



SINTEF

Solving train scheduling problems: reformulations, decompositions, and practice.

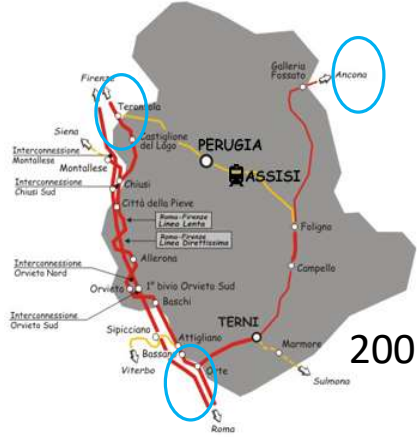
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Siemens Mobility

Aussois, January 2025

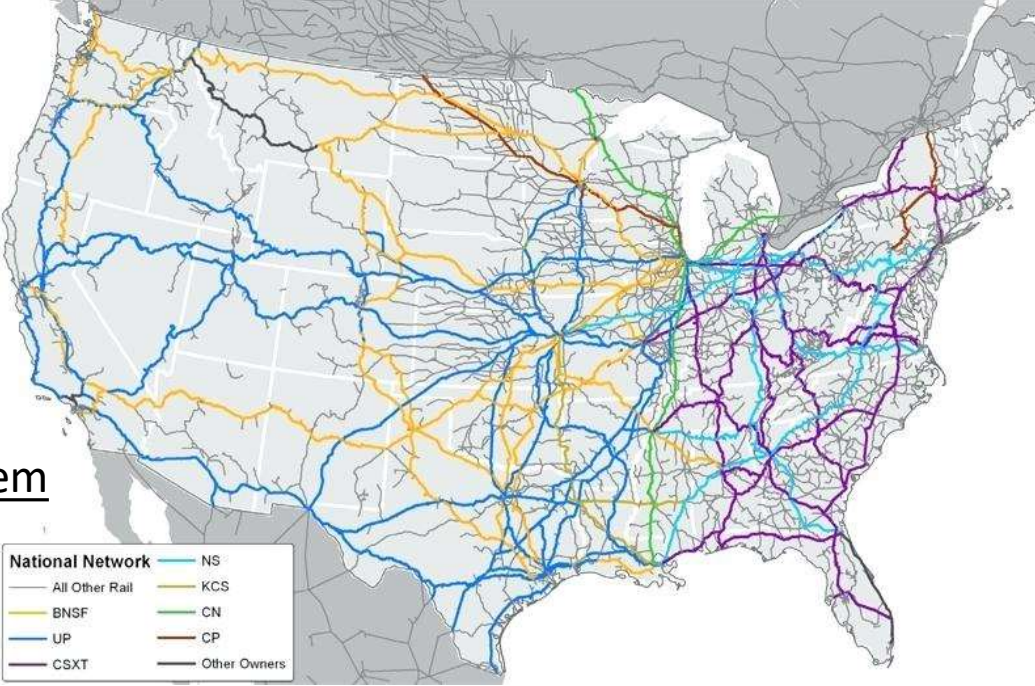
Photo: David Gubler

Technology for a better society

The challenge: dispatching in large networks



2009: REGIONAL LINES IN ITALY



National Network	
NS	Light Blue
KCS	Yellow
CN	Green
UP	Dark Blue
CP	Purple
Other Owners	Grey

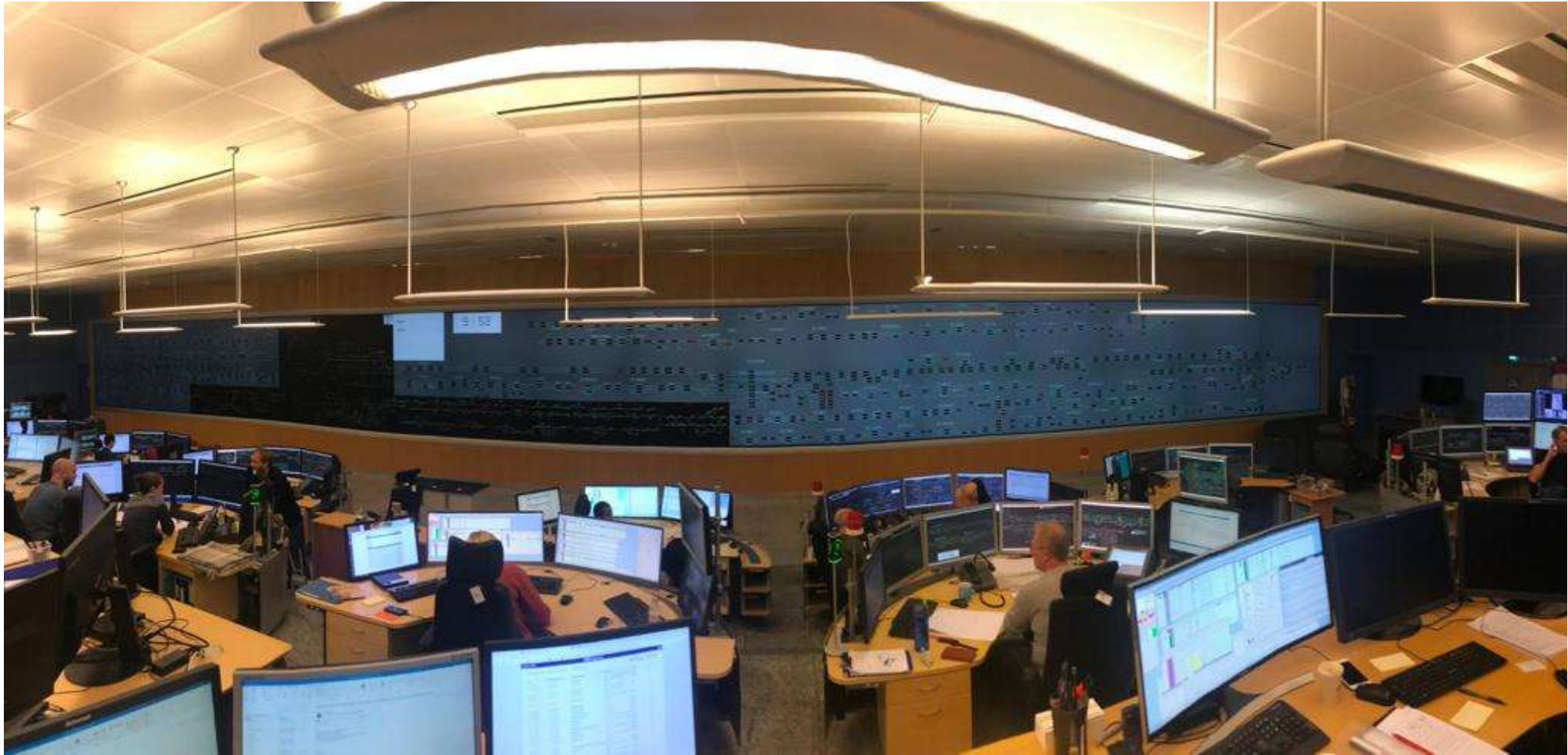
In 2017 we were asked to design and develop a system to dispatch trains in Union Pacific railway network.

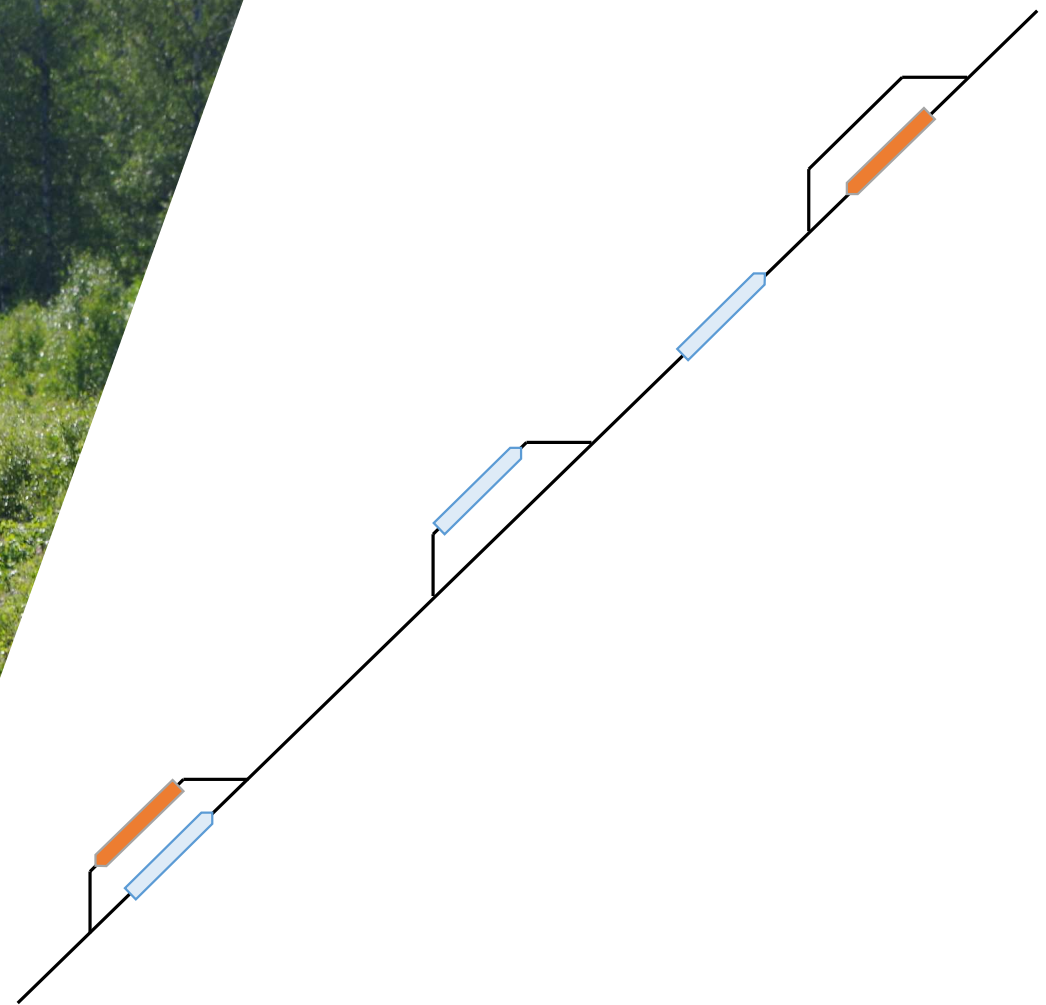
UP: Largest railroad in North America
32k miles, 23 states, 8.3k locomotives

Union Pacific: blue network.

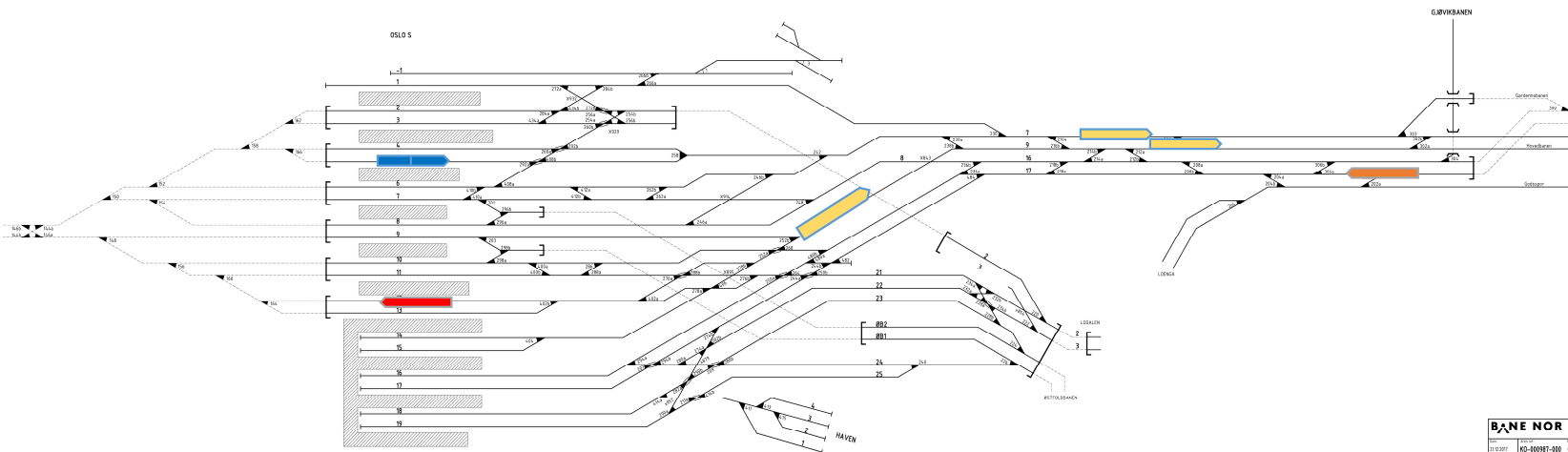
About 1000 trains running at any time any day of the year

Dispatchers at work





Train routing and scheduling (TRS)

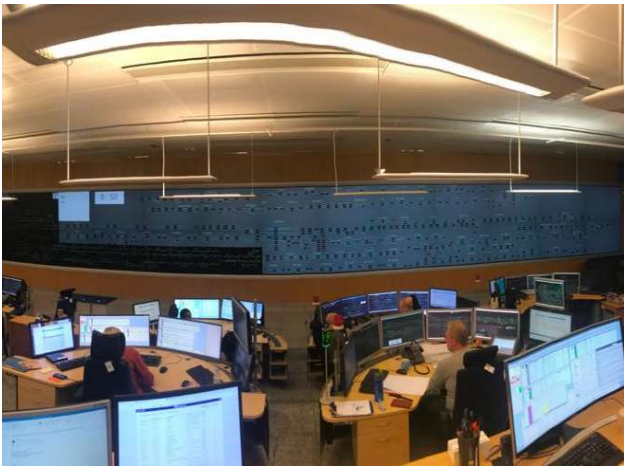


Roughly, the train routing and scheduling problem (TRS):

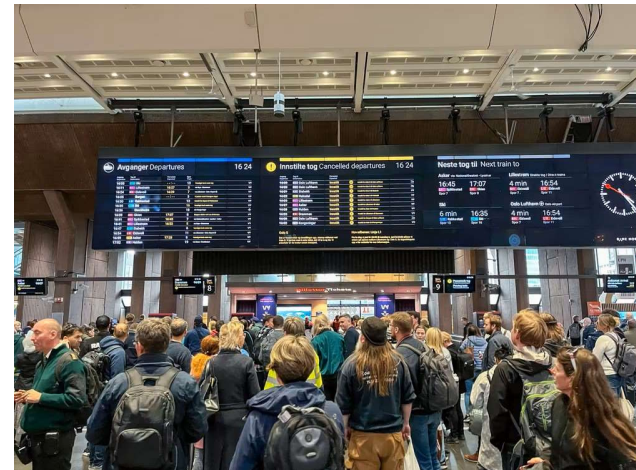
- **Given:** railway network, set of trains, traffic and time requirements, connections, ...
- **Find:** a route for each train and a schedule of the movement of each train along its route which minimizes the deviation from requirements

Versions of TRS

Operational (real-time)
Train dispatching (rescheduling)



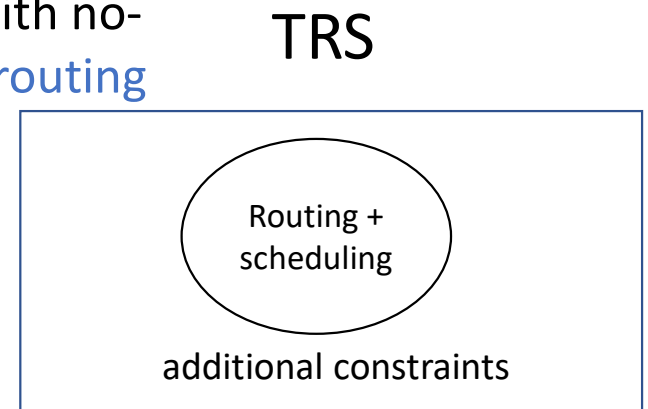
Tactical/Strategical
train timetabling



TRS is difficult

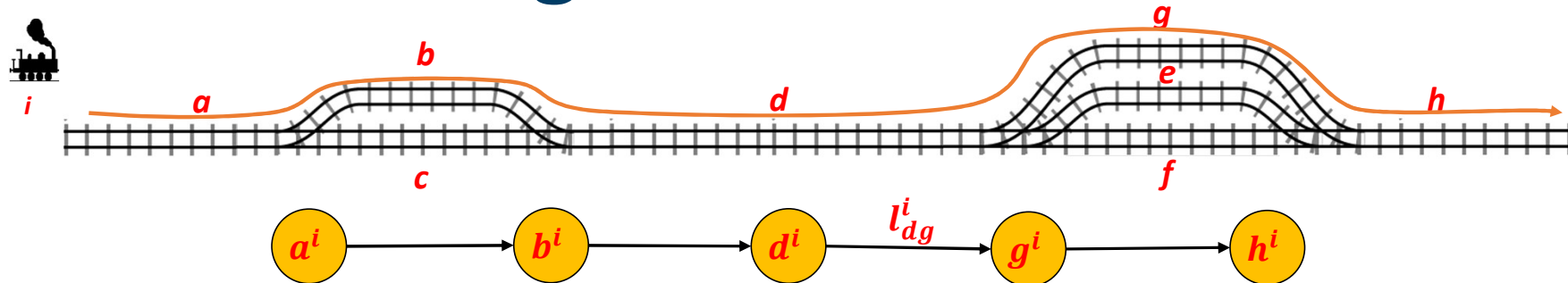
❑ TRS: Very hard to solve practical instances

1. The core is a **job-shop scheduling problem** (NP-hard) with no-wait blocking constraints (Mascis & Pacciarelli 2002) + **routing**
2. In practice, additional rules and constraints
3. Very large practical instances
4. Short computing time available (e.g. dispatching: 10-30 sec)

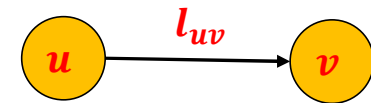


Modelling train movements

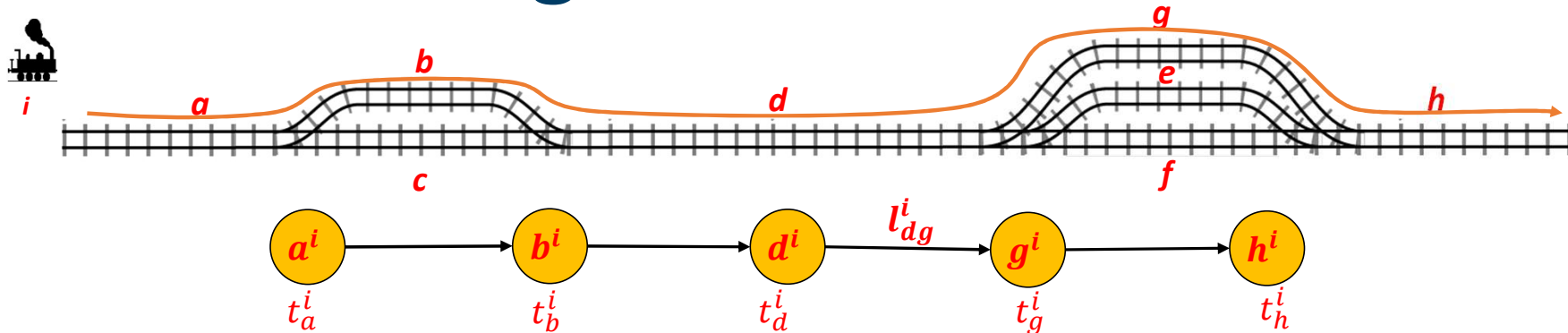
Modelling train movements



- ❑ Tracks are segmented into *track sections* (each can accommodate at most one train)
- ❑ The train movement is a sequence of track sections, represented by a **directed path** (graph)
- ❑ A node corresponds to (the event of) entering a **train entering a track section**
- ❑ An arc (u, v) represents that v is the track section following u
- ❑ The length l_{uv} of (u, v) is the minimizing running time of the train from u to v



Modelling train movements

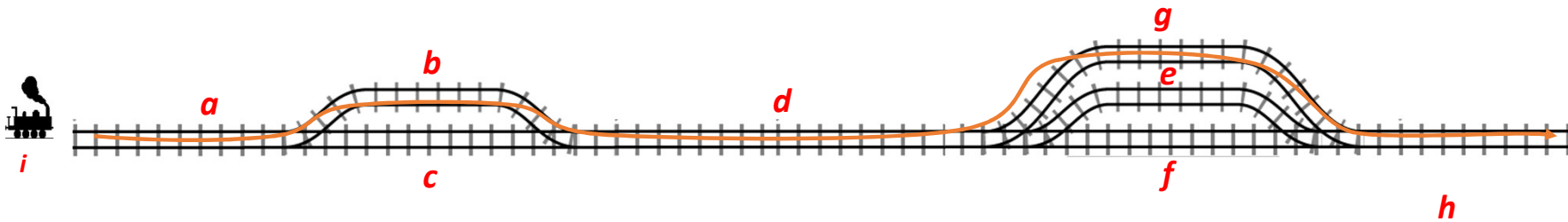


- A variable t_u is associated with node u representing **the time the train enters the track section**
- Arc (u, v) with length l_{uv} implies a time precedence constraint!

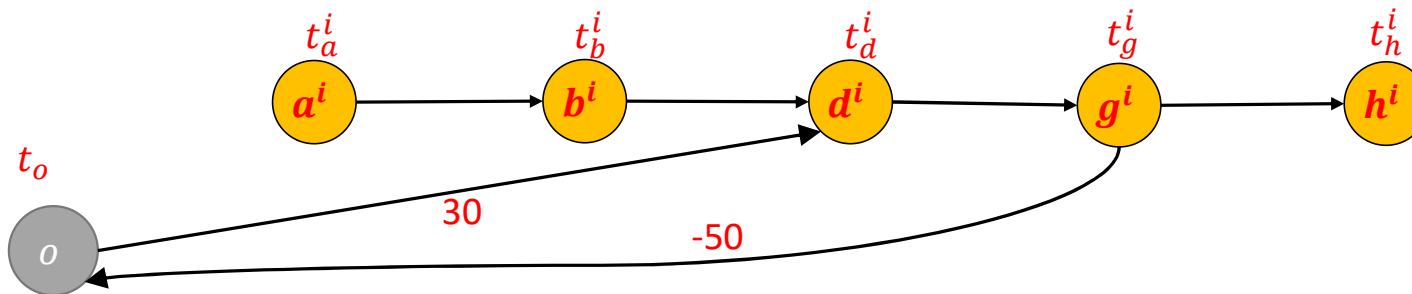
$$t_v - t_u \geq l_{uv}$$

The diagram shows a single arc between two nodes, u and v . The arc is labeled with length l_{uv} .

The time origin

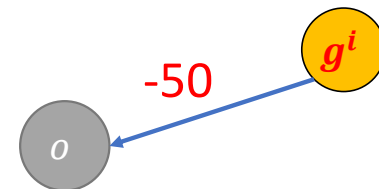


□ We have an extra node o representing the start t_o of the planning horizon

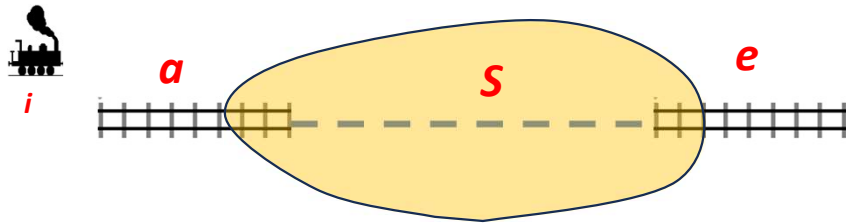


□ Departure from b not before 30: Event d^i starting not before 30: $t_d^i - t_o \geq 30$

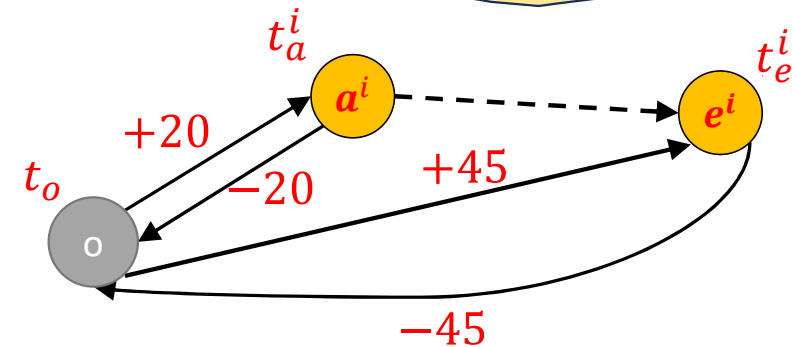
□ Arrival in g not after 50: $t_g^i - t_o \leq 50 \rightarrow t_o - t_g^i \geq -50$



Timetable arcs



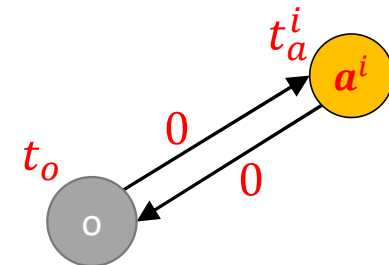
- A timetable provides arrival (departure) times of trains in certain subnetworks.
- Represented in the TRS graph by a pair of antiparallel arcs of opposite lengths.



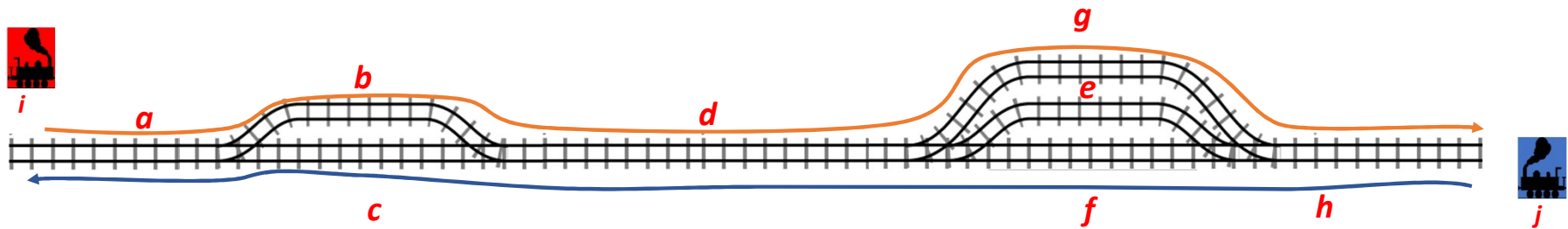
- Train i arrives in station S at time 20 ($t_a^i - t_o = 20$)
 $t_a^i - t_o \geq 20$ and $t_a^i - t_o \leq 20$

- Train i leaves S at time 45 ($t_e^i - t_o = 45$)

- OBS: Train i already in section a (starting now): $t_a^i - t_o = 0$



Conflict



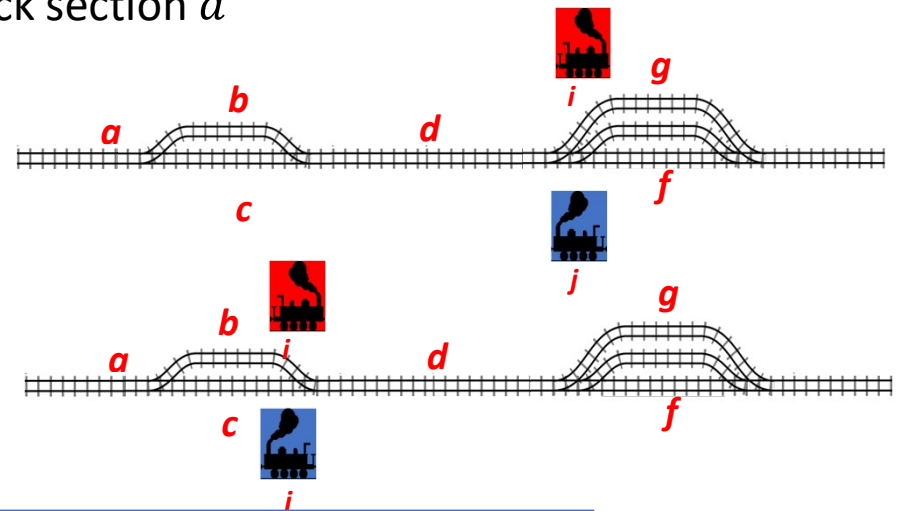
□ Conflict: both train i and train j wants to access track section d

Either i enters g before j enters d

$$t_d^j - t_g^i \geq 0$$

Or j enters c before i enters d

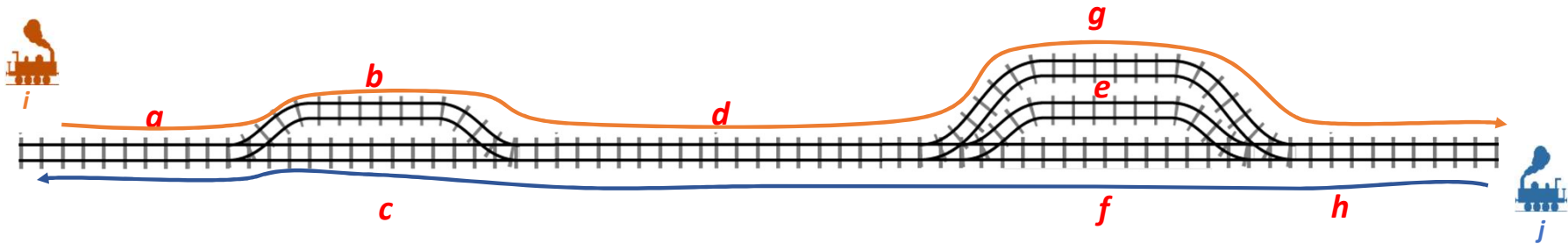
$$t_d^i - t_c^j \geq 0$$



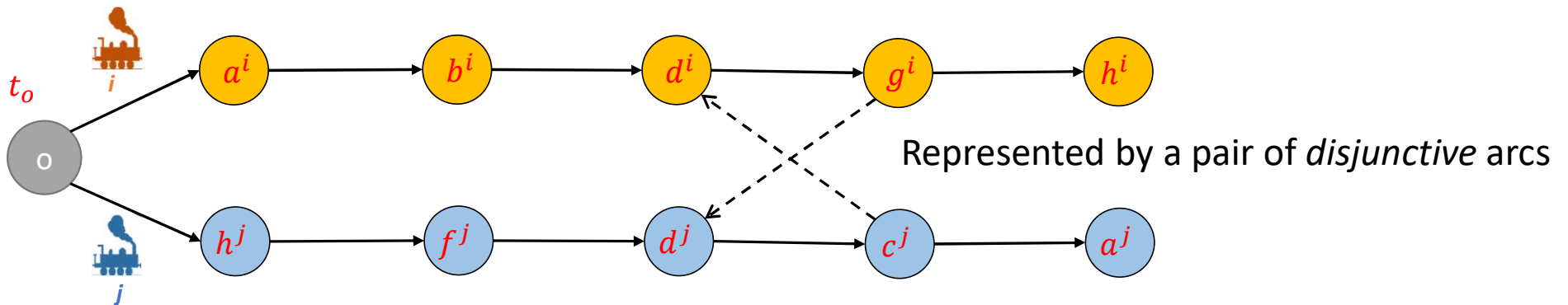
$$t_d^j - t_g^i \geq 0 \vee t_d^i - t_c^j \geq 0$$

Disjunctive (precedence) constraint

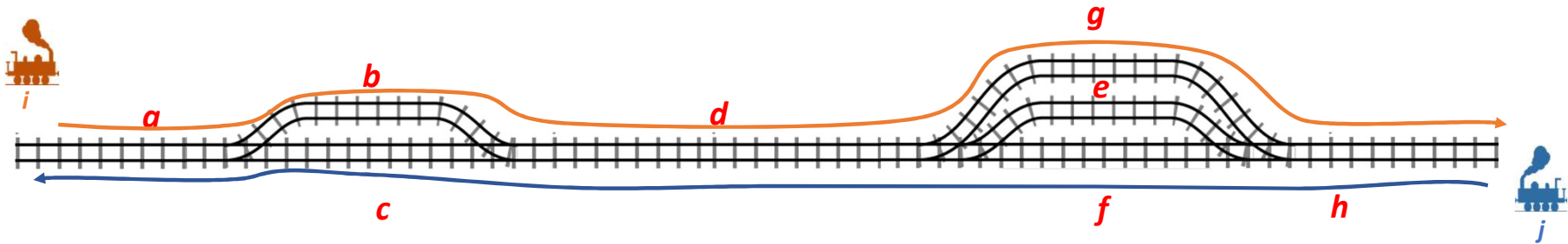
Disjunctive graph



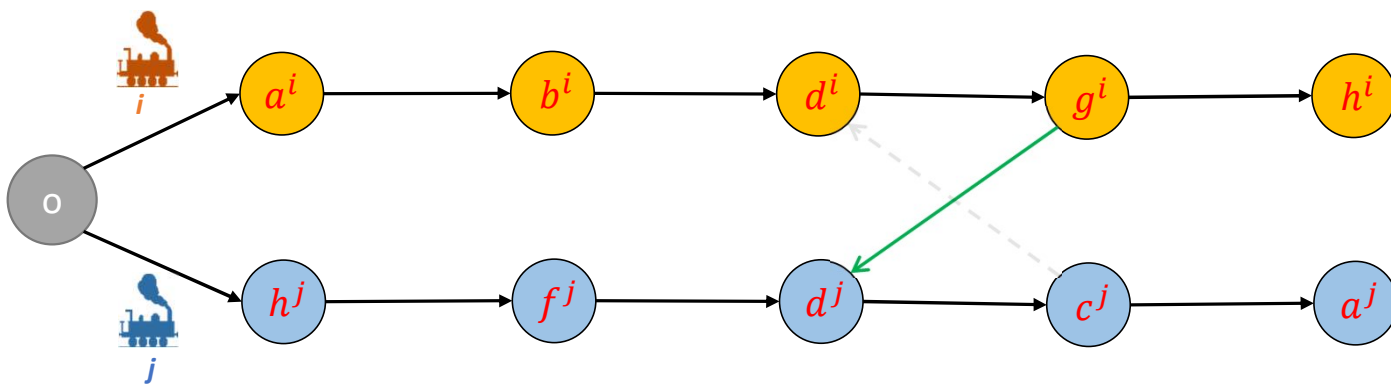
$$t_d^j - t_g^i \geq 0 \vee t_d^i - t_c^j \geq 0 \quad \text{Disjunctive precedence constraint}$$



"Solving" Conflicts

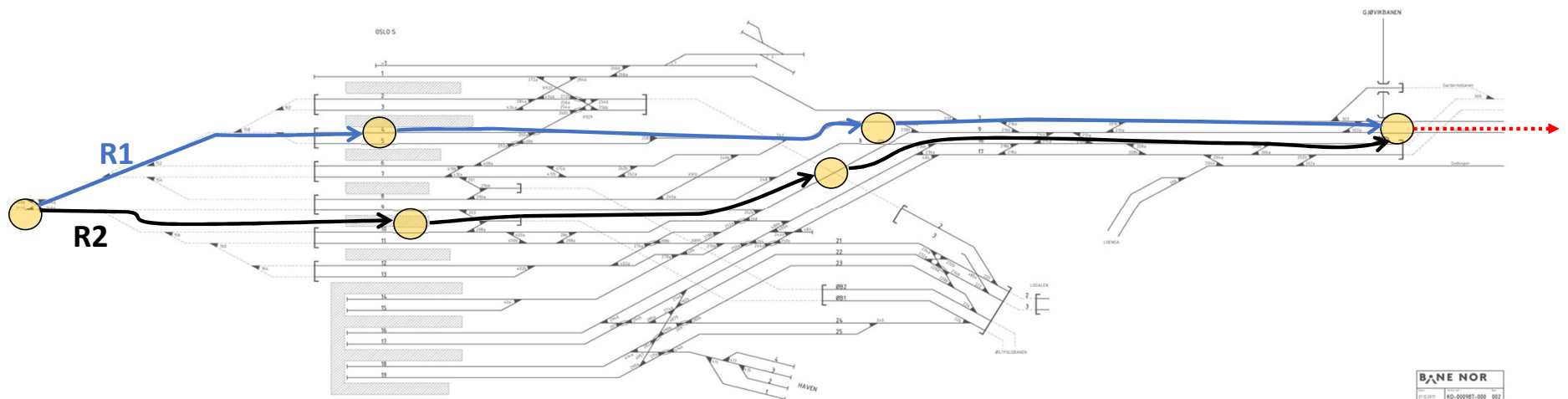


- ❑ Solving a conflict means deciding which term in $t_d^j - t_g^i \geq 0$ OR $t_d^i - t_c^j \geq 0$ to satisfy
- ❑ Equivalent to selecting an arc and dropping the other



train i goes first
 $t_d^j - t_g^i \geq 0$

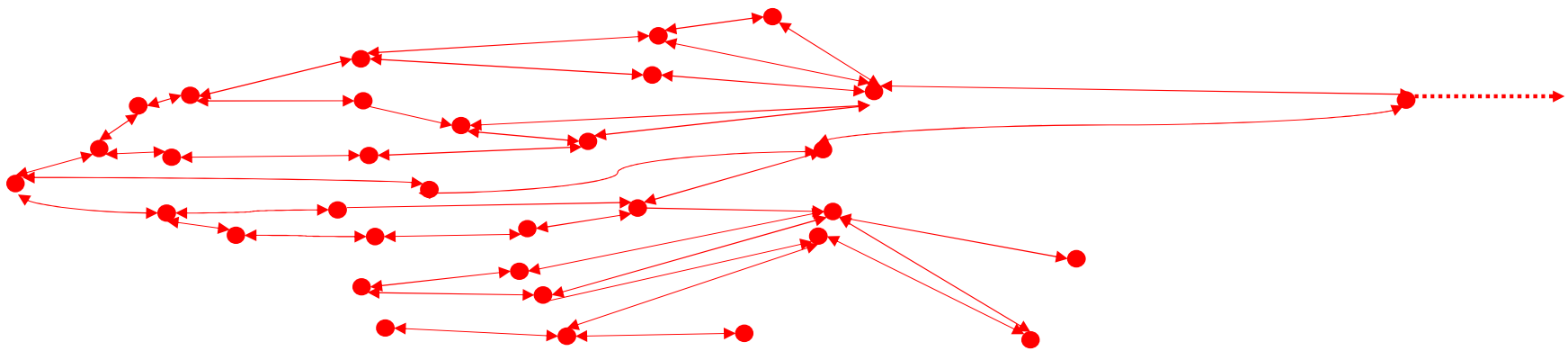
Alternative routes



- ❑ Two alternative routes for the same train in Oslo station

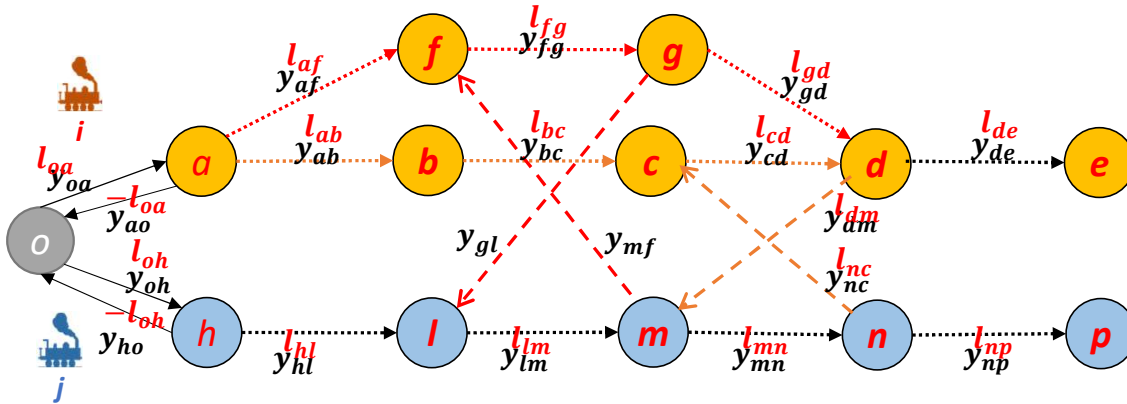
The rail network graph

- Networks with many routing options blow up computational complexity



A MIP formulation

An instance of TRS: TRS Graph



- The TRS graph $G = (V, A)$ contains all **routing, conflict** and **timetable arcs**
- Extra node o for **time synchronization**
- $l \in R^A$ **length vector**
- $G = (V, A), l$ **instance of TRS**

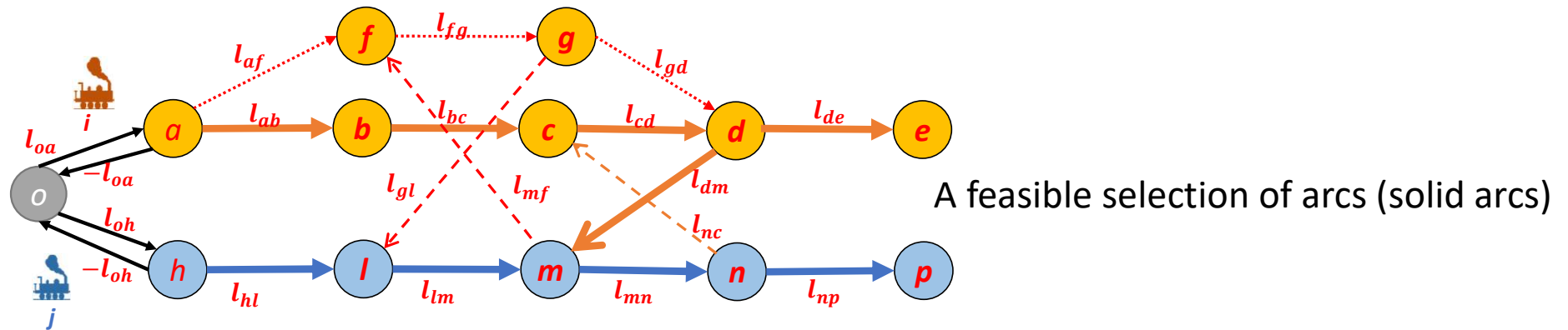
Routing and conflict-resolution decisions \iff selecting arcs in G

Mathematical model: $a \in A, y_a \in \{0,1\}$, ($y_a = 1$ if arc chosen)

Y set of **candidate solutions** $y \in \{0,1\}^A$ (can be described with linear constraints)

$\bar{y} \in Y$ incidence vector of arcs $A(\bar{y}) \subseteq A$ (*feasible selection of arcs*)

The big-M trick



$y_{uv} = 1$ = selecting arc (u, v) = imposing a time precedence constraint

$$t_v - t_u \geq +l_{uv} - (1 - y_{uv})M$$

$y_{uv} = 0$ = “deleting” (u, v)

M large positive constant

Big-M formulation for TRS

$G = (V, A), l$ instance of TRS, Y feasible (incidence vectors of) selections

$$\begin{aligned} \min \quad & f(t) \\ & t_v - t_u \geq l_{uv} - (1 - y_{uv})M \quad (u, v) \in A \\ & t \in \mathbb{R}^V, y \in Y \subseteq \{0, 1\}^A \end{aligned}$$

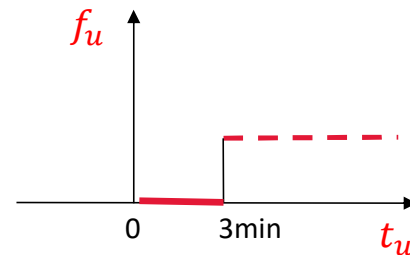
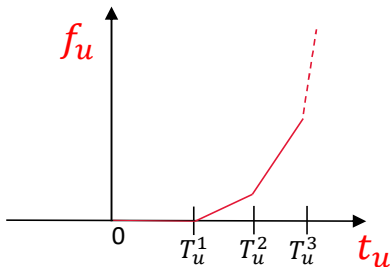
- For $\bar{y} \in Y$ TRS reduces to the following program (linear if $f(t)$ is linear):

Sched(\bar{y})

$$\begin{aligned} \min \quad & f(t) \\ & t_v - t_u \geq l_{uv} \quad (u, v) \in A(\bar{y}) \\ & t \in \mathbb{R}^V \end{aligned}$$

On the objective function $f(t)$

- ❑ Only computed in a few, special points (e.g. terminal stations $V^* \subset V$)
- ❑ $f(t)$ is separable: $f(t) = \sum_{u \in V^*} f_u(t_u)$
- ❑ Typically $f_u(t_u)$ is non-decreasing (*regular*).



- ❑ We only consider regular, separable cost functions

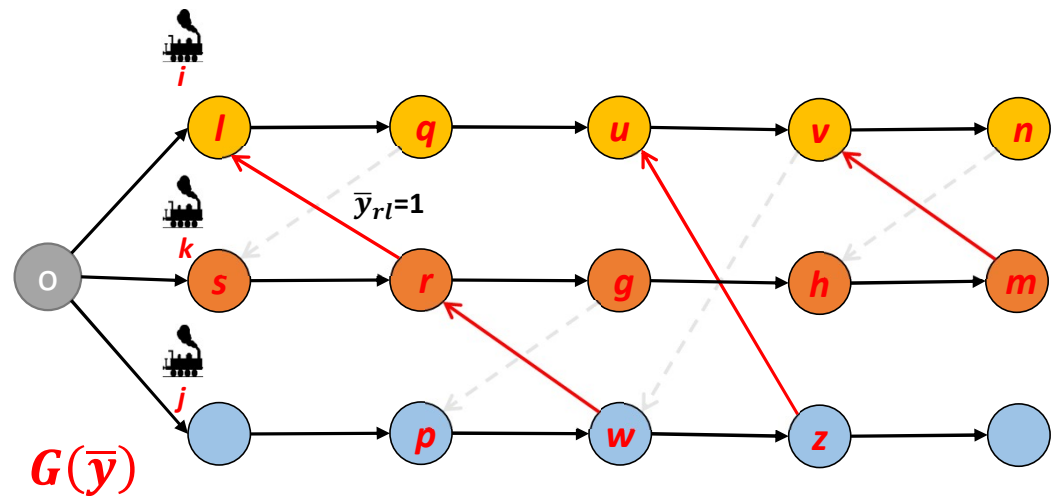
Important facts on feasible solutions

TRS graph and scheduling

□ Choose $\bar{y} \in Y$: the TRS graph reduces to a standard graph $G(\bar{y}) = (V, A(\bar{y}))$

Sched(\bar{y})

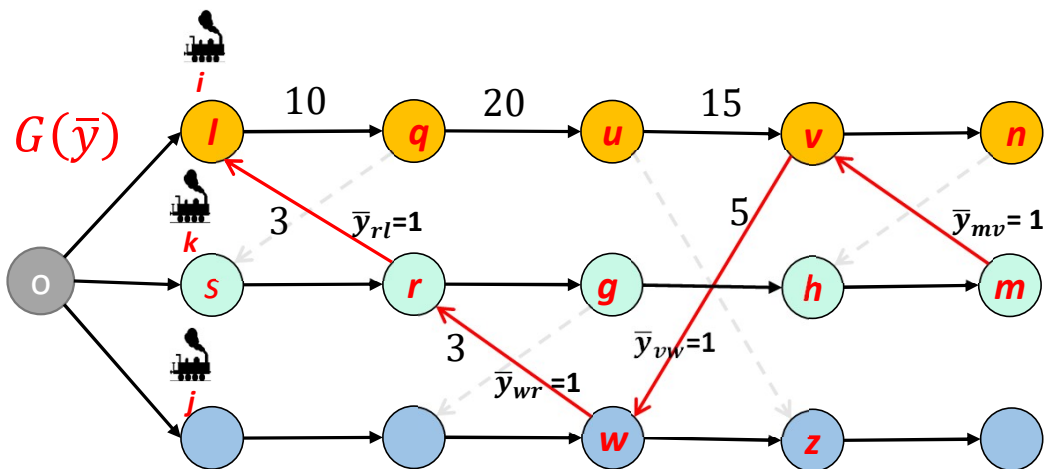
min $f(t)$
 $t_v - t_u \geq l_{uv} \quad (u, v) \in A(\bar{y})$
 $t \in \mathbb{R}^V$



□ How does $G(\bar{y})$ relate to the associated scheduling problem Sched(\bar{y})?

Feasibility

For $\bar{y} \in Y$, $\text{Sched}(\bar{y})$ has a solution, if and only if $G(\bar{y})$ does not contain a directed cycle C of positive length $l(C)$.

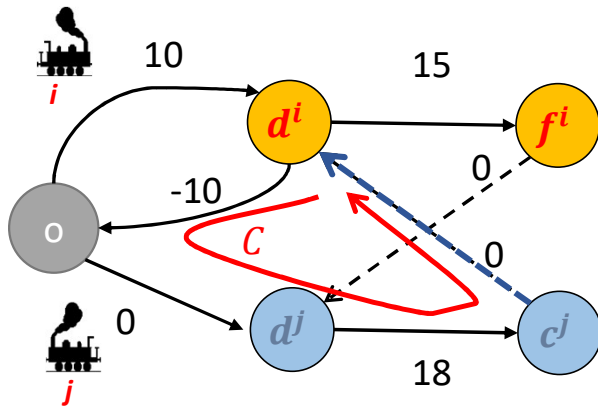
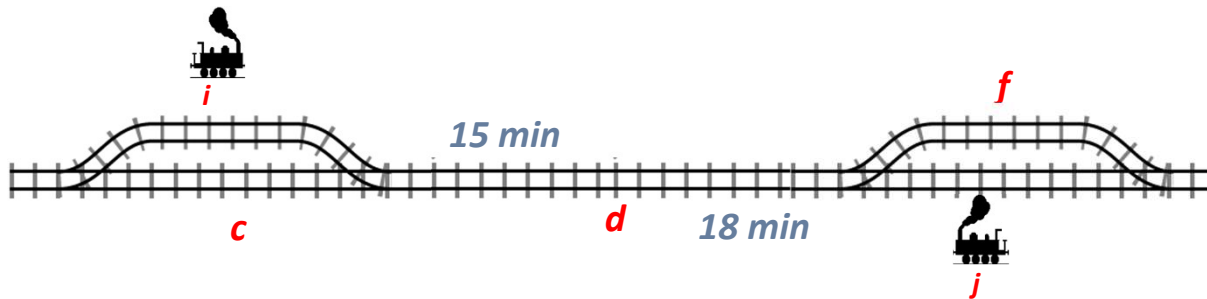


$$C = \{(lq), (qu), (uv), (vw), (wr), (rl)\}, \quad l(C) > 0$$

$\text{Sched}(\bar{y})$

min $f(t)$
 $t_v - t_u \geq l_{uv} \quad (u, v) \in A(\bar{y})$
 $t \in \mathbb{R}^V$

Example of infeasible solution



- Current time is **09:00**
- Train *i* leaves the station at **9:10** (exactly)
- Train *j* can leave the station at any time from now
- **Suppose** *j* wins the conflict on *d*

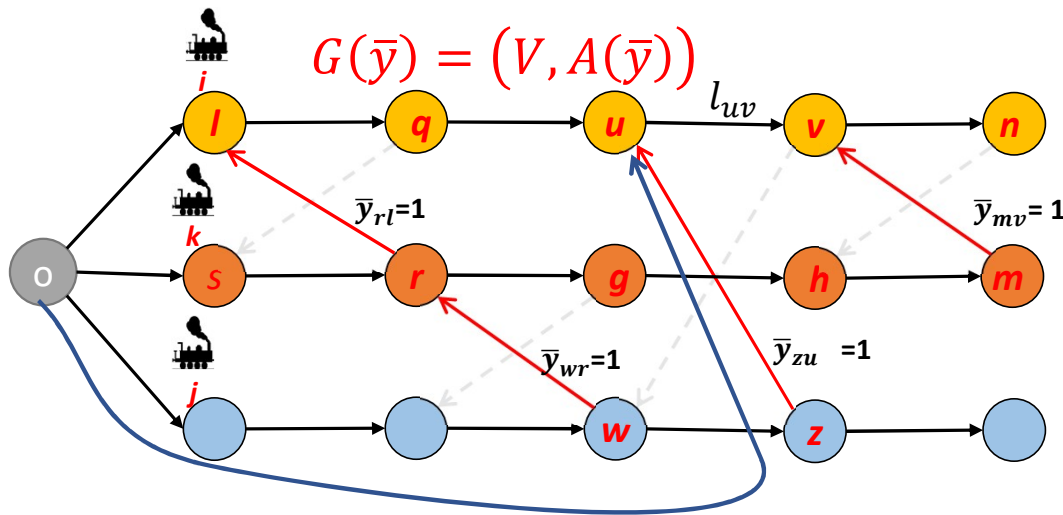
j wins → pick edge (c^j, d^i)

Cycle $C = \{c^j d^i, d^i o, o d^j, d^j c^j\}$, length $8 > 0$

Building feasible schedules

□ Instance $G = (V, A), l$. Solution $\bar{y} \in Y, G(\bar{y})$ no positive directed cycles.

• \bar{t}_u^* = length of longest path from o to $u \in V$ in $G(\bar{y})$ is feasible for Sched(\bar{y})



Sched(\bar{y})

min $f(t)$

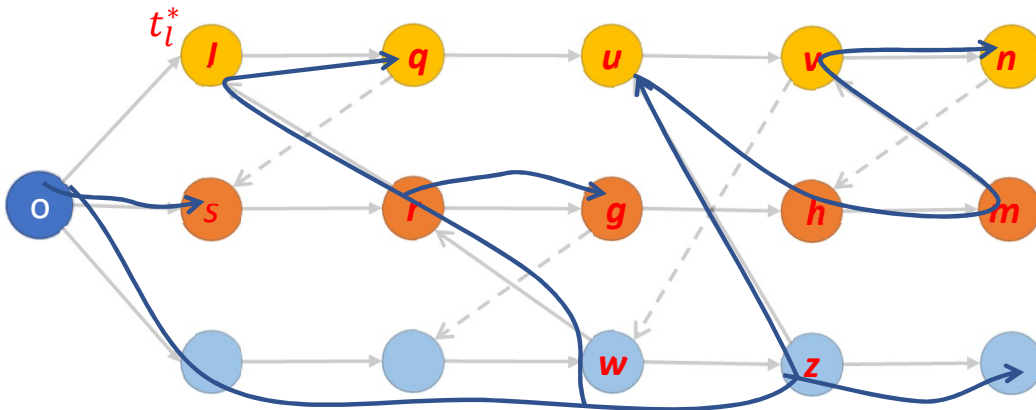
$t_v - t_u \geq l_{uv} \quad (u, v) \in A(\bar{y})$

$t \in \mathbb{R}^V$

Optimal solutions

□ $\bar{y} \in Y$, $G(\bar{y})$ no positive dicycles. For $u \in V$, $\bar{t}_u^* =$ length of longest ou -path in $G(\bar{y})$

If $f(t)$ is non-decreasing then \bar{t}^* is an optimal solution for $\text{Sched}(\bar{y})$



$\text{Sched}(\bar{y})$

min $f(t)$

$t_v - t_u \geq l_{uv} \quad (u, v) \in A(\bar{y})$

$t \in \mathbb{R}^V$

A reformulation for the TRS problem

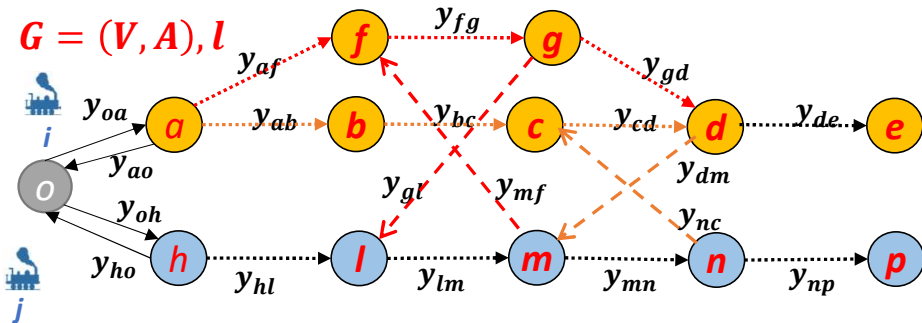
- Assuming $f(t)$ is non-decreasing, we can restate TRS as:

Find $y^ \in Y$, such that $G(y^*)$ has no positive directed cycles and the cost $f(t^*)$ of the schedule t^* associated with the longest path tree H^* in $G(y^*)$ is minimum.*

- This also leads to a non-compact, non-big-M reformulation (*Path&Cycle*, Lamorgese and M., 2019)
- The variables are again the arc variables y . But the constraints are associated with the positive directed cycles and the longest path trees of $G(V, A), l$.

Solving instances of TRS

Solving TRS by Branch & Bound



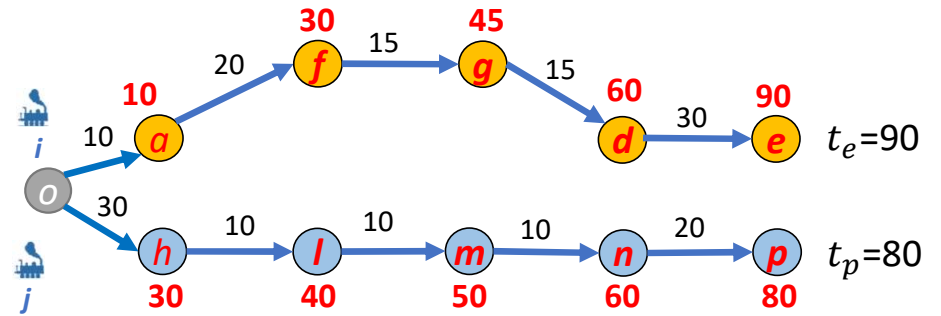
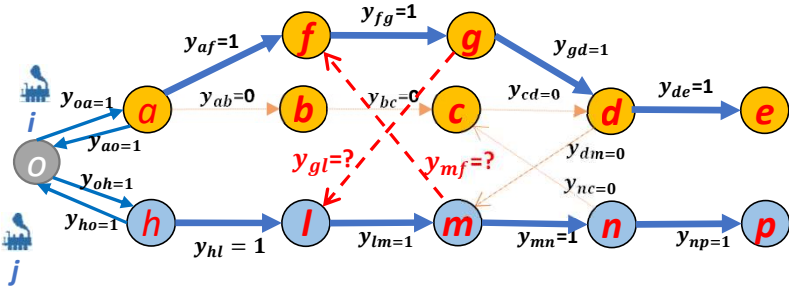
$$\begin{aligned} \min \quad & f(t) \\ & t_v - t_u \geq l_{uv} - (1 - y_{uv})M \quad (u, v) \in A \\ & t \in \mathbb{R}^V, y \in Y \subseteq \{0, 1\}^A \end{aligned}$$

- ❑ We assume $f(t)$ **non-decreasing**, e.g. linear $c^T t$, with $c \geq 0$
- ❑ MIP formulations can be solved by B&B, either ad-hoc or using off-the-shelf MIP-solvers
- ❑ In any node q of the B&B tree we have some y variables fixed to 0 or 1. I.e. we have selected $A^q \subseteq A$ arcs and excluded some other arcs.
- ❑ The associated MIP contains some standard precedence constraints and some big-M constraints. Solving linear relaxation = neglecting all residual big-M constraints.
- ❑ Computing the linear relaxation amounts to compute a longest path tree in $G(A^q)$

Into the branching node

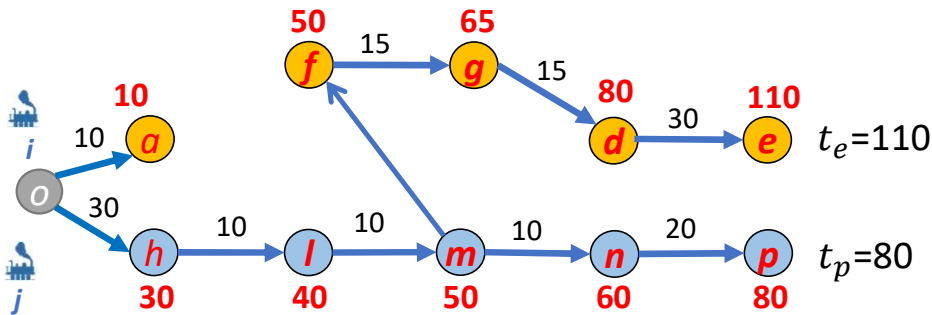
$$G^q = (V^q, A^q), l^q$$

Branch node q : Blue arcs selected, red arcs still free



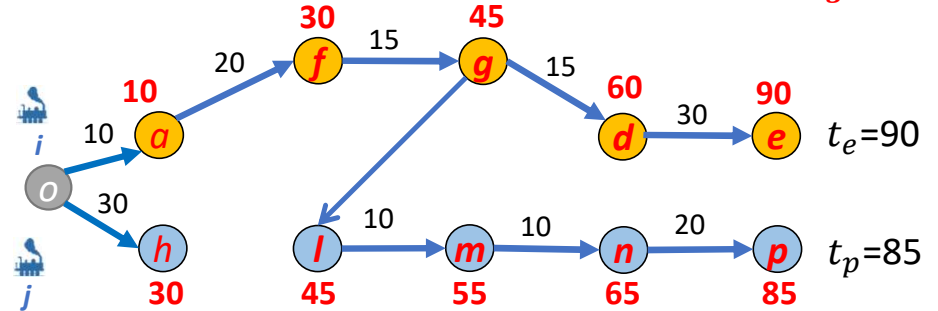
New branch on conflict $y_{mf} + y_{gl} = 1$

Child q_1 , select $(m, f), y_{mf}=1$



Longest path tree. Cost = $t_p + t_e = 190$

Child q_2 , select $(g, l), y_{gl}=1$



Longest path tree. Cost = $t_p + t_e = 175$

Incremental longest path tree

- ❑ Binary variables are associated with conflict- or routing-decisions (timetable arcs are fixed)
- ❑ At each branching node we either branch on a conflict or on a routing decision
- ❑ In either case, we add to the parent graph one or few arcs
- ❑ A positive cycle identification (infeasibility) or a partial feasible schedule with associated lower bound is computed effectively by **incremental longest path computations**.
- ❑ We extended previous results for incremental topological ordering (Haupler et al, 2011)

Is B&B good enough?

A modest example

An instance with **6 trains** (submitted to MIPLIB)

- TRS graph G contains 1954 nodes
- On average 213 alternative routes per train

Natural MIP model

- 1900 continuous variables (schedule t)
- 23000 binary variables (arc variables y)
- 61000 constraints, 300000 non-zeros

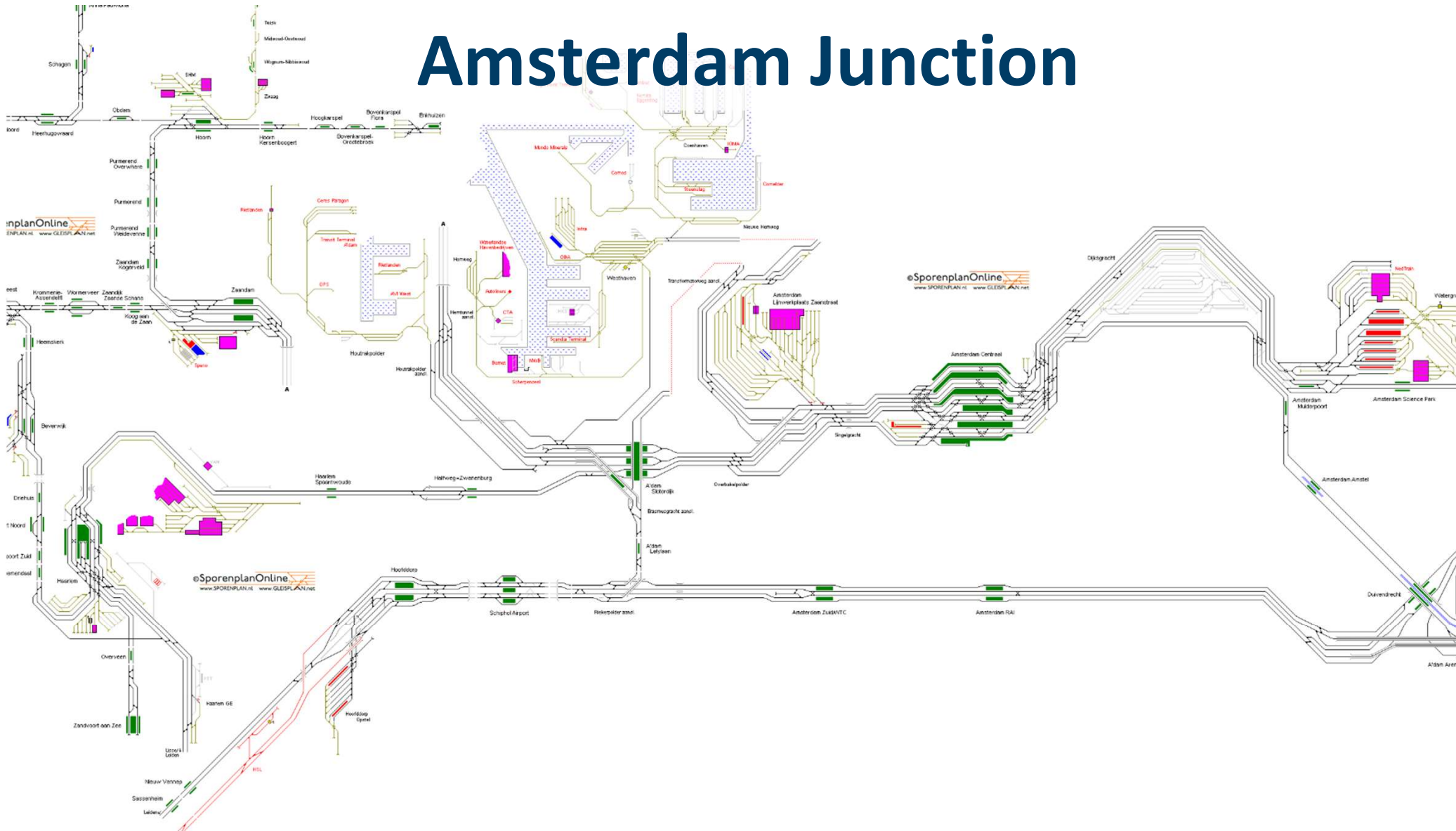
After 10 minutes computation Gap 100%.
(Gurobi 9.5, default settings, 12 cores)

In practice, we have 22 seconds ...



Decomposition

Amsterdam Junction



Logic Benders' reformulation

- We adopt a more general version of classical Benders' decomposition

$$\min f(x)$$

$$Ax \leq b$$

$$Bx + Cw \leq e$$

$$x \in X, w \in W$$

$$X \subseteq Z^{q_x} \times R^{s_x}, W \subseteq Z^{q_w} \times R^{s_w}$$

both x and w are mixed integer vectors

- Objective and a subset of constraints not involving w

- The target is to project out w variables and reformulate the problem as

$$\min f(x)$$

$$Ax \leq b$$

$$Dx \leq d \text{ feasibility cuts}$$

$$x \in X$$

- Is this convenient in our case?
- How to choose x, w ?
- How to generate the feasibility cuts

Master/worker decomposition

$$\begin{array}{l}
 \min f(x) \\
 Ax \leq b \\
 Bx + Cw \leq e \\
 x \in X, w \in W
 \end{array}
 \quad (\text{TRS})$$



$$\begin{array}{l}
 \min f(x) \\
 Ax \leq b \\
 Dx \leq d \quad \text{feasibility cuts} \\
 x \in X
 \end{array}
 \quad (P)$$

Constructed iteratively by row generation

1. Solve reformulation with subset $D^k x \leq d^k$ of feasibility cuts (**restricted master P^k**)
2. If P^k infeasible **STOP (TRS infeas)**. Otherwise let x^k optimal solution to P^k
3. If x^k can be extended to a solution (x^k, w^k) of TRS, **STOP** ((x^k, w^k) is optimal)
4. Else identify valid constraint $\delta^k x \leq \Delta^k$ violated by x^k . Generate $D^{k+1} x \leq d^{k+1}, P^{k+1}$
5. Iterate

Points 3., 4. and 5. is called **worker (sub)problem**:

Does there exist $w^k \in W$ such that $Cw^k \leq e - Bx^k$?

$$\begin{array}{l}
 \min f(x) \\
 Ax \leq b \\
 D^k x \leq d^k \\
 x \in X
 \end{array}
 \quad (P^k)$$



$$\delta^k x^k > \Delta^k$$

$$\begin{array}{l}
 \min f(x) \\
 Ax \leq b \\
 D^k x \leq d^k \\
 \delta^k x \leq \Delta^k \\
 x \in X
 \end{array}
 \quad (P^{k+1})$$

How to choose x and w

$$\begin{aligned} \min \quad & f(x) \\ & Ax \leq b \\ & Bx + Cw \leq e \\ & x \in X, w \in W \end{aligned}$$

$$\left(\begin{array}{cccc|cccc} & \mathbf{x} & & & & \mathbf{w} & & & \\ & \mathbf{A} & & & & \mathbf{0} & & & \\ \hline B^1 & 0 & \dots & 0 & C^1 & 0 & \dots & 0 \\ 0 & B^2 & \dots & 0 & 0 & C^2 & \dots & 0 \\ 0 & \ddots & 0 & 0 & 0 & \ddots & 0 & 0 \\ 0 & \dots & 0 & B^q & 0 & \dots & 0 & C^q \end{array} \right)$$

worker problem

Does there exist $w^k \in W$ such that $Cw^k \leq e - Bx^k$?

C decomposes into several, non-overlapping and small blocks

→ the worker problem decomposes into independent subproblems, solved individually

→ Workers solve quickly

B decomposes into small, non-overlapping blocks, each corresponding to a C -block

→ each worker subproblem “shares” few variables with master program

→ Strong infeasibility cuts

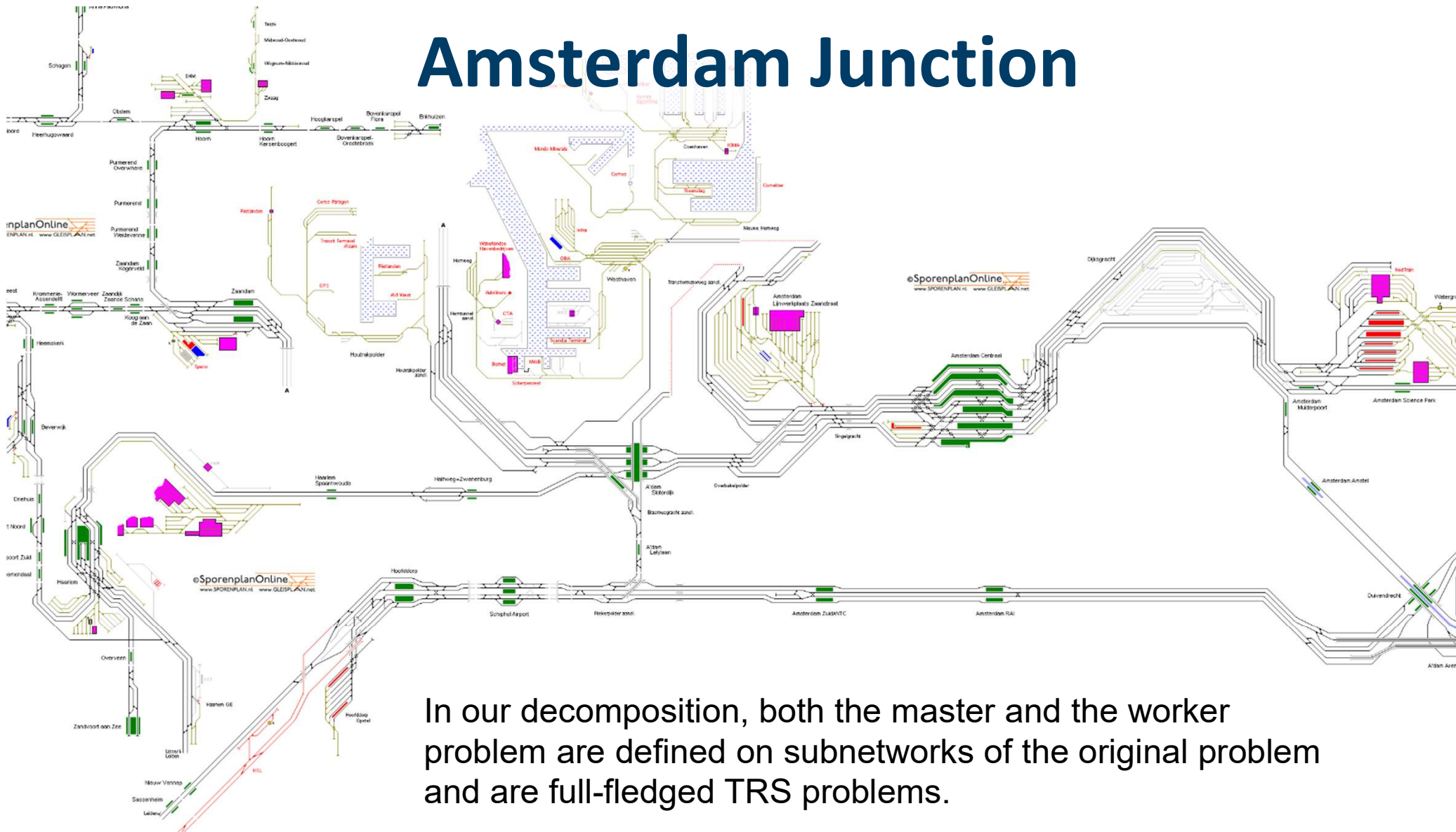
A small

→ the master program is small

→ Master solves quickly

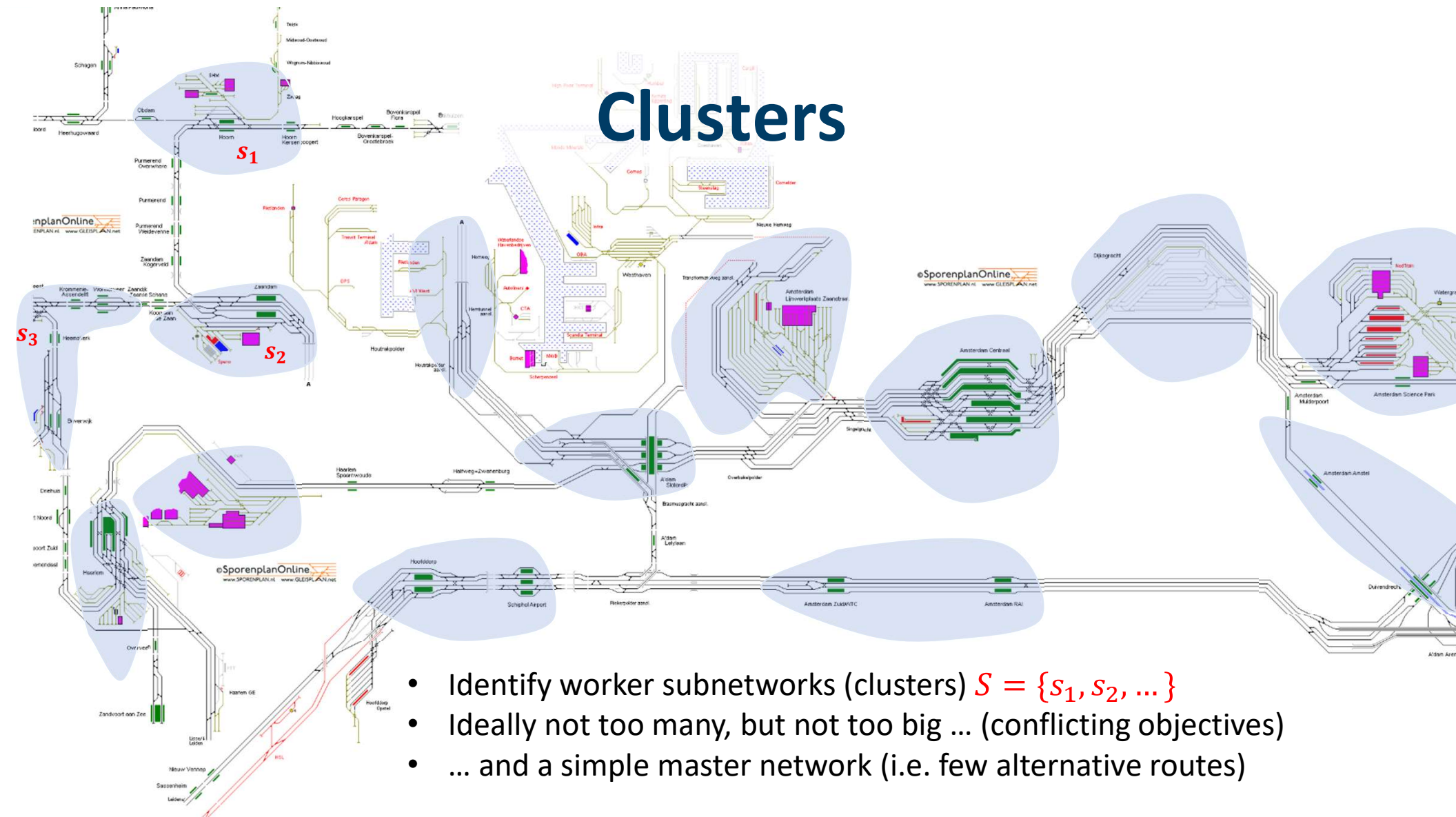
We identify x and w , thus A, B, C, b, e by **spatial decomposition**

Amsterdam Junction



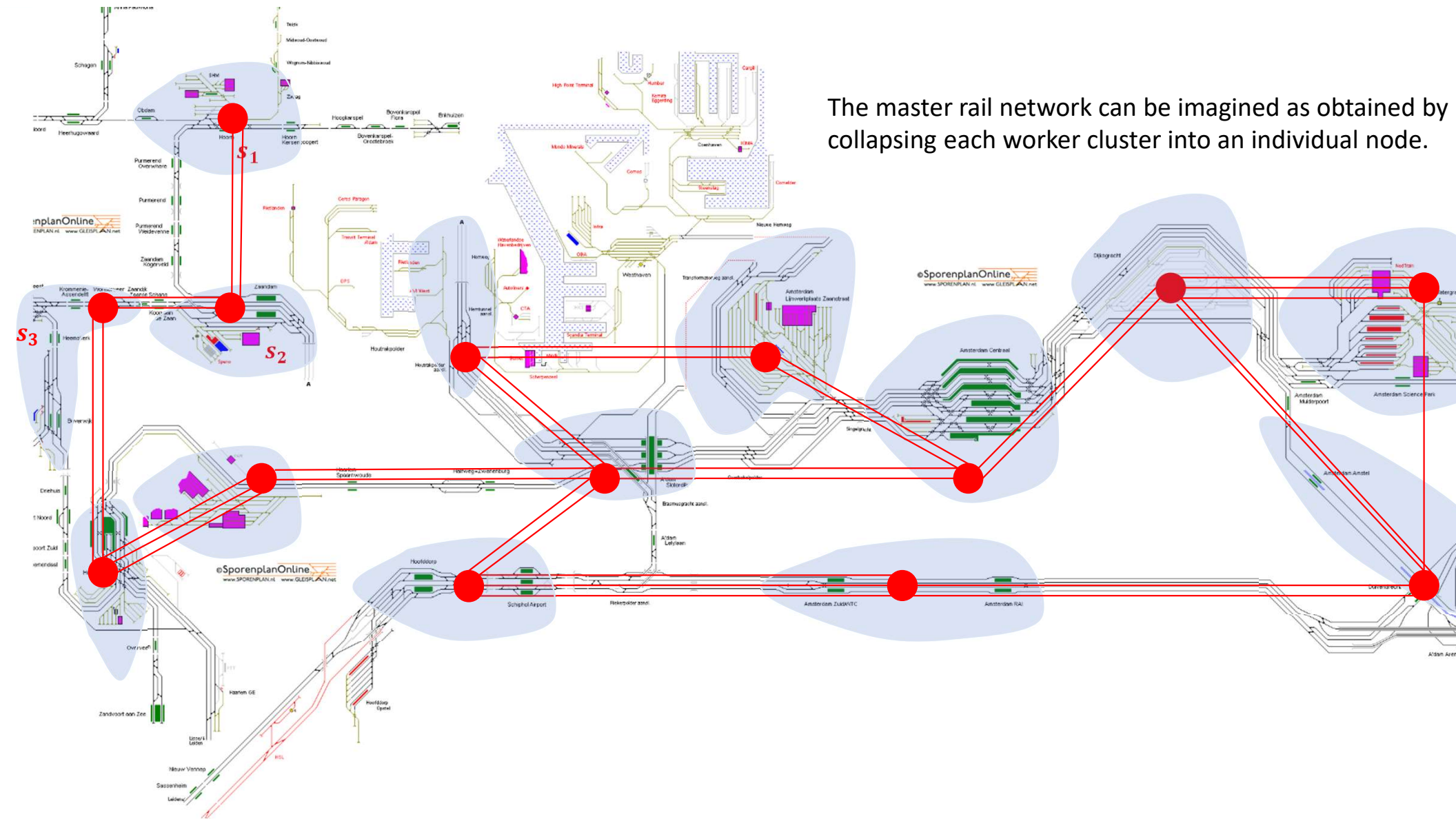
In our decomposition, both the master and the worker problem are defined on subnetworks of the original problem and are full-fledged TRS problems.

Clusters



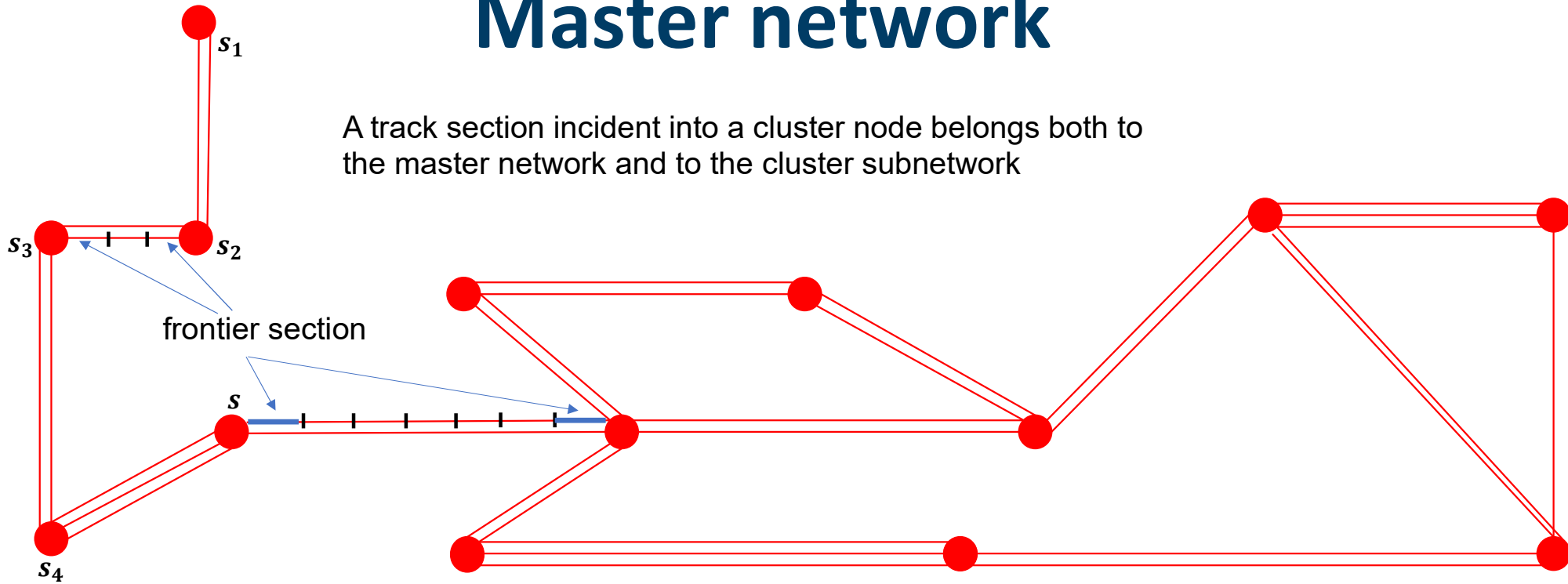
- Identify worker subnetworks (clusters) $S = \{s_1, s_2, \dots\}$
- Ideally not too many, but not too big ... (conflicting objectives)
- ... and a simple master network (i.e. few alternative routes)

The master rail network can be imagined as obtained by collapsing each worker cluster into an individual node.



Master network

A track section incident into a cluster node belongs both to the master network and to the cluster subnetwork

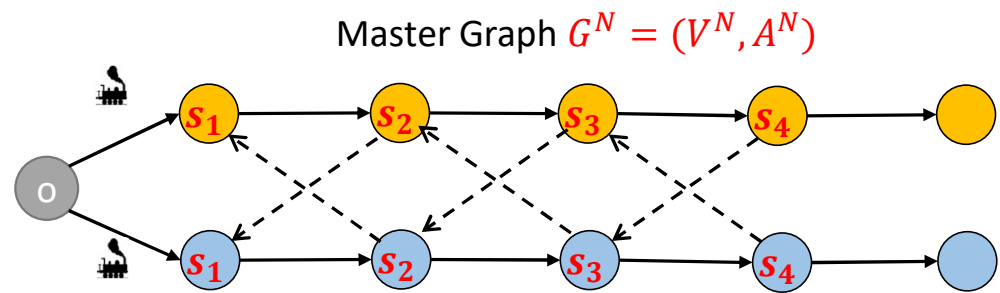
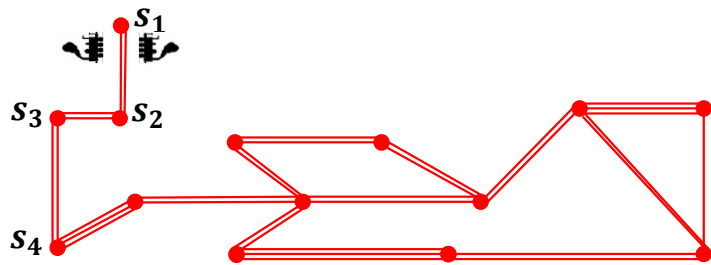


A schedule variable associated with a train and such frontier section belongs both to the master and to the worker

Spatial decomposition: master

- The master problem amounts to solving TRS on a collapsed railway network G^N

$$\begin{aligned} \min \quad & f(t) \\ & t_v^N - t_u^N \geq l_{uv}^N - (1 - y_{uv}^N)M \quad (u, v) \in A^N \\ & \dots \\ & \text{Feasibility cuts} \\ & \dots \\ & t^N \in \mathbb{R}^{V^N}, y^N \in Y^N \subseteq \{0, 1\}^{A^N} \end{aligned}$$

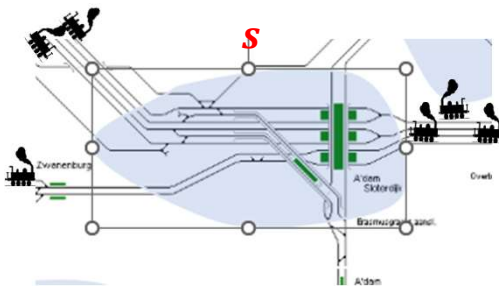


- t^N : master schedule solution
- For each cluster and each train, t^N provides arrival and departure times in the cluster
- t_s^N master (tentative) timetable for cluster s

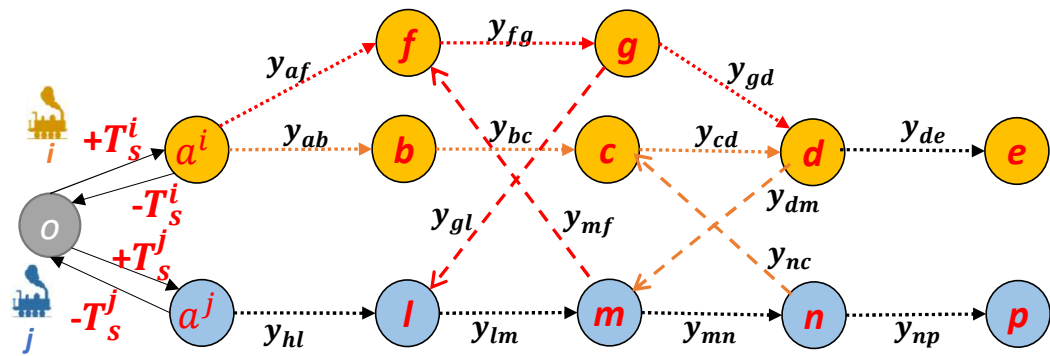
Worker subproblem s

Worker subproblem: Is the master timetable T feasible for (each) cluster s ?

T_s^i arrival time of train i in cluster s in current master solution (here we neglect departure times)



Cluster s



Worker TRS graph associated with s and master timetable T

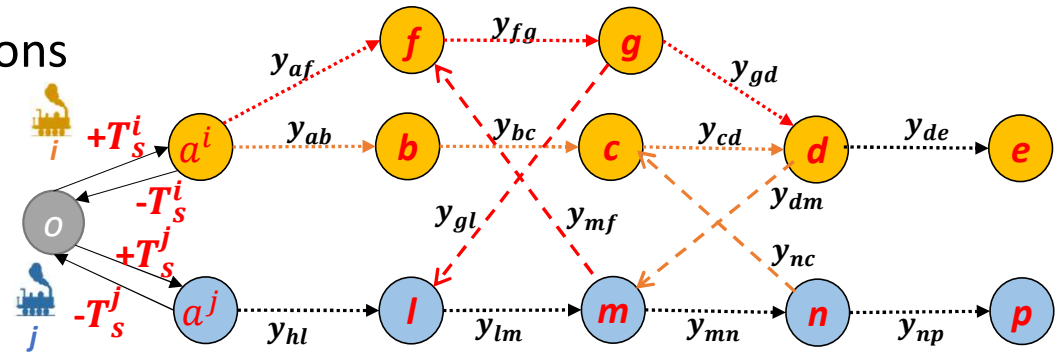
If worker infeasible, what is the feasibility cut violated by the current master timetable T ?

Cycle Obstructions and feasibility cuts

Cycle obstructions

TRS Instance $G = (V, A), l, Y$ feasible selections

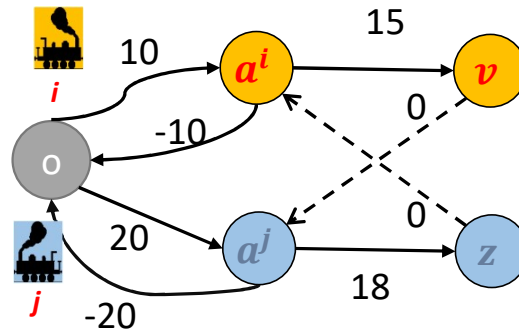
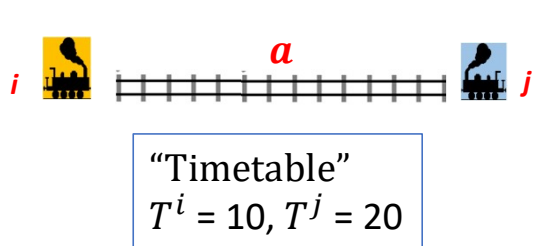
- Let $\bar{y} \in Y$. If $G(\bar{y})$ contains a positive dicycle, then $\text{Sched}(\bar{y})$ is infeasible.



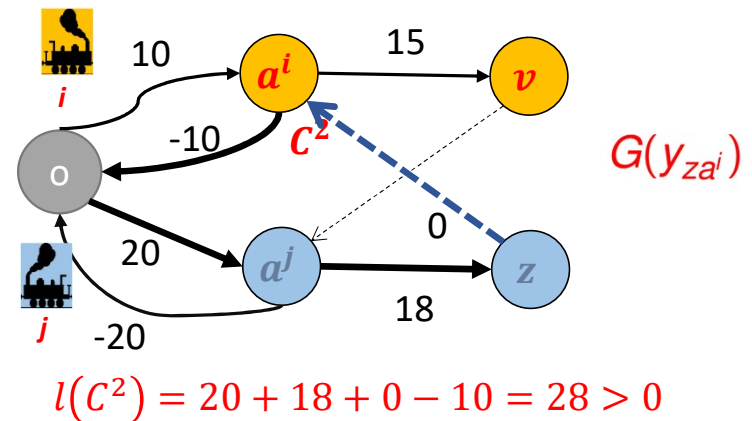
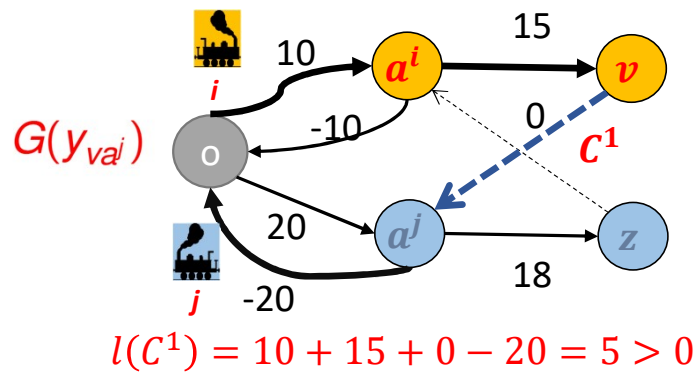
An instance $G = (V, A), l$ of TRS is *infeasible* iff, for all $\bar{y} \in Y$, $G(\bar{y})$ has a positive dicycle

- If TRS is infeasible, then there exists a family $\bar{\Omega} = \{C^1, C^2, \dots\}$ of **positive dicycles** of G, l such that for every selection $\bar{y} \in Y, G(\bar{y})$ contains (at least) a cycle of $\bar{\Omega}$.
- We call $\bar{\Omega} = \{C^1, C^2, \dots\}$ a **cycle obstruction** (for $G = (V, A), l$)

Example of cycle obstruction



Every solution contains
either (v, a^j) or (z, a^i)
 (every other arc is fixed)

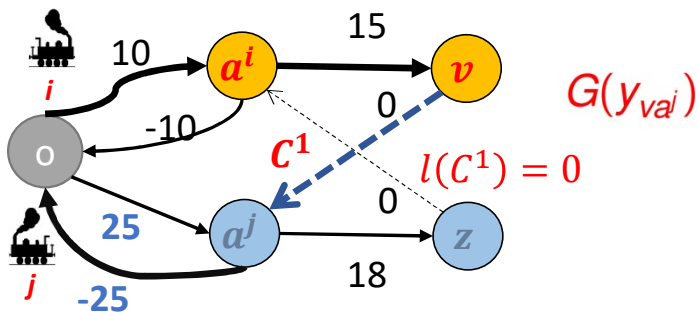


cycle obstruction $\bar{\Omega} = \{C^1, C^2\}$

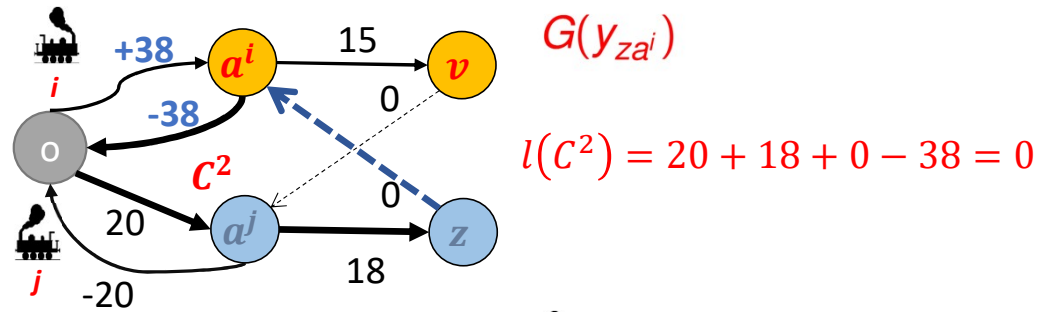
Avoiding cycle obstructions

- Cycle obstructions depend on the lengths l of the arcs of the instance $G = (V, A), l$
- In the example, if we modify the timetable, we can get rid of some positive cycles

New arrival "Timetable"
 $T^i = 10, T^j = 20$

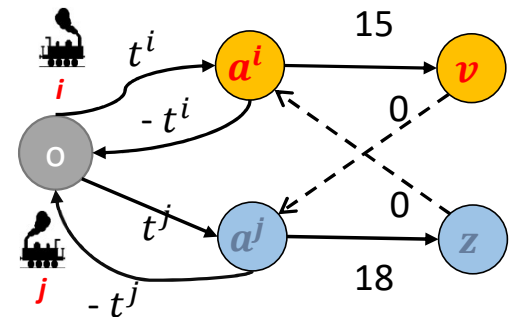


New arrival "Timetable"
 $T^i = 38, T^j = 20$



Any feasible (arrival) timetable t must satisfy one of two constraints:

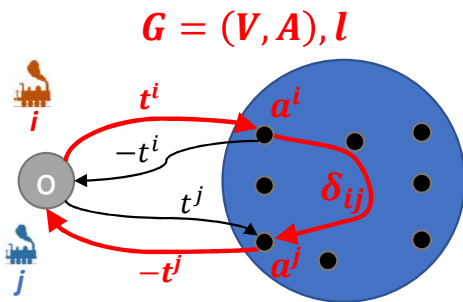
$$t_a^j - t_a^i \geq 15 \quad \vee \quad t_a^i - t_a^j \geq 18$$



Preventing positive timetable cycles

□ To prevent a positive «arrival» cycle to appear, we can constraint the arrival times of the trains involved

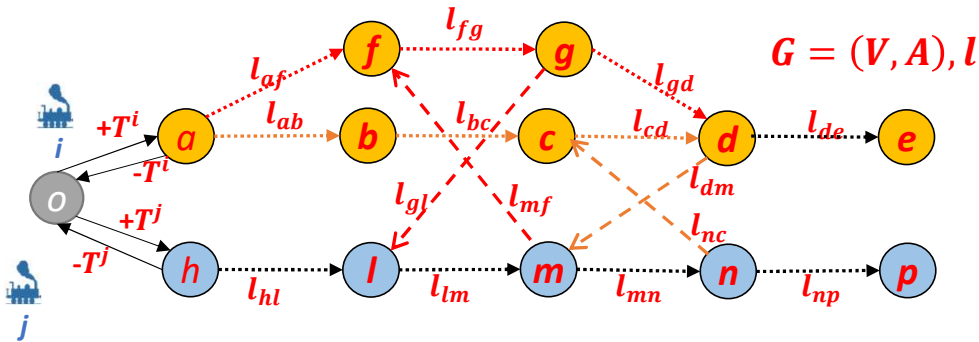
- t^i, t^j arrival times of train i, j
- a^i, a^j arrival nodes of train i, j
- δ_{ij} length of a maximum length path P_{ij} from a^i to a^j in $G = (V, A), l$



The directed cycle $(o, a^i), P_{ij}, (a^j, o)$ has **non-positive length** if

$$t^i - t^j + \delta_{ij} \leq 0 \quad \Rightarrow \quad t^j - t^i \geq \delta_{ij} \quad \text{time-precedence constraint}$$

Feasibility of timetables

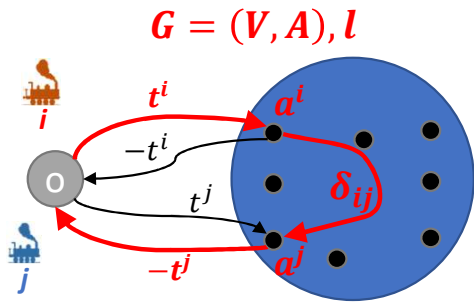


Timetable $\tau = T^1, T^2$

$\Omega = \{C^1, C^2, \dots\}$ cycle obstruction

- If a timetable $\tau = T^1, T^2, \dots$ is infeasible then there exists (at least) a cycle obstruction $\Omega = \{C^1, C^2, \dots, C^q\}$ associated with τ .
- We are only interested in cycles containing timetable arcs, i.e. through the origin.
- We may assume timetable cycles are simple, and Ω is minimal.
- WANT: find a timetable $\Delta = t^1, t^2, \dots$ so that Ω is not a cycle obstruction of Δ .
- At least one of the cycles in Ω should be non-positive w.r.t. Δ .

Disjunctive feasibility cut



Non “positiveness”
constraint

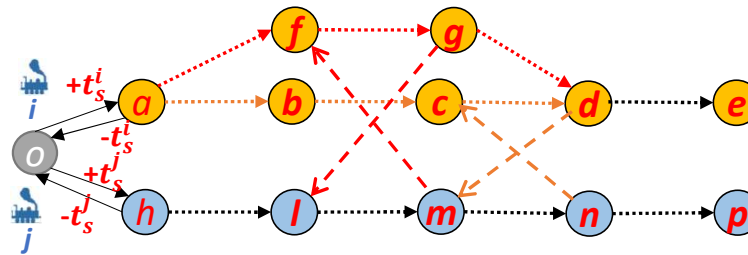
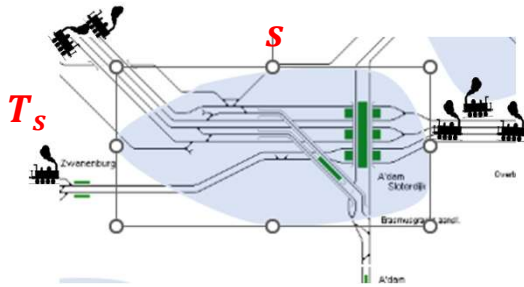
$$t^j - t^i \geq \delta_{ij}$$

- Cycle obstruction $\Omega = \{C^1, C^2, \dots, C^q\}$ timetable cycles of Ω
- Each cycle only involves two timetable arcs, associated with two trains
- $(i_1, j_1), (i_2, j_2), \dots, (i_q, j_q)$ pairs of trains involved in the obstruction
- OBS: With n trains, the number q of cycles of Ω is at most $n \times (n - 1)$
- Let $\Delta = t^1, t^2, \dots, t^n$ be the arrival time of train $1, \dots, n$
- Since at least one cycle of Ω must be non-positive in a feasible schedule, then the next **k-disjunctive constraint** is valid (Leutwiler & Corman 2023)

$$t^{j_1} - t^{i_1} \geq \delta_1 \vee t^{j_2} - t^{i_2} \geq \delta_2 \vee \dots \vee t^{j_q} - t^{i_q} \geq \delta_q$$

Worker subproblem

- Is the current master timetable t_k infeasible for some cluster $s \in S$?



- Worker s Infeasible: there exists a cycle obstruction $\Omega^s = \{C^1, C^2, \dots, C^q\}$ for t_k .
- To avoid at least one of such positive cycles, **the master timetable** must satisfy a q -disjunctive constraint

$$t_s^{j_1} - t_s^{i_1} \geq \delta_1 \vee t_s^{j_2} - t_s^{i_2} \geq \delta_2 \vee \dots \vee t_s^{j_k} - t_s^{i_k} \geq \delta_k$$

Remark: each term is a time-precedence constraint!

Master program: feasibility cuts

Each feasibility cut: q -disjunction of precedence constraints.

$$\begin{array}{ll}
 \min & f(t) \\
 & t_v^N - t_u^N \geq l_{uv}^N - (1 - y_{uv}^N)M \quad (u, v) \in A^N \\
 & \dots \\
 & t_{j_1}^N - t_{i_1}^N \geq \delta_1 \vee \dots \vee t_{j_q}^N - t_{i_q}^N \geq \delta_q \quad \Omega = \{C_1, \dots, C_q\} \in \mathcal{O} \\
 & \dots \\
 & t^N \in \mathbb{R}^{V^N}, y^N \in Y^N \subseteq \{0, 1\}^{A^N}
 \end{array}
 \quad
 \left\{ \begin{array}{l}
 t_{j_1}^N - t_{i_1}^N \geq \delta_1 - (1 - y_{i_1 j_1}^\Omega)M \\
 \dots \\
 t_{j_q}^N - t_{i_q}^N \geq \delta_q - (1 - y_{i_q j_q}^\Omega)M \\
 y_{i_1 j_1}^\Omega + \dots + y_{i_q j_q}^\Omega = 1
 \end{array} \right. \quad \Omega \in \mathcal{O}$$

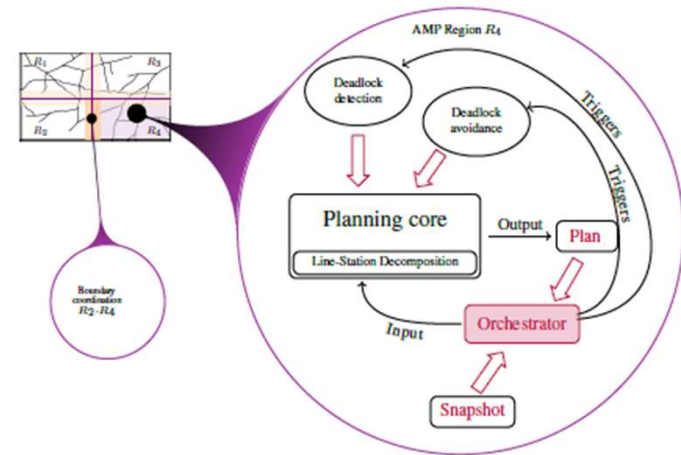
The “nature” of the formulation is unchanged after including the feasibility cuts

Include in G the “cycle” arcs $A^{\mathcal{O}} = \{(i_1, j_1)^\Omega, \dots, (i_q, j_q)^\Omega, \Omega \in \mathcal{O}\}$, the associated y variables and constraints

$$\begin{array}{ll}
 \min & f(t) \\
 & t_v - t_u \geq l_{uv} - (1 - y_{uv})M \quad (u, v) \in A^N \cup A^{\mathcal{O}} \\
 & t \in \mathbb{R}^V, y \in Y^N \cap Y^{\mathcal{O}} \subseteq \{0, 1\}^{A^N \cup A^{\mathcal{O}}}
 \end{array}$$

Conclusions

The automatic dispatching system was deployed in all 12 administrative regions of UP network in the fall 2021.

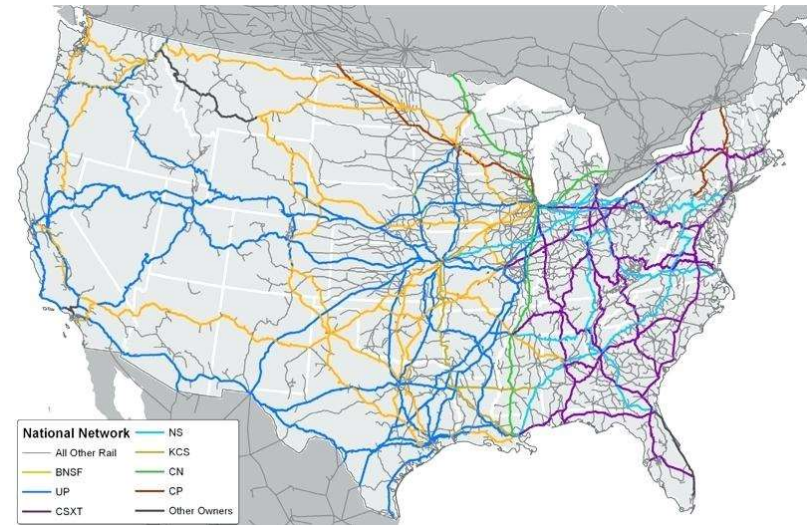


It required the development of many communicating algorithms and software components

Conclusions

Since then, we started new challenging projects with UP:

- **Block design:** how to group commodities and send them from origin to destination
- **Train design:** how to group the blocks and form trains
- **Deadlock detection**
- **Network decomposition**



Joint effort: Annese, Boccia, Dal Sasso, Lamorgese, Luteberget, **M.**, Onofri, **Oriolo**, Piacentini, Russo Russo, **Sartor**, **Ventura** (and others)

Very complex problems, very large instances asking for the development of new combinatorial optimization and integer programming methods.

- Lamorgese, L., & Mannino, C. (2015). An exact decomposition approach for the real-time train dispatching problem. *Operations Research*, 63(1), 48-64.
- Lamorgese, L., & Mannino, C. (2019). A noncompact formulation for job-shop scheduling problems in traffic management. *Operations Research*, 67(6), 1586-1609.
- Leutwiler, F., Corman, F. (2021). A logic Benders' decomposition for microscopic railway timetable planning, *EURO 2021 conference*, Athens, July 2021.
- Mascis, A., & Pacciarelli, D. (2002). Job-shop scheduling with blocking and no-wait constraints. *European Journal of Operational Research*, 143(3), 498-517.
- Haeupler, B., Kavitha, T., Mathew, R., Sen, S. and Tarjan, R.E., 2012. Incremental cycle detection, topological ordering, and strong component maintenance. *ACM Transactions on Algorithms (TALG)*, 8(1), pp.1-33.