# A branch-and-price method for the pickup and delivery problem with truck driver scheduling 

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## Outline

- Motivation
- Problem definition
- Solution Methodology
- Computational Results
- Summary


## Haste AS

- Start up company
- Portal for planning freight orders
- Market place for transporters and shippers
- Targeting Norwegian/ Scandinavian market



## Haste AS

- Main goal: Developed a Heuristic
- ALNS
- Need to benchmark this
- Exact CG method


Problem definition

## Problem definition



0

- A set of transportation requests with given:
- pickup and delivery locations
- weight/Volume
- Revenue (if transported)


## Problem definition



0

- A set of vehicles with:
- Individual starting location (no depot)
- Open ended
- All vehicles are identical

- Time windows at pickup and delivery


## Problem definition



- Pickup and Delivery Problem with time windows (PDPTW)
- No depot
- Open ended
- Only optional requests


## Relevant literature on PDPTW- CG stuff only

- Dumas et al (1991) - First BP alg.
- Røpke and Cordeau (2009) - First BPC alg.
- Baldacci et al. (2011) - Route enumerations
- Gschwind et al (2018) - Bi-directional
- Homsi et al (2020) - Ship routing and scheduling


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- Breaks


## Problem definition



- Pickup and Delivery Problem with time windows (PDPTW)
- No depot
- Open ended
- Only optional requests
- Breaks
- Rests


## European Hours of Service regulations

|  | Break $-\mathbf{4 5} \mathrm{min}$ | Rest -11 hrs | Weekly rest |
| :--- | :--- | :--- | :--- |
| Drive | 4.5 hrs | 9 hrs | 56 hrs |
| Work | 6.0 hrs | 13 hrs | 60 hrs |

## Relevant literature - CG stuff only

- VRP with Truck driver scheduling
- Goel and Irnich (2017) - First BP alg.
- Tilk and Goel (2020) - Bi-directional


## Solution method

Master problem

$$
\max z=\sum_{v \in \in V \in \sum_{v}} \sum_{v o r} P_{v r r}
$$

$$
\sum_{v \in \in \in \in \in \mathcal{R}_{v}} A_{i v i} A_{v r} \leq 1,
$$

$$
i \in N^{P}
$$

$$
\sum_{r \in \mathcal{R}_{v}} \lambda_{v r}=1
$$

$$
v \in \mathcal{V}
$$

$$
\lambda_{v r} \in\{0,1\},
$$

$$
\forall v \in \mathcal{V}, r \in \mathcal{R}_{v}
$$

## Subproblem

- Resource constrained shortest path problem
- Combination of:
- Røpke and Cordeau (2009)
- Goel and Irnich (2017)
- With some modifications


## One subproblem pr vehicle

- Vehicle 1:
- Vehicle 2:
- Vehicle n :



## One subproblem pr vehicle

- Vehicle 1:
- Vehicle 2:
- Vehicle n :



## One subproblem pr vehicle

- Vehicle 1:
- Vehicle 2 :
- Vehicle n :



## One subproblem pr vehicle

- Vehicle 1:
- Vehicle n :



## Resources needed for the PDPTW

## - Based on Røpke and Cordeau (2009)

| Resource | Resource Description | Resource Window |
| :--- | :--- | :--- |
| $\bar{p}$ | Accumulated reduced cost after visiting node $i$ | $[-\infty, \infty]$ |
| $l^{W}$ | Load of the vehicle after visiting node $i$ in terms of weight | $\left[0, W^{C}\right]$ |
| $l^{V}$ | Load of the vehicle after visiting node $i$ in terms of volume | $\left[0, V^{C}\right]$ |
| $t^{t i m e}$ | Time elapsed since start of route | $\left[\underline{T_{i}}, \overline{T_{i}}\right]$ |
| $\mathcal{U}$ | Set of unreachable nodes on the route | $\mathcal{U} \subseteq N^{P}$ |
| $\mathcal{O}$ | Set of requests started but not completed on this route | $\mathcal{O} \subseteq N^{P}$ |

## Resources needed for TDS

## - Based on Goel and Irnich (2017)

| Resource | Resource Description |
| :--- | :--- |
| $t^{\text {dist }}$ | Remaining driving time to the next node, $j$ |
| $t^{\text {drive } \mid R}$ | Accumulated driving time since the end of the last rest |
| $t^{\text {elapsed } \mid R}$ | Time elapsed since the end of the last rest |
| $t^{\text {lates } \mid R}$ | Latest time for when a rest must end |
| $t^{\text {drive } \mid B}$ | Accumulated driving time since the last break or rest |
| $t^{\text {elapsed } \mid B}$ | Time elapsed since the end of the last break or rest |
| $t^{\text {latest } \mid B}$ | Latest time for when a break must end |
| $t^{\text {drive } \mid W}$ | Total accumulated weekly driving time |
| $t^{\text {elapsed } \mid W}$ | Time elapsed since the end of the last weekly rest |
| $t^{\text {latest } \mid W}$ | Latest time for when a weekly rest must end |

Resource Window
$[0,0]$
$\left[0, T^{\text {drive } \mid R}\right]$
$\left[0, T^{\text {elapsed } \mid R}\right]$
$[0, \infty]$
$\left[0, T^{\text {drive } \mid B}\right]$
$\left[0, T^{\text {elapsed } \mid B}\right]$
$[0, \infty]$
$\left[0, T^{\text {drive } \mid W}\right]$
$\left[0, T^{\text {elapsed } \mid W}\right]$
$[0, \infty]$

## Network modification for the TDS



Figure 1: Auxiliary Network proposed by Goel and Irnich (2017). The network describes the possible extensions of a label by traversing the arc $(i, j)$.

## Resource extension functions



## Improvements

- Combining existing methods work, but can we improve?
- Lots of resources, and nodes gives lots of labels
- What can we do:
- Relax the subproblem
- Discard labels earlier
- Strengthen dominance


## Relaxing the subproblem

- Solve labeling alg. without «break»-resources
- Only affects feasibility, not optimality.
- Three cases:
- Finds no routes - > RMP is optimal
- Finds at least one feasible route with positive reduced cost -> new CG iteration
- Finds at least one route, but none are feasible -> solve full SP



## Discarding Labels

$$
j+n
$$



$$
\begin{gathered}
L=\left(i, L^{-}, T\right) \\
O(L)=\{j, k, l\}
\end{gathered}
$$


$2 n+1$

## Discarding Labels

$$
j+n
$$



What is the latest time we can leave $i$, given that we have to visit all these nodes?

## Discarding Labels

## 0

$$
\begin{gathered}
L=\left(i, L^{-}, T\right) \\
O(L)=\{j, k, l\}
\end{gathered}
$$

Determine $t_{i}^{\text {Late }}(W)$, where $W \subset N^{P}$

## Discarding Labels



This can be solved as a backward labeling, given one additional resource to ensure all nodes are visited

## Discarding Labels



This can be solved as a backward labeling, given one additional resource to ensure all nodes are visited If $t^{\text {Time }}(L)>t_{i}^{\text {Late }}(O(L))$ we may discard L

## Increasing the unreachable set



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## Increasing the unreachable set



## Increasing the unreachable set



## Increasing the unreachable set



Determine $t_{i}^{\text {Late }}(l, W)$, where $l \in N^{P}, W \subset N^{P}$
This can be solved as a backward labeling, ensuring all nodes are visited and presedence for $l, l+n$
$U(L)=U(L) \cup\left\{l \in N^{P} \mid t^{\text {Time }}(L)>t_{i}^{\text {Late }}(l, O(L))\right\}$

## Preprocessing of $t^{\text {Late }}$

- $\forall i \in N, W \subset N^{P},|W| \leq 3$, calculate $t_{i}^{\text {Late }}(W)$
- $\forall i \in N, l \in N^{P}, W \subset N^{P},|W| \leq 2$, calculate $t_{i}^{\text {Late }}(l, W)$
- In both cases we omit the break resources


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- $\forall i \in N, l \in N^{P}, W \subset N^{P},|W| \leq 2$, calculate $t_{i}^{\text {Late }}(l, W)$
- In both cases we omit the break resources
- Note that if

$$
\begin{aligned}
& \text { - } \exists W \subseteq O(L), t^{\text {Time }}(L)>t_{i}^{\text {Late }}(W) \text {, we can discard } \mathrm{L} \\
& -U(L)=U(L) \cup\left\{l \in N^{P} \mid \exists W \subseteq O(L), t^{\text {Time }}(L)>t_{i}^{\text {Late }}(l, W)\right\}
\end{aligned}
$$

## Computational Results

## Test instances

- 132 locations from central and southern Norway
- Distances and times based on Google Maps
- Probability of drawing each location proportional to population
- Planning horizon of 144 hours
- Three time windows widths: 12-24, 24-48, 48-144 hours
- Two cargo sizes: 1-10, 10-20 (capacity of vehicle 30 )
- \# requests $=10,15,20,25,50,75,100,150,200$
- \# Vehicles $=\left\lfloor\frac{\# \text { requests }}{4}\right\rfloor,\left\lfloor\frac{\# \text { requests }}{5}\right],\left\lfloor\frac{\# \text { requests }}{6}\right\rfloor$
- Four instances of each setting gives 648 instances total



## Effect of the preprocessing

Table 3: The number of instances with a given number of requests where the BP method can find an optimal solution (\# opt.) and a dual bound (\# DB), and the average computing time (Avg. time) used, with and without the preprocessing techniques from Section 4.5.

|  | Without preprocess |  |  | With preprocess |  |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: |
| \# requests | \# opt. | \# DB | Avg. time | \# opt. | \# DB | Avg. time |
| $\mathbf{1 0}$ | 63 | 63 | 1218.79 | 66 | 68 | 631.99 |
| $\mathbf{1 5}$ | 46 | 47 | 3063.39 | 60 | 60 | 1217.78 |
| $\mathbf{2 0}$ | 28 | 29 | 4575.96 | 58 | 60 | 1679.74 |
| $\mathbf{2 5}$ | 23 | 24 | 4989.11 | 58 | 60 | 1801.29 |
| $\mathbf{5 0}$ | 6 | 6 | 6873.35 | 31 | 37 | 4357.43 |
| Total | 166 | 169 | 4144.12 | 273 | 285 | 1937.64 |

## Effect of preprocessing

Table 4: The number of instances with a given time window width where the BP method can find an optimal solution (\# opt.) and a dual bound (\# DB), and the average computing time (Avg. time) used, with and without the preprocessing techniques from Section 4.5 .

|  | Without preprocess |  |  | With preprocess |  |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: |
| Time Windows | \# opt. | \# DB | Avg. time | \# opt. | \# DB | Avg. time |
| $12-24$ | 81 | 81 | 2695.85 | 120 | 120 | 40.23 |
| $24-48$ | 61 | 63 | 3775.15 | 103 | 109 | 1236.03 |
| $48-144$ | 24 | 25 | 5961.36 | 50 | 56 | 4536.68 |

## Effect of preprocessing on PDPTW

- Tested on the instances propoced by Røpke and Cordeau (2009)

|  | without preprocess | with preprocess |
| :--- | :---: | :---: |
| AA | 637.7 | 591.9 |
| BB | 841.1 | 785.4 |
| CC | 3047.8 | 3038.9 |
| DD | 3124.0 | 3040.7 |

## Summary

- Presented new problem (and acronym) to the research community - PDPTDS
- Preprocesing of time windows based on the open set can reduce computational time significantly
- Does this carry over to the bi-directional case?
- Instances, results and preprint of paper found:
- http://axiomresearchproject.com/publications

