Continuous Approximation Model for Vehicle Routing -Tank Sizing Optimization

Fengqi You Elisabet Capon Ignacio E. Grossmann Jose M. Pinto





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Tank Sizing Problem

- New customers need new tanks
- All or some of existing customers subject to tank upgrades or downgrades
- Different trailer sizes and available tank sizes



Tank Sizing Issues

- Major Features
 - Key tradeoff: distribution (routing) vs. capital (tank sizing) cost
 - Capture the effects of customer synergies and tanker availability
- Nontrivial Problem
 - Non obvious ways of grouping customers for lower total costs
 - Tank size at a particular customer may influence distribution (routing) of all other customers
 - Many thousands of customer combinations are possible
 - Analyzing a small set of clusters at one time is useful
 - Lots of possible routes for each fixed tank sizing decision



Ex. a large customer may be delivered along with a small customer to empty tanker

Algorithm Flowchart for Detailed Model



Why we need continuous approximation?

- Detailed Integrated Model (DIM)
 - Solve vehicle routing and tank sizing simultaneously
 - Tradeoff routing and tank sizing cost directly
 - Pros: more accurate result
 - Cons: very large scale MILP, long CPU time
- Continuous Approximation Model (CAM)
 - Approximate the routing cost in the long run (e.g. annually)
 - Tradeoff the tank sizing cost with approximated routing cost
 - Pros: smaller model, fast computation



• Cons: total cost is approximated





"Cyclic" Inventory-Routing in CAM

- Key Assumption: each customer is replenished in a "cyclic" way with fixed interval *T*
- Required tank size \geq max. inv. = min. inv. + demand rate $\times T$



Routing & Replenishment in CAM

- T = R / (ave. speed)
 - *T* replenishment interval
 - *R* minimum distance to replenish all the customer in a cluster once
 - Average travelling speed is known
- If only one trip for each replenishment
 - R = TSP distance of the cluster & plant
- If allowing multiple trips for replenishment



• R = ?







CAM for Capacitated Routing Problems*

• Bounds for minimum routing distance **R**

$$\max\{2\frac{n}{q} \cdot r, \text{ TSP}\} \le R \le 2\lceil \frac{n}{q}\rceil r + (1 - \frac{1}{q}) \cdot \text{TSP}$$

- n # of customers in the cluster
- q capacity, max. # of customers that can be visited in one trip or volume in terms of # of customers with unit demand
- r average distance from customers to the plant
- TSP traveling salesman distance to visit all customers once
- Examples
 - Cluster 1: q=1, TSP=0, r = 67
 - $2\lceil \frac{n}{q} \rceil r + (1 \frac{1}{q}) \cdot \text{TSP} = 2r = 2 \times 67 = 134 \text{km}$
 - Cluster 2: q=1, same as Cluster 1, R = 4,400km

 $2\left\lceil \frac{n}{a}\right\rceil r + (1 - \frac{1}{a}) \cdot \text{TSP} = 2r + \frac{\text{TSP}}{2} = 2,225 \text{km}$

Cluster 2: q=2, TSP=50, r = 1,100





* M Haimovich, AHG Rinnooy Kan, "Bounds and heuristics for CRP", <u>Math. of Oper. Res.</u>, 1985, 10(4), 527-541 8

Algorithm Flowchart for CAM



Example: Comparison of DIM and CAM

- Problem Size
 - 3 customers (2) is new) in 2 clusters
 - 6 available tank size, 4 types of trucks
- Detailed Integrated Model (DIM)
 - CPU time: ~ 8 min. (5% gap)
 - 747 disc. Var., 1,606 cont. var., 2,121 constraints
 - Total cost: \$23,087
 - Upgrade 1 by 6,000 L, add a new one for 2 with 10,000 L, no change for 3
- Continuous Approximation Model (CAM)
 - CPU time: ~ 1 sec. for CAM, ~ 5 sec. for routing problem (5% gap)
 - 47 disc. Var., 26 cont. var., 35 constraints (CAM)



Total cost: \$23,405, **Same** tank sizing decisions



Conclusion / Future Work

- Conclusion
 - Continuous approximation model (CAM) for vehicle routing tank sizing problem
 - Novel algorithm framework to reduce the computational effort w/o too much sacrifice in solution quality
- Future Work
 - Simplify MINLP model to MILP model
 - Bi-level decomposition algorithm
 - Consider uncertainties such as demand variation and adding or losing customers

