

Planning Shared Passenger and Freight Transport on a Fixed Infrastructure

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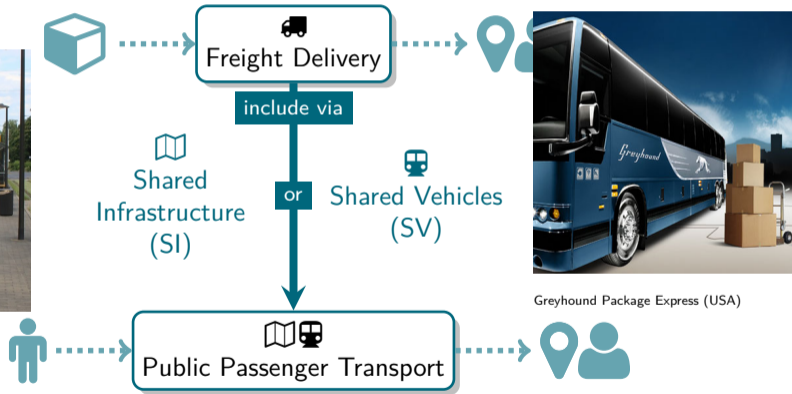
Last-Mile Delivery Workshop, Vallendar

27.06.2022

What is shared passenger and freight transport?



CargoTram in Dresden, Germany (2001-2020)



Greyhound Package Express (USA)

LastMileTram, Frankfurt (Main)

Start: 2018

<http://www.relut.de>

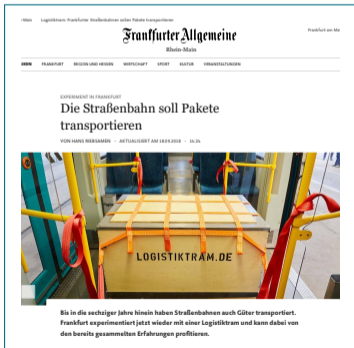
LogIKTram, Karlsruhe

Start: 2021

www.logiktram.de

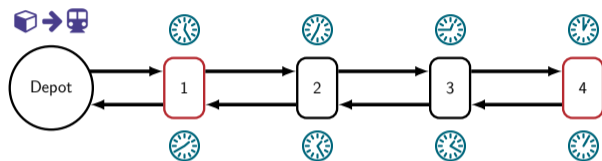
Cargo-Tram, Berlin

Start: 2021

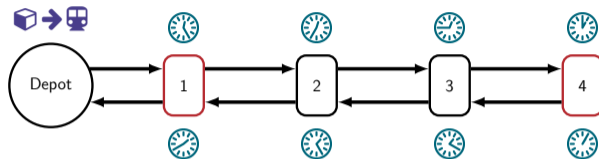




How can we efficiently schedule passengers and freight?



- ▶ Circular bi-directional single route
- ▶ Given passenger demand as traffic flow (arrival and alighting rate)
- ▶ Given freight demand, i.e., freight containers with
 - individual release time
 - soft (and hard) deadline
 - destination stop



Aperiodic train schedule

- ▶ Departure $d_{v,s} \geq 0$ of vehicle v at stop s
- ▶ Arrival $a_{v,s} \geq 0$ of vehicle v at stop s
- ▶ Dwell time $h_{v,s} \geq 0$ of vehicle v at stop s
- ▶ Is vehicle v operating ($o_v = 1$); or not ($o_v = 0$)?



Freight allocation

- ▶ Assign individual freight requests to vehicles
- ▶ Is request r loaded to vehicle v at stop s ($y_{r,v,s} = 1$); or not ($y_{r,v,s} = 0$)?
- ▶ Accept ($u_r = 1$) or reject ($u_r = 0$) request r ?

Linear program: Lexicographical objective

\mathcal{S} : stops, \mathcal{P} : periods, \mathcal{R} : freight requests, P : rejection penalty

$$\text{lex min } \left\{ \sum_{s \in \mathcal{S}} \sum_{p \in \mathcal{P}} \frac{w_{s,p}}{|\mathcal{S}| \cdot |\mathcal{P}|}, \sum_{r \in \mathcal{R}} \frac{t_r}{|\mathcal{R}|} + P \cdot (1 - u_r) \right\} \quad (1)$$



1st objective:
Passengers have priority!

- ▶ Minimize number of waiting passengers w_{sp} across stops s and periods p



2nd objective:
Reduce freight delay

- ▶ Minimize delay t_r of freight request r
- ▶ Minimise the number of rejected requests ($u_r = 0$)

Constraints: Headway times, dwell times, service times, capacity, etc.



Vehicle capacity



©Kay Dreyer, HfG Offenbach, 2019



Dwell time



©J. Schwarze, HfG Offenbach



Station capacity



©Porstner u. Qu, 2019

Image sources: Schocke et al., 2020

periods	stops	containers	time (sec.)	1st objective			2nd objective		
				sec.	value	gap	sec.	value	gap
60	14	0	2091	2091	5.59	0	0	0	0
60	14	10	1604	1554	5.59	0	49	0	0
60	14	20	2522	2502	5.59	0	20	0	0
60	14	30	35067	2658	5.59	0	32400*	0.1	1
60	14	40	4444	2825	5.59	0	1619	0.08	0
60	14	50	2117	2048	5.59	0	69	50000.24	0
180	28	0	32482	32471*	74.09	0.96	1	0	0
180	28	10	36147	32476*	43.7	0.93	3670	0	0
180	28	20	38774	32618*	49.64	0.94	6156	0	0
180	28	30	36065	32618*	45.54	0.93	3456	0	0
180	28	40	36106	32618*	51.96	0.94	3628	0	0
180	28	50	37844	32618*	26.04	0.81	5390	0	0

*runtime-limit exceeded

► Definition by Pisinger and Ropke, 2010

Algorithm 2 Adaptive large neighborhood search

```

1: input: a feasible solution  $x$ 
2:  $x^b = x$ ;  $\rho^- = (1, \dots, 1)$ ;  $\rho^+ = (1, \dots, 1)$ ;
3: repeat
4:   select destroy and repair methods  $d \in \Omega^-$  and  $r \in \Omega^+$  using  $\rho^-$  and  $\rho^+$ ;
5:    $x^r = r(d(x))$ ;
6:   if  $\text{accept}(x^r, x)$  then
7:      $x = x^r$ ;
8:   end if
9:   if  $c(x^r) < c(x^b)$  then
10:     $x^b = x^r$ ;
11:  end if
12:  update  $\rho^-$  and  $\rho^+$ ;
13: until stop criterion is met
14: return  $x^b$ 

```

► Work of Yin et al., 2021

- ALNS for demand-oriented timetabling problem
- Each neighbor represents a timetable

► This work's extension

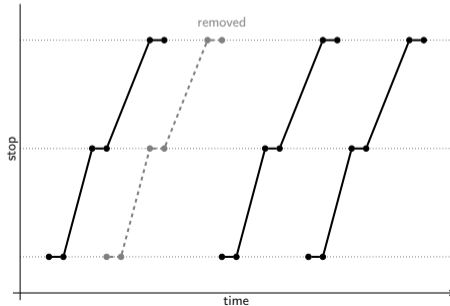
- Addition of cargo favouring repair operators
- Inclusion of cargo allocation problem (solved with Gurobi)

Adaptive Large Neighborhood Search

Destroy Operators (Yin et al., 2021)

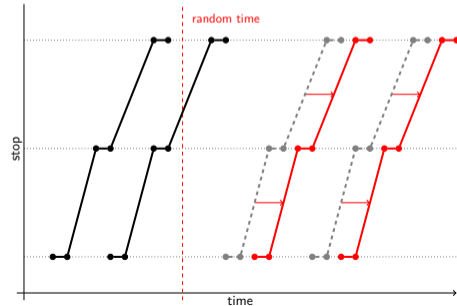
Random deletion

- ▶ Randomly select a vehicle
- ▶ Remove selected vehicle



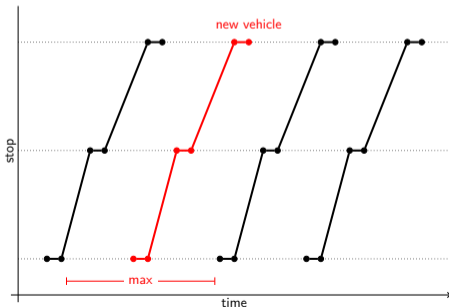
Random shift

- ▶ Randomly select time, direction of shift and shift interval
- ▶ Shift all vehicles departure before/after the selected time



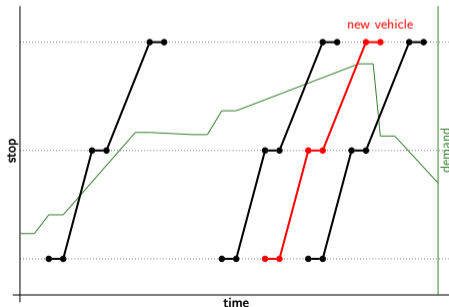
Greedy temporal insertion

- ▶ Detect the largest (feasible) time slot in-between two vehicles.
- ▶ Insert vehicle in the centre of the time slot.
- ▶ Repeat.



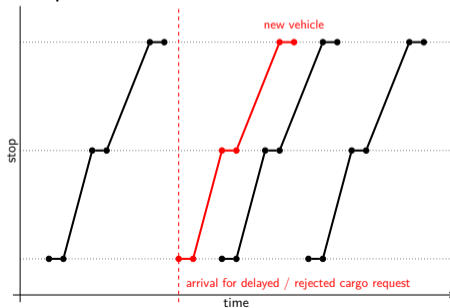
Greedy passenger demand insertion

- ▶ Detect (feasible) time slot with highest passenger congestion.
- ▶ Insert vehicle in the time slot.
- ▶ Repeat.



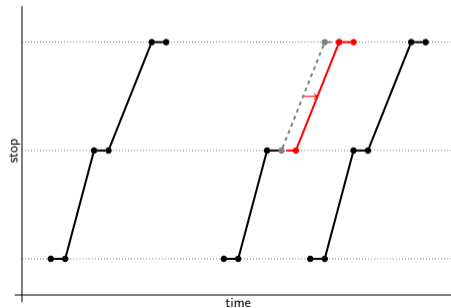
Greedy freight demand insertion

- ▶ Detect rejected cargo request OR cargo request with highest delay
- ▶ Insert vehicle in the next feasible time slot after cargo request release.
- ▶ Repeat.



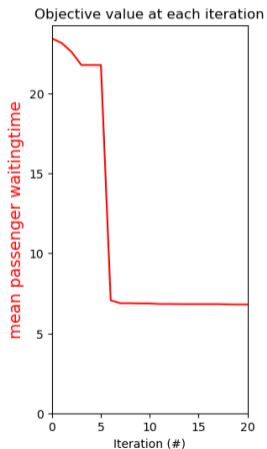
Increase service time buffer

- ▶ Detect the vehicle and stop with minimal slack between service and dwell time.
- ▶ Increase the dwell time by one time unit.

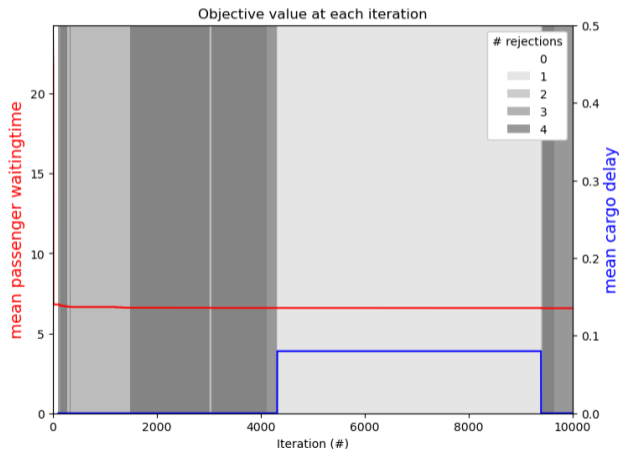


Development of the lexicographical objective function

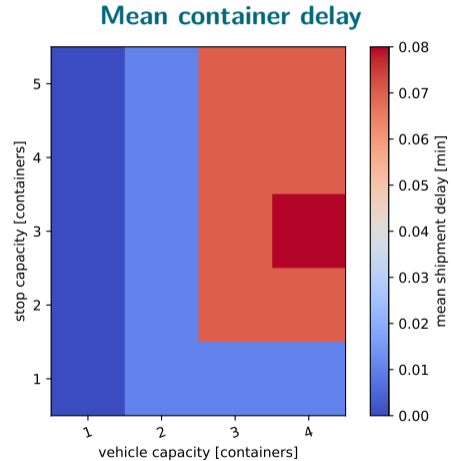
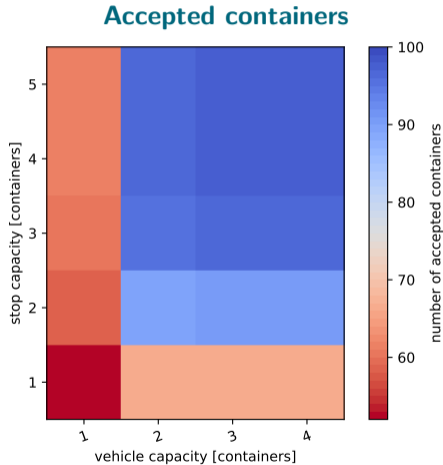
20 iterations



10000 iterations



Cargo service evaluation for varying capacities



- Model supports, e.g., estimation of required cargo capacity for stops and vehicles.

Conclusion



Summary

- Bi-objective train schedule and cargo allocation: Passengers first, cargo second
- Extension of Adaptive Large Neighbourhood Search by Yin et al., 2021
- Integration of cargo allocation problem

Next steps



- Store and consider intermediate solutions
- Different representative instances to consider stochasticity (currently expected values)
- Tram für Kiel*: Real data provided by city Kiel

Thank you for your attention! Questions?

Working paper available at SSRN: <https://ssrn.com/abstract=3886691>

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