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

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- Faster Relaxations
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## 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, ...)


## 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

(a) backhaul deliveries

(b) mixed deliveries

(c) selective pickups

Selective pickups case can lead to lowest costs and can generalize the other cases (Golden\&Assad, Oper. Res., 1986)

## 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 $V=\{0\} \cup P \cup D \cup P D$
- $P=\{$ pickup cust. $\}, D=\{$ delivery cust. $\}, P D=\{$ 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 $j \in P D$ has both delivery and pickup
- a vehicle of capacity $Q$
- a traveling cost $c_{i j}$ 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

## 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

$$
D^{\prime}=D \cup\left\{j^{D} \forall j \in P D\right\}
$$


(b) selective pickups - network of the duplicates

## 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

## Two-commodity Hamiltonian (TCH) formulation

$$
\begin{array}{lrl}
\min z_{T C H}=\sum_{(i, j) \in A^{\prime}} c_{i j} x_{i j}+\sum_{j \in P^{\prime}} r_{j}\left(1-y_{j}\right) \\
\forall j \in D^{\prime} \cup\{0\}, & \\
\sum_{i \in V^{\prime}} x_{i j}=1 & \forall j \in P^{\prime}, & \bullet V^{\prime}=\{0\} \cup P^{\prime} \cup D^{\prime} \\
\sum_{i \in V^{\prime}} x_{i j}=y_{j} & \forall j \in V^{\prime}, & \begin{array}{l}
x_{i j}=1 \text { if vehicle travels } \\
\text { along arc }(i, j), 0
\end{array} \\
\sum_{i \in V^{\prime}}\left(x_{i j}-x_{j i}\right)=0 & \forall(i, j) \in A^{\prime}, & \begin{array}{l}
\text { otherwise } \\
y_{j}=1 \text { if pickup of } j \text { is }
\end{array} \\
f_{i j}^{d}+f_{i j}^{p} \leq Q x_{i j} & \forall j \in V^{\prime} \backslash\{0\}, & \begin{array}{l}
\text { performed, } 0 \text { otherwise }
\end{array} \\
\sum_{i \in V^{\prime}}\left(f_{i j}^{d}-f_{j i}^{d}\right)=d_{j} & \forall j \in P^{\prime}, & \begin{array}{l}
f_{i j}^{d}=\text { flow of delivery } \\
\text { commodity }
\end{array} \\
\sum_{i \in V^{\prime}}\left(f_{j i}^{p}-f_{i j}^{p}\right)=p_{j} y_{j} & \forall j \in D^{\prime}, & \bullet f_{i j}^{p}=\text { flow of pickup } \\
\sum_{i \in V^{\prime}}\left(f_{j i}^{p}-f_{i j}^{p}\right)=0 & \forall(i, j) \in A^{\prime}, & \text { commodity }
\end{array}
$$

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

$$
\begin{array}{lr}
\min z_{T C N H}=\sum_{(i, j) \in A} c_{i j} x_{i j}+\sum_{j \in P \cup P D} r_{j}\left(1-y_{j}\right) \\
\sum_{i \in V} x_{i j}=1 & \forall j \in D \cup\{0 \\
\sum_{i \in V} x_{i j}=y_{j} & \forall j \in I \\
\sum_{i \in V} x_{i j} \geq 1 & \forall j \in P L \\
\sum_{i \in V} x_{i j} \leq y_{j}+1 & \forall j \in P L \\
\sum_{i \in V}\left(x_{i j}-x_{j i}\right)=0 & \forall j \in l \\
f_{i j}^{d}+f_{i j}^{p} \leq Q x_{i j} & \forall(i, j) \in 1 \\
\sum_{i \in V}\left(f_{i j}^{d}-f_{j i}^{d}\right)=d_{j} & \forall j \in V \backslash\{0 \\
\sum_{i \in V}\left(f_{j i}^{p}-f_{i j}^{p}\right)=p_{j} y_{j} & \forall j \in P \cup P L \\
\sum_{i \in V}\left(f_{j i}^{p}-f_{i j}^{p}\right)=0 & \forall j \in L \\
x_{i j} \in\{0,1\} & \forall(i, j) \in A \backslash A(P D \\
x_{i j} \in\{0,1,2\} & \forall(i, j) \in A(P D \\
y_{j} \in\{0,1\} & \forall j \in P \cup P L \\
f_{i j}^{d}, f_{i j}^{p} \geq 0 & \forall(i, j) \in l
\end{array}
$$

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


## 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 $P D=\emptyset$

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

Trivially checked by setting $P D=\emptyset$
Property 2
For the $C D$ case, $z_{T C N H} \leq z_{T C H}$ and thus $T C N H$ 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

## Solutions with Split Deliveries

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

(a) pickup $p_{1}$ is split in two visits

(b) no split of $p_{1}$

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## Property 3

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

(a) dropoff of 2 units of load in vertex 1

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:

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

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:

$$
\sum_{(i, j) \in A}\left(Q t_{i j}^{r}\right) x_{i j}+\sum_{j \in V \backslash\{0\}} d_{j} v_{j}^{r}+\sum_{j \in P \cup P D}\left(p_{j} w_{j}^{r}\right) y_{j} \leq 0 \quad \forall r \in R,
$$

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


## An example of a Valid Cut

## Property 4

The following capacity constraint is valid for the SDSP:

$$
\begin{aligned}
& Q \times(\bar{S}: S)-\sum_{j \in S} p_{j} y_{j} \geq d(V)-d(S)-Q \quad \forall S \subseteq V \backslash\{0\}: p(S)-d(S)>Q-d(V) \\
& \text { where } \bar{S}=V \backslash S \backslash\{0\}, d(S)=\sum_{j \in S} d_{j}, p(S)=\sum_{j \in S} p_{j}
\end{aligned}
$$

## Property 5

Capacity constraints can be separated in polynomial time


## 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:
(1) 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) 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)
(3) 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)

## Single-Demand Case

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

| SB benchmark | \# | literature |  |  |  | new |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | SB |  | GMO* |  | TCNH |  | BBNH |  | TINH |  |
|  |  | 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 |
| $(V \mid=20)$ | 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 |
| $(25 \leq\|V\| \leq 30)$ | 18 | 18 | 32.3 | 18 | 2.2 | 18 | 1.5 | 18 | 4.8 | 18 | 0.1 |
| $(40 \leq\|V\| \leq 60)$ | 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 |

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


## Combined-Demand Case

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

| GLS benchmark |  | Hamiltonian |  |  |  |  |  | non-Hamiltonian |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | GLS |  | SB |  | TCH |  | TA |  | 2S |  | MND |  |
|  | \# |  | sec |  | sec | opt | sec | opt |  |  | sec | opt | sec |
| $(15 \leq\|V\| \leq 30)$ | 28 |  | 3487 |  | 2961 | 23 | 915 | 28 | 0.4 | 28 | 0.4 | 28 | 0.2 |
| $(32 \leq\|V\| \leq 50)$ | 24 | 0 | 3600 |  | 3600 | 1 | 3592 | 21 | 837.3 | 19 | 833.2 | 24 | 5.2 |
| $(71 \leq\|V\| \leq 100)$ | 16 | 0 | 3600 |  | 3600 |  | 3600 | 10 | 1412.1 | 10 | 1369.0 | ${ }^{*} 15$ | 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


## 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)


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

