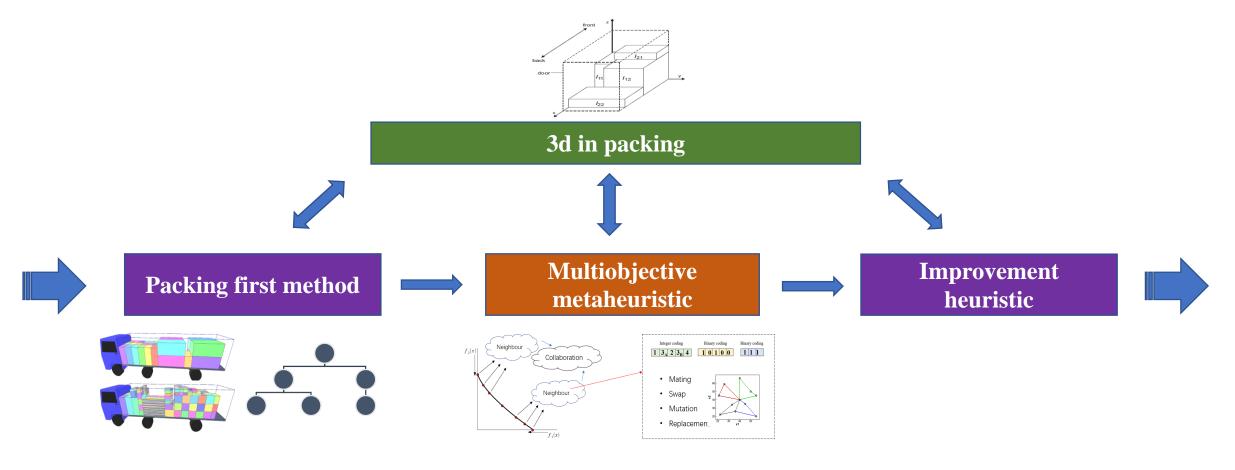


Solution for Huawei 3L-SDVRP





Final mixed version for Huawei 3L-SDVRP



https://www.noahlab.com.hk/logistics-ranking/#/ranking



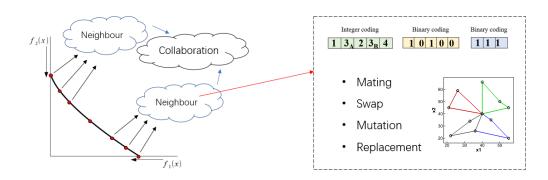
Outline

- 1. EMO2021 Competition Huawei 3L-SDVRP
- 2. A Metaheuristic Approach for 3L-SDVRP
- 3. Conclusion



Conclusion

- ➤ A metaheuristic approach is proposed for 3L-SDVRP
- The method is validated on a simple instance and the EMO competition
- Results show the flexibility of metaheuristic





Future works

- ➤ General encoding of design space & encoding of exploration and exploitation operators in metaheuristic
- Use offline & online knowledge in VRP and SDR
- Experiments on benchmark SDVRP instances, middle & large scale SDVRP, and compare with other methods
- **>** ...

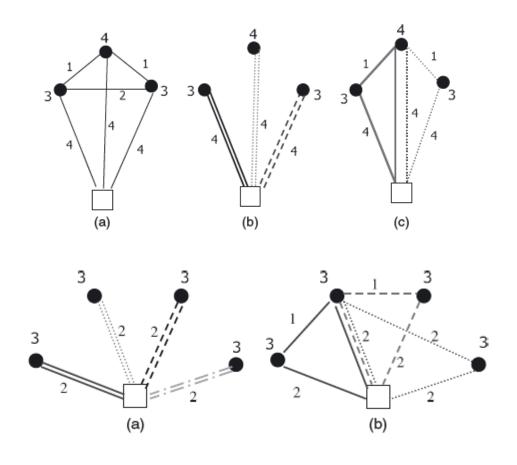


Thanks!

- Fei Liu, Qingling Zhu
- March 1, 2021
- Email: fliu36-c@my.cityu.edu.hk



Some conclusions on SDVRP



- The delivery cost reductions obtained when allowing split deliveries appear to be due primarily to the ability to <u>reduce</u> <u>the number of delivery routes</u>.
- The largest benefits are obtained when the mean demand is greater than half but less than three quarters of the vehicle capacity.
- The benefits from allowing split deliveries mainly depends on the relation between mean demand and vehicle capacity and on demand variance; there <u>does not appear to be a</u> <u>dependence on customer locations.</u>

- Archetti, Claudia, Martin WP Savelsbergh, and M. Grazia Speranza. "To split or not to split: That is the question." Transportation Research Part E: Logistics and Transportation Review 44.1 (2008): 114-123.
- Archetti, Claudia, and Maria Grazia Speranza. "Vehicle routing problems with split deliveries." International transactions in operational research 19.1-2 (2012): 3-22.



Some conclusions on SDVRP

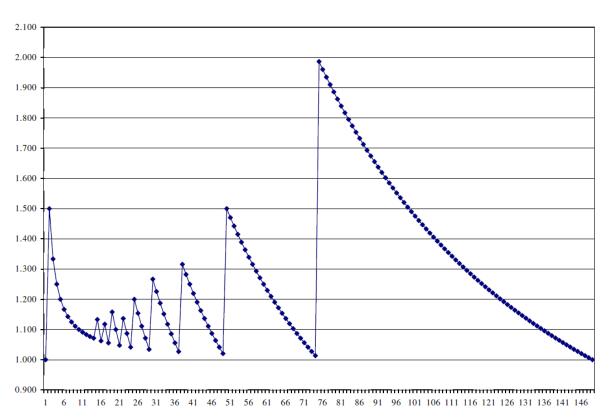
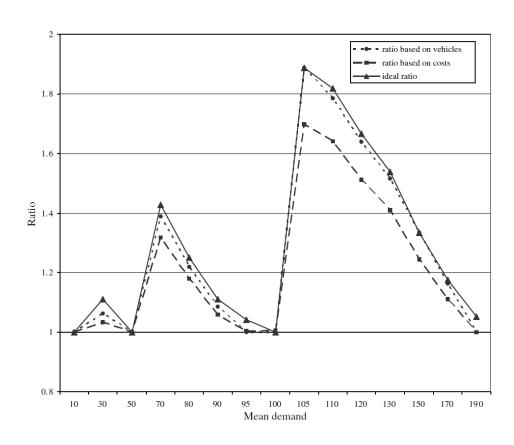


Fig. 1. Ratio $\frac{r(VRP)}{r(SDVRP)}$ as a function of d for an instance with 149 customers and vehicles with capacity 149.





Some examples on coding for VRP

coding

Natural number coding

For example: Chromosome: (1465327)Corresponding string:

(146513271)

crossover

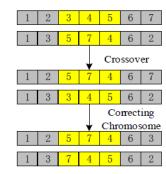


Figure 1. Chromosome Crossover

mutation

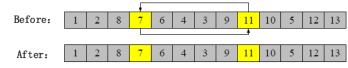


Figure 2. Chromosome Mutation

selection

copying the best individuals, weeding out the worst individuals.

Gao, Zhao, Guohua Sun, and Zhenglei Yuan. "Genetic Algorithm to the Split Delivery Vehicle Routing Problem." 2019 6th International Conference on Systems and Informatics (ICSAI). IEEE, 2019.