## Non-Hamiltonian Formulations for the Single Vehicle Routing Problem with Deliveries and Selective Pickups

### Bruno P. Bruck Manuel Iori

DISMI, University of Modena and Reggio Emilia, Italy

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#### Motivation

### One-to-Many-to-One Transportation

We study the area of one-to-many-to-one single vehicle pickup and delivery problems (1-M-1 SVPDPs):

• A single vehicle laves the depot and delivers a first commodity to customers, it also collects a second commodity and brings it back to the depot



- Classical example: the vehicle delivers full drink bottles to customers and retrieves empty bottles to the depot
- 1-M-1 SVPDPs model many other real-world situations (distribution of electrical and electronic devices, courier service transportation, reverse logistics, ...)

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### General 1-M-1 SVPDPs Variants

Several variants are studied to better model the different real-world issues:

- Here we focus on the *general 1-M-1 SVPDPs* (Gribkovskaia and Laporte, 2008), where customers are not restrained to be visited once
- The most common general variants are the following ones



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### Single Vehicle with Deliveries and Selective Pickups

We study the Single VRP with Deliveries and Selective Pickups (SVDSP):

- we are given a graph G = (V, A), where  $V = \{0\} \cup P \cup D \cup PD$
- $P = \{ pickup cust. \}, D = \{ delivery cust. \}, PD = \{ combined cust. \}$
- each  $j \in D$  asks for a delivery of weight  $d_j$
- each  $j \in P$  offers a pickup of weight  $p_j$  and revenue  $r_j$
- each  $j \in PD$  has both delivery and pickup
- a vehicle of capacity Q
- a traveling cost  $c_{ij}$  is associated with each arc  $(i,j) \in A$

The SVDSP is to find a route that performs all deliveries and possibly some pickups, and minimizes the total cost

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### Single-demand vs Combined-demand

Two SVDSP cases are known in the literature:

- single-demand (SD) case contains only customers in P or in D
- combined-demand (CD) case may contain any type of customer

These two versions are, up to a certain extent, interchangeable:

- CD generalizes SD
- CD can be transformed into SD by *duplicating* combined customers



(a) selective pickups



(b) selective pickups - network of the duplicates

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### Prior Work

Probably all SVDSP algorithms in the literature refer to the network of the duplicates and solve the SD case. In terms of exact algorithms:

- Süral and Bookbinder (*Networks*, 2003): MILP formulation
- Gribkovskaia, Laporte and Shyshou (C&OR, 2008): MILP formulation + Tabu Search
- Gutiérrez-Jarpa, Marianov and Obreque (*IIE Transactions*, 2009): branch-and-cut algorithm

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Pros and cons of this idea:

- Advantage: optimal solutions are Hamiltonian, so many classical results from the literature can be reused
- Disadvantage: network might be doubled in size

In our work we consider instead *non-Hamiltonian* solutions on the original problem network

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## Two-commodity Hamiltonian (TCH) formulation

$$\begin{array}{ll} \min \ z_{TCH} = \sum_{(i,j) \in A'} c_{ij} x_{ij} + \sum_{j \in P'} r_j (1 - y_j) \\ \sum_{i \in V'} x_{ij} = 1 & \forall \ j \in D' \cup \{0\}, \\ \sum_{i \in V'} x_{ij} = y_j & \forall \ j \in P', \\ \sum_{i \in V'} (x_{ij} - x_{ji}) = 0 & \forall \ j \in V', \\ f_{ij}^d + f_{ij}^p \leq Q x_{ij} & \forall \ (i,j) \in A', \\ \sum_{i \in V'} \left( f_{ij}^d - f_{ji}^d \right) = d_j & \forall \ j \in V' \setminus \{0\}, \\ \sum_{i \in V'} \left( f_{ji}^p - f_{ij}^p \right) = p_j y_j & \forall \ j \in P', \\ \sum_{i \in V'} \left( f_{ji}^p - f_{ij}^p \right) = 0 & \forall \ j \in D', \\ x_{ij} \in \{0, 1\} & \forall \ (i,j) \in A', \\ y_j \in \{0, 1\} & \forall \ j \in P', \\ f_{ij}^d, f_{ij}^p \geq 0 & \forall \ (i,j) \in A'. \end{array}$$

- $V' = \{0\} \cup P' \cup D'$
- x<sub>ij</sub> = 1 if vehicle travels along arc (i, j), 0 otherwise
- y<sub>j</sub> = 1 if pickup of j is performed, 0 otherwise
- $f_{ij}^d$  = flow of delivery commodity
- $f_{ij}^{p}$  = flow of pickup commodity

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### *Two-commodity non-Hamiltonian* (TCNH) formulation

min $z_{TCNH} = \sum_{(i,j)\in A} c_{ij} x_{ij} +$	$\sum_{j \in P \cup PD} r_j(1-y_j)$
$\sum_{i \in V} x_{ij} = 1$	$\forall j \in D \cup \{0\},$
$\sum_{i\in V} x_{ij} = y_j$	$\forall \ j \in P,$
$\sum_{i \in V} x_{ij} \geq 1$	$\forall \ j \in \textit{PD},$
$\sum_{i \in V} x_{ij} \leq y_j + 1$	$\forall j \in PD,$
$\sum\nolimits_{i \in V} (x_{ij} - x_{ji}) = 0$	$\forall j \in V,$
$f_{ij}^d + f_{ij}^p \leq Q x_{ij}$	$\forall (i,j) \in A,$
$\sum\nolimits_{i \in V} \left( f_{ij}^d - f_{ji}^d \right) = d_j$	$\forall j \in V \setminus \{0\},$
$\sum\nolimits_{i \in V} \left( f_{ji}^{p} - f_{ij}^{p} \right) = p_{j} y_{j}$	$\forall j \in P \cup PD,$
$\sum_{i\in V} \left( f_{ji}^{p} - f_{ij}^{p} \right) = 0$	$\forall \ j \in D,$
$x_{ij} \in \{0,1\}$	$\forall (i,j) \in A \setminus A(PD),$
$x_{ij} \in \{0,1,2\}$	$\forall (i,j) \in A(PD),$
$y_j \in \{0,1\}$	$\forall j \in P \cup PD,$
$f_{ij}^d, f_{ij}^p \geq 0$	$\forall (i,j) \in A.$

• 
$$V = \{0\} \cup P \cup D \cup PD$$

- $x_{ii}$  = number of times that vehicle travels along arc (i, j)
- $y_i = 1$  if pickup of j is performed, 0 otherwise
- $f_{ii}^d$  = flow of delivery commodity
- $f_{ii}^{p} =$ flow of pickup commodity

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### Some Properties

#### Property 1

For the SD case, TCNH is equivalent to TCH and thus provides the optimal SVDSP solution value

Trivially checked by setting  $PD = \emptyset$ 

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Trivially checked by setting  $PD = \emptyset$ 

#### Property 2

For the CD case,  $z_{TCNH} \leq z_{TCH}$  and thus TCNH provides a relaxation of the SVDSP

Proven by showing that any TCH solution is mapped into a TCNH one, but the opposite does not hold

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### Solutions with Split Deliveries

TCNH accepts solutions where pickups and/or deliveries are *split* in the two visits to combined customers



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A TCNH solution with split deliveries or pickups can be transformed into a solution having the same cost and for which no delivery or pickup is split

Done with a path-search algorithm (that might take exponential time!)

### Solutions with Temporary Dropoffs

TCNH also accepts solutions where some part of the load is temporary dropped off in the first visit to a customer, and then recollected during the second visit



No nice property can help us here, so we deal with dropoffs with tailored *branch-and-cut* (B&Cut) algorithms

### Benders' Decomposition

We first obtain a *Benders' Based non-Hamiltonian* (BBNH) formulation, by projecting out from TCNH the two-commodity flow variables:

$$\begin{split} f_{ij}^{d} + f_{ij}^{p} &\leq Q x_{ij} & \forall (i,j) \in A, \\ \sum_{i \in V} \left( f_{ij}^{d} - f_{ji}^{d} \right) &= d_{j} & \forall j \in V \setminus \{0\}, \\ \sum_{i \in V} \left( f_{ji}^{p} - f_{ij}^{p} \right) &= p_{j} y_{j} & \forall j \in P \cup PD, \\ \sum_{i \in V} \left( f_{ji}^{p} - f_{ij}^{p} \right) &= 0 & \forall j \in D, \\ f_{ij}^{d}, f_{ij}^{p} &\geq 0 & \forall (i,j) \in A. \end{split}$$

When a solution is found for the "difficult" (x, y) variables, we solve the dual of the above linear subproblem (DSP). If unbounded, we obtain the following *Benders' feasibility cut*:

$$\sum_{(i,j)\in A} (Qt_{ij}^r) x_{ij} + \sum_{j\in V\setminus\{0\}} d_j v_j^r + \sum_{j\in P\cup PD} (p_j w_j^r) y_j \leq 0 \qquad \forall r\in R,$$

where R denotes the set of the extreme rays of the DSP

### Cutting Planes Generation

Benders' cuts are weak, so we tried to improve them (but failed). However, we found additional families of valid inequalities, and provided for them polynomial-time separation procedures (max-flow). We solved the resulting *Two-Index Non-Hamiltonian* (TINH) formulation by B&Cut

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There are several ways to implement a "modern" B&Cut:

- separate at fractional nodes (classical)
- separate at integer nodes (Subramanian et al., OR Letters, 2011)
- separate outside the MILP (Pferschy and Staněk, tech. rep., 2014)

Our best B&Cut for TINH:

- separate Benders' cuts only when new incumbent is found
- separate all other cuts at integer nodes

#### Faster Relaxations

## An example of a Valid Cut

#### Property 4

The following capacity constraint is valid for the SDSP:

$$\begin{aligned} Qx(\bar{S}:S) - \sum_{j \in S} p_j y_j \geq d(V) - d(S) - Q \quad \forall S \subseteq V \setminus \{0\} : p(S) - d(S) > Q - d(V) \\ \text{where } \bar{S} = V \setminus S \setminus \{0\}, \ d(S) = \sum_{j \in S} d_j, \ p(S) = \sum_{j \in S} p_j \end{aligned}$$

#### $\bar{x}_{ik}$ $\frac{d_j}{Q} + \frac{p_j(1-\bar{y}_j)}{O}$ **Property 5** $\bar{x}_{ki}$ $\frac{p_k(1-\bar{y}_k)}{O}$ $\bar{x}_{ji} + \frac{p_j}{O}$ Capacity constraints $\frac{d_i}{O} + \frac{p_i(1-\bar{y}_i)}{O}$ $\mathbf{n}+$ i can be separated in polynomial time $\bar{x}_{ji} + \frac{p_j}{O}$ $\frac{p_i(1-\bar{y}_j)}{O}$ $\frac{d_j}{Q}$ э

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### B&Cut algorithms for the CD case

The B&Cut for TINH solves to optimality the SD case, but only provides a relaxation for the CD case. We used it as a basis for CD exact algorithms:

- Throw-Away (TA): inside B&Cut for TINH, when a new incumbent is found, check if it is feasible (i.e., it has no dropoffs). If not, simply disregard solution
- 2-Steps (2S): first invoke B&Cut for TINH. If final solution is not feasible, invoke TA (but initialize it with all cuts already generated)
- Minimal Network of the Duplicates (MND): invoke B&Cut for TINH but disregard Benders' cuts. Check if capacity or dropoff are violated in final solution. If so, duplicate the CD customer originating the violation and re-iterate (inserting all generated cuts)

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### Single-Demand Case

Algorithms coded in C++, all tests run with Cplex 12.6, i7 3.4 GHz, 3600 sec t.lim.

			litera	ature		new						
			SB	G	MO*	TCNH		BBNH		TINH		
SB benchmark	#	opt	sec	opt	sec	opt	sec	opt	sec	opt	sec	
( V =10)	24	24	1.8	24	0.0	24	0.1	24	0.0	24	0.0	
( V =20)	21	18	461.2	21	0.2	21	0.5	21	0.5	21	0.0	
( V =30)	18	14	823.2	18	0.6	18	2.3	18	3.1	18	0.1	
SB All	63	56	389.6	63	0.2	63	0.9	63	1.1	63	0.0	
GMO benchmark	#	opt	sec	opt	sec	opt	sec	opt	sec	opt	sec	
$(25 \le  V  \le 30)$	18	18	32.3	18	2.2	18	1.5	18	4.8	18	0.1	
$(40 \le  V  \le 60)$	39	21	2441.8	39	47.0	39	44.1	36	466.4	39	3.6	
(68 $\leq$ $ V  \leq$ 90)	17	3	3898.4	16	3510.8	17	314.6	6	2513.0	17	37.3	
GMO All	74	42	2190.4	73	831.9	74	95.9	60	824.3	74	10.5	

- SB = Süral and Bookbinder (2003)
- GMO = Gutiérrez-Jarpa, Marianov and Obreque (2009), run with with Cplex 10, Dual Core AMD 2.7 GHz, 21000 sec t.lim.

### **Combined-Demand Case**

Algorithms coded in C++, all tests run with Cplex 12.6, i7 3.4 GHz, 3600 sec t.lim.

		Hamiltonian							non-Hamiltonian						
		GLS		SB		ТСН		TA		25		MND			
GLS benchmark	#	opt	sec	opt	sec	opt	sec	opt	sec	opt	sec	opt	sec		
(15 <  V  < 30)	28	1	3487	5	2961	23	915	28	0.4	28	0.4	28	0.2		
$(32 \leq  V  \leq 50)$	24	0	3600	0	3600	1	3592	21	837.3	19	833.2	24	5.2		
$(71 \le  V  \le 100)$	16	0	3600	0	3600	0	3600	10	1412.1	10	1369.0	<sup>(*)</sup> 15	375.5		
All GLS	68	1	3553	5	3336	24	2492	59	628.0	57	616.4	67	90.3		

- GLS = Gribkovskaia, Laporte and Shyshou (2008)
- SB = Süral and Bookbinder (2003)
- \* = remaining instance solved to proven optimality in about 4.7 CPU hours

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### Conclusions

We developed new formulations that exploit the original non-Hamiltonian structure of the problem and provided a few theoretical properties

By developing several B&Cut algorithms we found good computational results, especially for the more general combined-demand case

Solutions are frequently non-Hamiltonian. Temporary dropoffs, if allowed, are frequent

Ideas can be extended to several other general PD problems:

- Mixed deliveries
- Backhaul deliveries
- Single vehicle problems with dropoff
- Multiple vehicle problems (with and without transhipment)

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# Thank you very much for your attention Comments and questions are welcome