

Transportation Models

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- large variety of models due to the many modes of transportation
 - roads
 - railroad
 - shipping
 - airlines
- as a consequence different type of equipment and resources with different characteristics are involved
 - cars, trucks, roads
 - trains, tracks and stations
 - ships and ports
 - planes and airports
- consider two specific problems

Tanker Scheduling

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Basic Characteristics

- consider the problem from the view of a company
- the planning process normally is done in a 'rolling horizon' fashion
- company operates a fleet of ships consisting of
 - own ships $\{1, \dots, T\}$
 - chartered ships
- the operating costs of these two types are different
- only the own ships are scheduled
- using chartered ships only leads to costs which are given by the spot market

Tanker Scheduling

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Basic Characteristics (cont.)

- each own ship i is characterized by its
 - capacity cap_i
 - draught dr_i
 - range of possible speeds
 - location l_i and time r_i at which it is ready to start next trip
 - ...

Tanker Scheduling

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Basic Characteristics (cont.)

- the company has n cargos to be transported
- cargo j is characterized by
 - type t_j (e.g. crude type)
 - quantity p_j
 - load port $port_j^l$ and delivery port $port_j^d$
 - time windows $[r_j^l, d_j^l]$ and $[r_j^d, d_j^d]$ for loading and delivery
 - load and unload times t_j^l and t_j^d
 - costs c_j^* denoting the price which has to be paid on the spot market to transport cargo j

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Basic Characteristics (cont.)

- there are p different ports
- port k is characterized by
 - its location
 - limitations on the physical characteristics (e.g. length, draught, deadweight, ...) of the ships which may enter the port
 - local government rules (e.g. in Nigeria a ship has to be loaded above 90% to be allowed to sail)
 - ...

Tanker Scheduling

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Basic Characteristics (cont.)

- the objective is to minimize the total cost of transporting all cargos
- hereby a cargo can be assigned to a ship of the company or 'sold' on the spot market and thus be transported by a chartered ship
- costs consist of
 - operating costs for own ships
 - spot charter rates
 - fuel costs
 - port charges, which depend on the deadweight of the ship

Tanker Scheduling

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ILP modeling

- straightforward choice of variables would be to use 0 – 1-variables for assigning cargos to ships
- problem: these assignment variables do not define the schedule/route for the ship and thus feasibility and costs of the assignment can not be determined
- alternative approach: generate a set of possible schedules/routes for each ship and afterwards use assignment variables to assign schedules/routes to ships
- problem splits up into two subproblems:
 - generate schedules for ships
 - assign schedules to ships

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ILP modeling - generate schedules

- a schedule for a ship consist of an assignment of cargos to the ship and a sequence in which the corresponding ports are visited
- generation of schedules can be done by ad-hoc heuristics which consider
 - ship constraints like capacity, speed, availability, ...
 - port constraints
 - time windows of cargos
- each schedule leads to a certain cost
- for each ship enough potential schedules should be generated in order to get feasible and good solutions for the second subproblem

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ILP modeling - generate schedules (cont.)

- the output of the first subproblem is
 - a set S_i of possible schedules for ship i
 - each schedule $l \in S_i$ is characterized by
 - * a vector $(a_{i1}^l, \dots, a_{in}^l)$ where $a_{ij}^l = 1$ if cargo j is transported by ship i in schedule l and 0 otherwise
 - * costs c_i^l denoting the incremental costs of operating ship i under schedule l versus keeping it idle over the planning horizon
 - * profit $\pi_i^l = \sum_{j=1}^n a_{ij}^l c_j^* - c_i^l$ by using schedule l for ship i instead of paying the spot market

Tanker Scheduling

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ILP modeling - generate schedules (cont.)

- Remarks:
 - all the feasibility constraints of the ports and ships are now within the schedule
 - all cost aspects are summarized in the values c_i^l resp. π_i^l
 - the sequences belonging to the schedules determine feasibility and the costs c_i^l but are not part of the output since they are not needed in the second subproblem

Tanker Scheduling

ILP modeling - assign schedules to ships

- variables $x_i^l = \begin{cases} 1 & \text{if ship } i \text{ follows schedule } l \\ 0 & \text{else} \end{cases}$
- objective: $\max \sum_{i=1}^T \sum_{l \in S_i} \pi_i^l x_i^l$
- constraint:
 - $\sum_{i=1}^T \sum_{l \in S_i} a_{ij}^l x_i^l \leq 1; \quad j = 1, \dots, n$ (each cargo at most once)
 - $\sum_{l \in S_i} x_i^l \leq 1; \quad i = 1, \dots, T$ (each ship at most one schedule)

Tanker Scheduling

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ILP modeling - assign schedules to ships (cont.)

- the ILP model is a set-packing problem and well studied in the literature
- can be solved by branch and bound procedures
- possible branchings:
 - chose a variable x_i^l and branch on the two possibilities $x_i^l = 0$ and $x_i^l = 1$
select x_i^l on base of the solution of the LP-relaxation: choose a variable with value close to 0.5
 - chose a ship i and branch on the possible schedules $l \in S_i$
selection of ship i is e.g. be done using the LP-relaxation: choose a ship with a highly fractional solution

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ILP modeling - assign schedules to ships (cont.)

- lower bounds can be achieved by generating feasible solutions via clever heuristics (feasible solution = lower bound since we have a maximization problem)
- upper bounds can be obtained via relaxing the integrality constraints and solving the resulting LP (note, that this LP-solution is also used for branching!)
- for a small example, the behavior of the branch and bound method is given in the handouts

Tanker Scheduling

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Remarks Two Phase Approach

- in general the solution after solving the two subproblems is only a heuristic solution of the overall problem
- if in the first subproblem all possible schedules/routes for each ship are generated (i.e. S_i is equal to the set S_i^{all} of all feasible schedules for ship i), the optimal solution of the second subproblem is an optimal solution for the overall problem
- for real life instances the cardinalities of the sets S_i^{all} are too large to allow a complete generation (i.e. S_i is always a (small) subset of S_i^{all})
- column generation can be used to improve the overall quality of the resulting solution

Train Timetabling

-1-

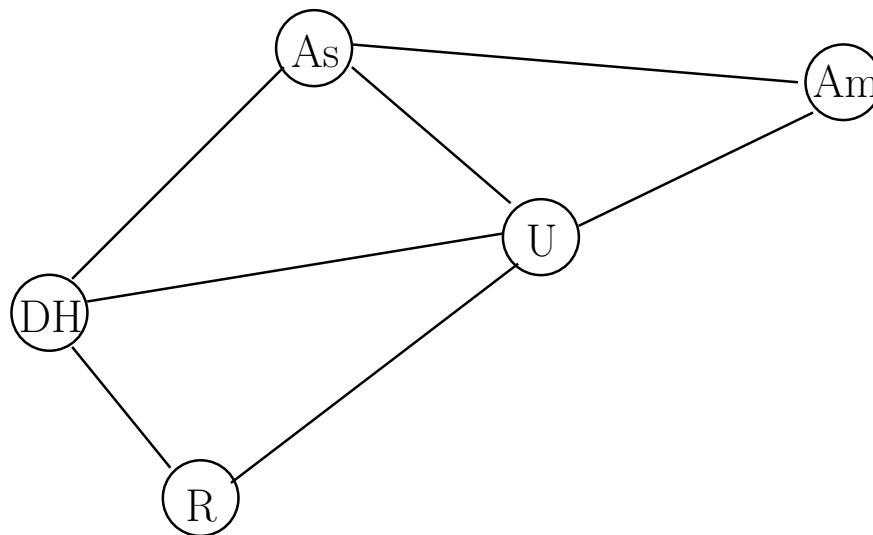
General Remarks

- in the railway world lots of scheduling problems are of importance
 - scheduling trains in a timetable
 - routing of material
 - staff planning
 - ...
- currently lots of subproblems are investigated
- the goal is to achieve an overall decision support system for the whole planning process
- we consider one important subproblem

Train Timetabling

Decomposition of the Train Timetabling

- mostly the overall railway network consists of some mayor stations and 'lines/corridors' connecting them



Am Amersfoort
As Amsterdam Centraal
DH Den Haag Centraal
R Rotterdam Centraal
U Utrecht Centraal

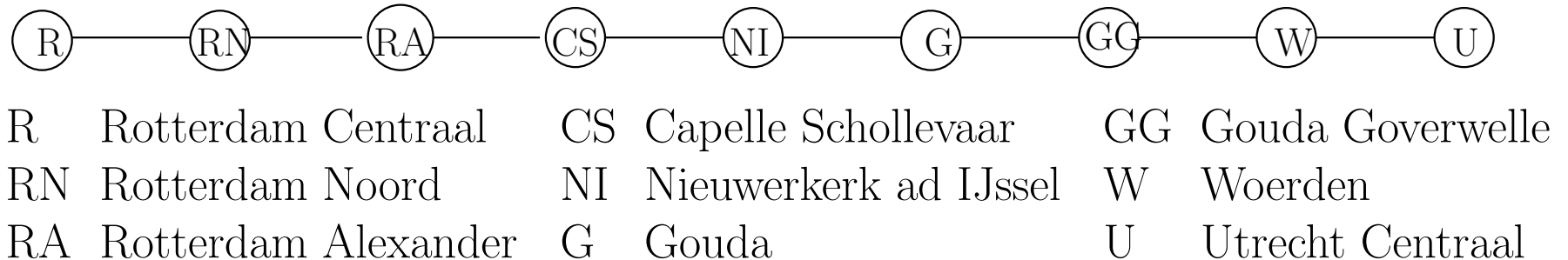
- a corridor normally consists of two independent one-way tracks
- having good timetables for the trains in the corridors makes it often easy to find timetables for the trains on the other lines

Train Timetabling

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Scheduling Train on a Track

- consider a track between two mayor stations
- in between the two mayor stations several smaller stations exists



- trains may or may not stop at these stations
- trains can only overtake each other at stations

Train Timetabling

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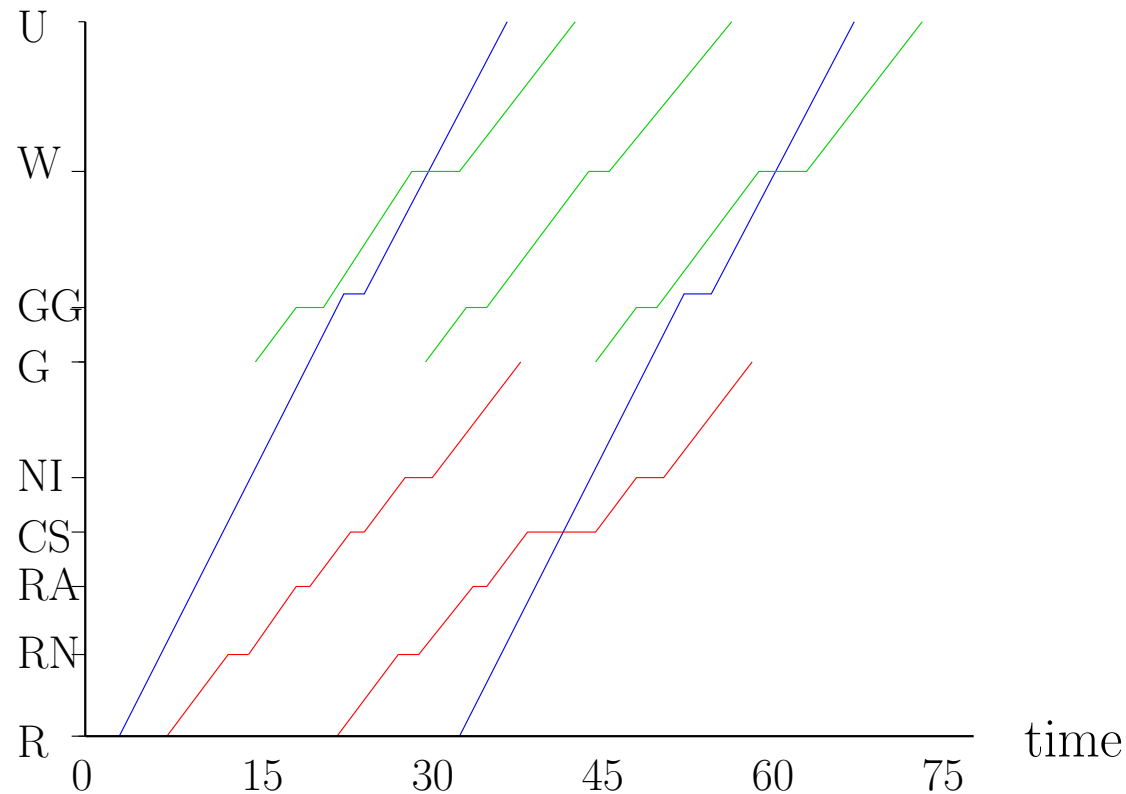
Problem Definition Track Scheduling

- time period $1, \dots, q$, where q is the length of the planning period (typically measured in minutes; e.g. $q = 1440$)
- $L + 1$ stations $0, \dots, L$
- L consecutive links;
- link j connects station $j - 1$ and j
- trains travel in the direction from station 0 to L
- T : set of trains that are candidates to run during planning period
- for link j , $T_j \subset T$ denotes the trains passing the link

Train Timetabling

Problem Definition Track Scheduling (cont.)

- train schedules are usually depicted in so-called time-space diagrams

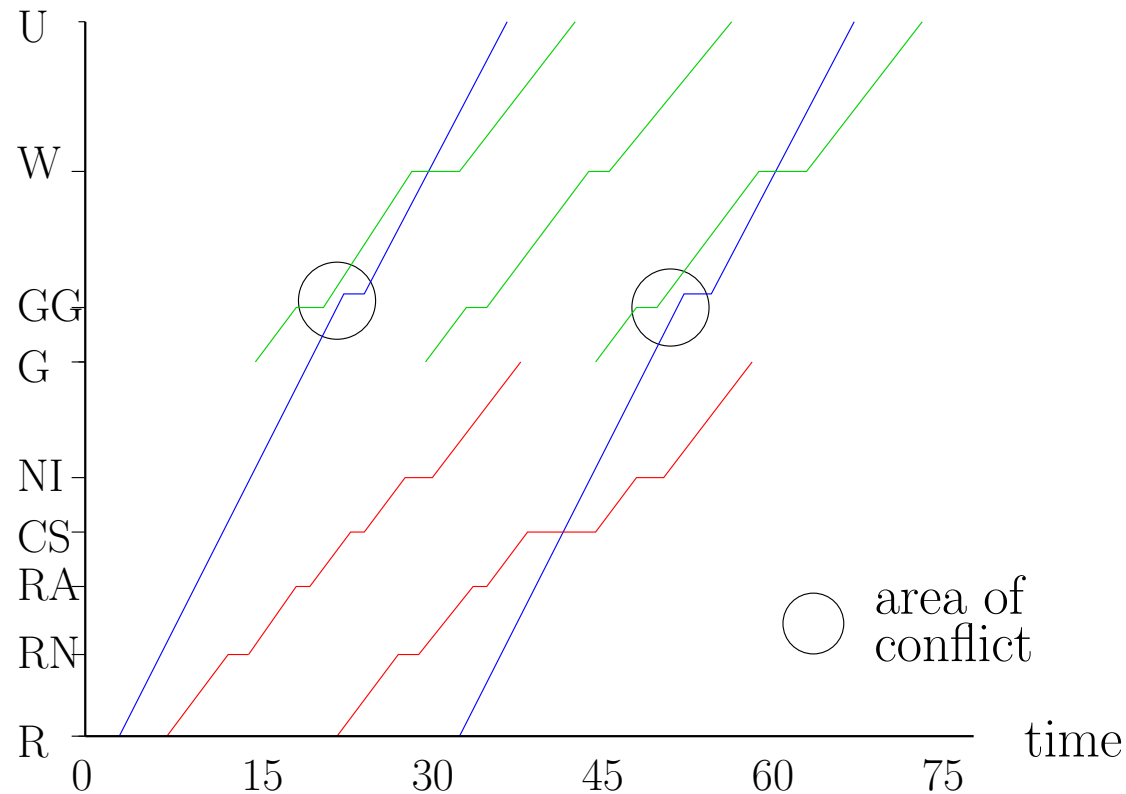


- diagrams enable user to see conflicts

Train Timetabling

Problem Definition Track Scheduling (cont.)

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Train Timetabling

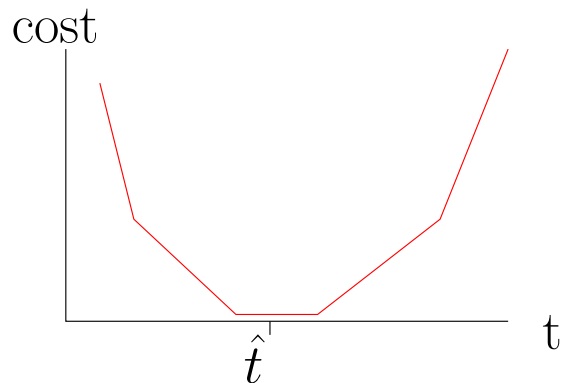
Problem Definition Track Scheduling (cont.)

- each train has an most desirable timetable (arrivals, departures, travel time on links, stopping time at stations), achieved e.g. via marketing department
- putting all these most desirable timetables together, surely will lead to conflicts on the track
- possibilities to change a timetable:
 - slow down train on link
 - increase stopping time at a station
 - modify departure time at first station
 - cancel the train

Train Timetabling

Problem Definition Track Scheduling (cont.)

- cost of deviating from a given time \hat{t} :
 - specifies the revenue loss due to a deviation from \hat{t}
 - the cost function has its minimum in \hat{t} , is convex, and often modeled by a piecewise linear function



- piecewise linear helps in ILP models!

Train Timetabling

Variables for Track Scheduling

- variables represent departure and arrival times from stations
 - y_{ij} : time train i enters link j
= time train i departs from station $j - 1$
(defined if $i \in T_j$)
 - z_{ij} : time train i leaves link j
= time train i arrives at station j
(defined if $i \in T_j$)
- $c_{ij}^d(y_{ij})$ ($c_{ij}^a(z_{ij})$) denotes the cost resulting from the deviation of the departure time y_{ij} (arrival time z_{ij}) from its most desirable value

Train Timetabling

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Variables for Track Scheduling (cont.)

- variables resulting from the departures and arrivals times:
 - $\tau_{ij} = z_{ij} - y_{ij}$: travel time of train i on link j
 - $\delta_{ij} = y_{i,j+1} - z_{ij}$: stopping time of train i at station j
- $c_{ij}^\tau(\tau_{ij})$ ($c_{ij}^\delta(\delta_{ij})$) denotes the cost resulting from the deviation of the travel time τ_{ij} (stopping time δ_{ij}) from its most desirable value
- all cost functions $c_{ij}^d, c_{ij}^a, c_{ij}^\tau, c_{ij}^\delta$ have the mentioned structure

Train Timetabling

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Objective function

- minimize

$$\sum_{j=1}^L \sum_{i \in T_j} (c_{ij}^d(y_{ij}) + c_{ij}^a(z_{ij}) + c_{ij}^T(z_{ij} - y_{ij})) \\ + \sum_{j=1}^{L-1} \sum_{i \in T_j} c_{ij}^\delta(y_{i,j+1} - z_{ij})$$

Train Timetabling

Constraints

- minimum travel times for train i over link j : τ_{ij}^{min}
- minimum stopping times for train i at station j : δ_{ij}^{min}
- safety distance:
 - minimum headway between departure times of train h and train i from station j : H_{hij}^d
 - minimum headway between arrival times of train h and train i from station j : H_{hij}^a
- lower and upper bounds on departure and arrival times:
 $y_{ij}^{min}, y_{ij}^{max}, z_{ij}^{min}, z_{ij}^{max}$

Train Timetabling

Constraints (cont.)

- to be able to model the minimum headway constraints, variables have to be introduced which control the order of the trains on the links
- $x_{hij} = \begin{cases} 1 & \text{if train } h \text{ immediately preceeds train } i \text{ on link } j \\ 0 & \text{else} \end{cases}$
- using the variables x_{hij} , the minimum headway constraints can be formulated via 'big M'-constraints:

$$y_{i,j+1} - y_{h,j+1} + (1 - x_{hij})M \geq H_{hij}^d$$

$$z_{ij} - z_{hj} + (1 - x_{hij})M \geq H_{hij}^a$$

Train Timetabling

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Constraints (cont.)

- two dummy trains 0 and * are added, representing the start and end of the planning period (fix departure and arrival times appropriate ensuring that 0 is sequenced before all other trains and * after all other trains)

Train Timetabling

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Constraints (cont.)

$$y_{ij} \geq y_{ij}^{\min} \quad j = 1, \dots, L; \quad i \in T_j$$

$$y_{ij} \leq y_{ij}^{\max} \quad j = 1, \dots, L; \quad i \in T_j$$

$$z_{ij} \geq z_{ij}^{\min} \quad j = 1, \dots, L; \quad i \in T_j$$

$$z_{ij} \leq z_{ij}^{\max} \quad j = 1, \dots, L; \quad i \in T_j$$

$$z_{ij} - y_{ij} \geq \tau_{ij}^{\min} \quad j = 1, \dots, L; \quad i \in T_j$$

$$y_{i,j+1} - z_{ij} \geq \delta_{ij}^{\min} \quad j = 1, \dots, L-1; \quad i \in T_j$$

$$y_{i,j+1} - y_{h,j+1} + (1 - x_{hij})M \geq H_{hij}^d \quad j = 0, \dots, L-1; \quad i, h \in T_j$$

$$z_{ij} - z_{hj} + (1 - x_{hij})M \geq H_{hij}^a \quad j = 1, \dots, L; \quad i, h \in T_j$$

$$\sum_{h \in T_j \setminus \{i\}} x_{hij} = 1 \quad j = 1, \dots, L; \quad i \in T_j$$

$$x_{hij} \in \{0, 1\} \quad j = 1, \dots, L; \quad i, h \in T_j$$

Train Timetabling

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Remarks on ILP Model

- the number of 0-1 variables gets already for moderate instances quite large
- the single track problem is only a subproblem in the whole time tabling process and needs therefore to be solved often
- as a consequence, the computational time for solving the single track problem must be small
- this asks for heuristic approaches to solve the single track problem

Train Timetabling

Decomposition Approach: General Idea

- schedule the trains iteratively one by one
- initially, the two dummy trains 0 and * are scheduled
- the selection of the next train to be scheduled is done on base of priorities
- possible priorities are
 - earliest desired departure time
 - decreasing order of importance (importance may be e.g. measured by train type, speed, expected revenue, ...)
 - smallest flexibility in departure and arrival
 - combinations of the above

Train Timetabling

Decomposition Approach: Realization

- T_0 : set of already scheduled trains
- initially $T_0 = \{0, *\}$
- after each iteration a schedule of the trains from T_0 is given
- however, for the next iteration only the sequence in which the trains from T_0 traverse the links is taken into account
- $S_j = (0 = j_0, j_1, \dots, j_{n_j}, j_{n_j+1} = *)$: sequence of trains from T_0 on link j
- if train k is chosen to be scheduled in an iteration, we have to insert k in all sequences S_j where $k \in T_j$
- this problem is called $Insert(k, T_0)$

Train Timetabling

ILP Formulation of $Insert(k, T_0)$

Adapt the 'standard' constraints and the objective to T_0 :

$$\begin{aligned} \min \quad & \sum_{j=1}^L \sum_{i \in T_j} (c_{ij}^d(y_{ij}) + c_{ij}^a(z_{ij}) + c_{ij}^\tau(z_{ij} - y_{ij})) \\ & + \sum_{j=1}^{L_1} \sum_{i \in T_j} c_{ij}^\delta(y_{i,j+1} - z_{ij}) \end{aligned}$$

subject to

$$\begin{aligned} y_{ij} &\geq y_{ij}^{\min} & j = 1, \dots, L; \ i \in T_0 \cap T_j \\ y_{ij} &\leq y_{ij}^{\max} & j = 1, \dots, L; \ i \in T_0 \cap T_j \\ z_{ij} &\geq z_{ij}^{\min} & j = 1, \dots, L; \ i \in T_0 \cap T_j \\ z_{ij} &\leq z_{ij}^{\max} & j = 1, \dots, L; \ i \in T_0 \cap T_j \\ z_{ij} - y_{ij} &\geq \tau_{ij}^{\min} & j = 1, \dots, L; \ i \in T_0 \cap T_j \\ y_{i,j+1} - z_{ij} &\geq \delta_{ij}^{\min} & j = 1, \dots, L-1; \ i \in T_0 \cap T_j \end{aligned}$$

Train Timetabling

ILP Formulation of $Insert(k, T_0)$ (cont.)

- adapt $y_{i,j+1} - y_{h,j+1} + (1 - x_{hij})M \geq H_{hij}^d$ for trains from T_0

$$y_{j_{i+1},j} - y_{j_i,j} \geq H_{j_i j_{i+1},j-1}^d \quad \text{for } j = 1, \dots, L, i = 0, \dots, n_j$$

- adapt $z_{ij} - z_{hj} + (1 - x_{hij})M \geq H_{hij}^a$ for trains from T_0

$$z_{j_{i+1},j} - z_{j_i,j} \geq H_{j_i j_{i+1}j}^a \quad \text{for } j = 1, \dots, L, i = 0, \dots, n_j$$

Train Timetabling

ILP Formulation of $Insert(k, T_0)$ (cont.)

- insert k on link j via variables

$$x_{ij} = \begin{cases} 1 & \text{if train } k \text{ immediately precedes train } j_i \text{ on link } j \\ 0 & \text{else} \end{cases}$$

- new constraints for $j = 1, \dots, L, i = 0, \dots, n_j$:

$$-y_{k,j} - y_{j_i,j} + (1 - x_{ij})M \geq H_{j_i k j}^d$$

$$-y_{j_{i+1},j} - y_{k,j} + (1 - x_{ij})M \geq H_{k j_{i+1} j}^d$$

$$-z_{k,j} - z_{j_i,j} + (1 - x_{ij})M \geq H_{j_i k j}^a$$

$$-z_{j_{i+1},j} - z_{k,j} + (1 - x_{ij})M \geq H_{k j_{i+1} j}^a$$

- 0-1 constraints and sum constraint on x_{ij} values

Train Timetabling

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Remarks on ILP Formulation of $Insert(k, T_0)$

- the ILP Formulation of $Insert(k, T_0)$ has the same order of continuous constraints (y_{ij}, z_{ij}) but far fewer 0-1 variables than the original MIP
- a preprocessing may help to fix x_{ij} variables since on base of the lower and upper bound on the departure and arrival times of train k many options may be impossible
- solving $Insert(k, T_0)$ may be done by branch and bound

Train Timetabling

Solving the overall problem

- an heuristic for the overall problem may follow the ideas of the shifting bottleneck heuristic
 - select a new train k (machine) which is most 'urgent'
 - solve for this new train k the problem $Insert(k, T_0)$
 - reoptimize the resulting schedule by rescheduling the trains from T_0
- rescheduling of a train $l \in T_0$ can be done by solving the problem $Insert(l, T_0 \cup \{k\} \setminus \{l\})$ using the schedule which results from deleting train l from the schedule achieved by $Insert(k, T_0)$