

Delay-Robust Event-Scheduling *In Memoriam* of A. Caprara

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Delay-Robust Event-Scheduling
Robustness
Framework

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Train Platforming Problem

- In and Out

- TPP deterministic model

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- Buffers linking constraints

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Cons

Strict robustness is generally overconservative, because:

- ▶ solutions must cope with every likely scenarios without any recovery

Recoverable Robustness

Informally speaking, a solution to an optimization problem is called *recovery robust* if it can be adjusted to all likely scenarios by limited recovery action. Thus a recovery-robust solution provides a service guarantee (Liebchen *et al.* 2009 [3], Stiller 2008 [4]).

Robust Network Buffering

We are interested in the special case in which the recovery problem is a delay (d_e) propagation in some directed graph $N = (\mathcal{E}, A)$, which is buffered on the arcs against disturbances on the nodes $e \in \mathcal{E}$.

Details in Liebchen *et al.* 2009 [3], Stiller 2008 [4].

A framework

- ▶ A set \mathcal{E} of *events* to be scheduled.
- ▶ Denoting by F the set of *feasible* schedulings of events \mathcal{E} , i.e., the feasible region, a *nominal problem* of the form $\min\{c(x) : x \in F\}$.
- ▶ A set \mathcal{S} of possible *scenarios*, where each scenario $s \in \mathcal{S}$ is defined by the external *disturbance* $\delta_e^s \geq 0$ assigned to each event $e \in \mathcal{E}$.
- ▶ A *delay-propagation network* $N = (\mathcal{E}, A)$, with $(e', e) \in A$ if a delay $d_{e'}$ on event e' may propagate to a delay d_e on event e .
- ▶ A delay $d_{e'}$ on event e' may propagate to a delay d_e on event e according to the relation $d_e \geq d_{e'} - b((e', e), x)$, where $b((e', e), x)$ is a *buffer time* function.

Mathematical model

The delay-robust problem can be formulated as:

$$\min c(x) + D, \quad (1)$$

subject to

$$x \in F, \quad (2)$$

$$D \geq \sum_{e \in \mathcal{E}} d_e^s, \quad s \in \mathcal{S}, \quad (3)$$

$$d_e^s \geq \delta_e^s, \quad e \in \mathcal{E}, s \in \mathcal{S}, \quad (4)$$

$$d_e^s \geq d_{e'}^s - f_{(e',e)}, \quad (e',e) \in A, s \in \mathcal{S}, \quad (5)$$

$$f_{(e',e)} = b((e',e), x), \quad (e',e) \in A. \quad (6)$$

Scenario set

$$\{\delta^s \in \mathbb{R}_+^{|\mathcal{E}|} : \|\delta^s\|_1 \leq \Delta\} \quad (7)$$

These are *uncountably* many scenarios, which can however be handled easily thanks to the following observation.

Proposition

For the delay-robust problem (1)-(6), the compact scenario set defined by (7) is equivalent to the finite scenario set S defined by the vectors

$$\{\delta^s \in \{0, \Delta\}^{|\mathcal{E}|} : \|\delta^s\|_1 = \Delta\},$$

which contains only $|\mathcal{E}|$ scenarios.

Quadratic buffer time function

In the most natural case, the buffer time $b((e', e), x)$ only depends on the way in which events e' and e are scheduled in x , and is independent of the scheduling of the other events:

$$f_{(e', e)} = \sum_{h' \in \mathcal{H}_{e'}} \sum_{h \in \mathcal{H}_e} \varphi(e', h', e, h) x_{e', h'} x_{e, h}, \quad (e', e) \in A.$$

These quadratic constraints can be linearised in a few ways.

Projecting out delay variables

One may directly optimise over the feasible region obtained by projecting out the d_e^s variables (and removing the constraints (3)-(5)).

Let $\sigma_{\bar{f}}(s, \bar{e})$ be the value of the shortest path from s to $\bar{e} \in \mathcal{E}$ on the delay propagation network $N = (\mathcal{E}, A)$ with arc lengths $\bar{f}_{(e', e)}$ for $(e', e) \in A$. Moreover, let $P_{\bar{f}}(s, \bar{e}) \subseteq A$ denote such a shortest path.

Lemma

Given buffer time values $\bar{f}_{(e', e)}$, an external disturbance equal to Δ on event $s \in \mathcal{E}$ (and null for the other events) propagates to a delay of value $\max\{0, \Delta - \sigma_{\bar{f}}(s, \bar{e})\}$ on event $\bar{e} \in \mathcal{E}$.

Projecting out delay variables (cont.)

A direct consequence is:

Lemma

Given buffer time values $\bar{f}_{(e',e)}$ and maximum cumulative delay \bar{D} , for every scenario $s \in \mathcal{S}$ there exist values of the d_e^s variables satisfying (3)-(5) if and only if $\sum_{e \in \mathcal{E}} \max\{0, \Delta - \sigma_{\bar{f}}(s, e)\} \leq \bar{D}$. Otherwise, for the considered scenario s , letting $\mathcal{E}_{\bar{f}}(s) \subseteq \mathcal{E}$ be the set of events such that $\Delta - \sigma_{\bar{f}}(s, e) > 0$, a valid inequality that is violated by $\bar{f}_{(e',e)}$ and \bar{D} is

$$\sum_{e \in \mathcal{E}_{\bar{f}}(s)} \sum_{a \in P_{\bar{f}}(s,e)} f_a \geq \Delta |\mathcal{E}_{\bar{f}}(s)| - D.$$

Problem definition

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- ▶ for a specific railway station
- ▶ after the timetable has been defined

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- ▶ Train schedule : arrival, departure times, directions and allowed shifts
- ▶ Railway station topology: platforms, paths and directions

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Output

- ▶ Assign each train a platform and two paths for arrival and departure s.t. no operational constraint is violated

Train schedule

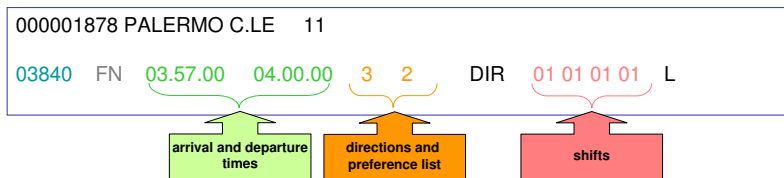


Figure: Train Schedule

The train schedule of a railway station contains info on arrival and departure times, directions and allowed shifts of each train passing through it.

Railway station topology

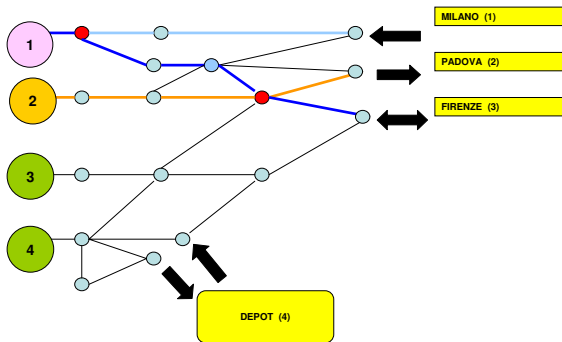


Figure: Topology

The topology of a railway station includes platforms, paths and directions.

Resources and operational constraints

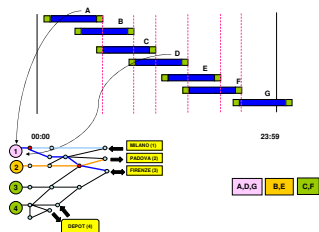


Figure: Platform and path.

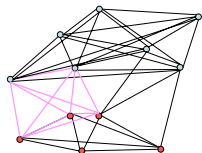


Figure: Incompatibility graph.

Platform conflicts are forbidden, path conflicts are allowed to some extent.

A pattern P for a train t is a 5-tuple defining: platform, arrival/departure paths and shifts. Operational constraints can be expressed using an incompatibility graph among patterns.

TPP deterministic model

$$\min \sum_{t \in T} \sum_{P \in \mathcal{P}_t} c_{t,P} x_{t,P}$$

s.t.

$$\sum_{P \in \mathcal{P}_t} x_{t,P} = 1, \quad t \in T$$

$$\sum_{(t,P) \in K} x_{t,P} \leq 1, \quad K \in \mathcal{K}$$

$$x_{t,P} \in \{0, 1\}, \quad t \in T, P \in \mathcal{P}_t$$

Details in Caprara *et al.* 2011 [1].

Delay propagation network

The platforming gives rise to a network in which the delay caused by disturbances propagates.

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Hence, the following two events are associated with each train $t \in T$ and define \mathcal{E} :

- ▶ *arrival* a_t : the train occupies a given arrival path at a given instant
- ▶ *departure* p_t : the train frees its platform at a given instant

Example

Assume that in the nominal solution trains t' and t stop at the same platform, event $p_{t'}$ is scheduled at 10:00, and event a_t at 10:04 with 3 minutes to travel along the arrival path, so that t occupies the platform at 10:07. Moreover, assume that the headway time that must elapse between t' freeing the platform and t occupying it is 2 minutes and that the 3-minute travel time for t along its arrival path is fixed. If the departure path of t' and the arrival path of t are compatible, the buffer time $f_{(p_{t'}, a_t)}$ is equal to 5 minutes, as a delay of up to 5 minutes on $p_{t'}$ does not affect a_t , whereas a larger one does.

Delay-robust mathematical model

$$\min \sum_{t \in T} \sum_{P \in \mathcal{P}_t} c_{t,P} x_{t,P} + D$$

subject to

$$\sum_{P \in \mathcal{P}_t} x_{t,P} = 1, \quad t \in T,$$

$$\sum_{(t,P) \in K} x_{t,P} \leq 1, \quad K \in \mathcal{K},$$

$$x_{t,P} \in \{0, 1\}, \quad t \in T, P \in \mathcal{P}_t,$$

$$f_{(e',e)} = \sum_{P_1 \in \mathcal{P}_{t_1}} \sum_{P_2 \in \mathcal{P}_{t_2}} c_{P_1, P_2, a} x_{t_1, P_1} x_{t_2, P_2}, \quad (e', e) \in A,$$

$$D \geq \sum_{e \in \mathcal{E}} d_e^s, \quad s \in \mathcal{S},$$

$$d_e^s \geq \delta_e^s, \quad e \in \mathcal{E}, s \in \mathcal{S},$$

$$d_e^s \geq d_{e'}^s - f_{(e',e)}, \quad (e', e) \in A, s \in \mathcal{S}.$$

Buffers linking constraints

A straightforward link between the buffer value of a given arc $a \in A(N)$ associated with train pair $(t_1, t_2) \in T^2$ and the choice of patterns for the given pair of trains is the following:

$$\sum_{P_1 \in \mathcal{P}_{t_1}} \sum_{P_2 \in \mathcal{P}_{t_2}} c_{P_1, P_2, a} x_{t_1, P_1} x_{t_2, P_2}$$

where c_{a, P_1, P_2} is a constant associated to arc a and to the corresponding choice of patterns (P_1, P_2) for trains (t_1, t_2) .

Buffers linking constraints

$$f_a \leq \sum_{P_1 \in \mathcal{P}_{t_1}} \alpha_{P_1}^a x_{t_1, P_1} + \sum_{P_2 \in \mathcal{P}_{t_2}} \beta_{P_2}^a x_{t_2, P_2} - \gamma^a, \quad a \in A(N), (\alpha, \beta, \gamma) \in \mathcal{F}_a \quad (8)$$

Following Caprara *et al.* 2011 [1], the separation of Constraints (8) is done by a sort of polyhedral brute force, given that, for each pair of trains t_1, t_2 , and for each arc $a \in A(N)$ the number of vertices in $Q_{t_1, t_2, a}$ is small. Specifically, $Q_{t_1, t_2, a}$ has $|\mathcal{P}_{t_1}| |\mathcal{P}_{t_2}|$ vertices and lies in $\mathbb{R}^{|\mathcal{P}_{t_1}| + |\mathcal{P}_{t_2}| + 1}$, so we can separate over it by solving an LP with $|\mathcal{P}_{t_1}| |\mathcal{P}_{t_2}|$ variables and $|\mathcal{P}_{t_1}| + |\mathcal{P}_{t_2}| + 1$ constraints.

Computational results: Palermo C.Le.

instance	[1] Solution		Delay-Robust LP		Delay-Robust Solution			
	D	time	D	time	D	%gap	%reduction	time
PA I	1112	5	451.88	131	791	43%	29%	9788
PA II	694	5	252.00	49	584	57%	16%	3892
PA III	756	5	285.00	47	593	52%	22%	5411
PA IV	525	5	207.00	48	459	55%	13%	1417

Table: Results for the real-world instances of Rete Ferroviaria Italiana.

Computational results: Genova P.Principe

instance	[1] Solution		Delay-Robust LP		Delay-Robust Solution			
	D	time	D	time	D	%gap	%reduction	time
GE I	561	5	193.11	30	453	57%	19%	147
GE II	1252	5	388.00	201	810	52%	35%	30417
GE III	1073	5	297.95	118	865	66%	19%	1086
GE IV	758	5	231.00	54	498	54%	34%	274

Table: Results for the real-world instances of Rete Ferroviaria Italiana.

Computational results: Bari C.Le.

instance	[1] Solution		Delay-Robust LP		Delay-Robust Solution			
	D	time	D	time	D	%gap	%reduction	time
BA I	770	5	341.00	68	592	42%	23%	1095
BA II	959	5	508.08	118	959	47%	0%	5
BA III	711	5	293.91	115	651	55%	8%	525
BA IV	733	5	308.40	103	686	55%	6%	2431
BA V	786	5	245.26	165	734	67%	7%	3204
BA VI	819	5	312.00	51	819	62%	0%	5
BA VII	1025	5	315.18	77	743	58%	27%	4126
BA VIII	714	5	240.00	45	580	59%	19%	535

Table: Results for the real-world instances of Rete Ferroviaria Italiana.

Computational results: Milano C.Le.

instance	[1] Solution		Delay-Robust LP		Delay-Robust Solution			
	D	time	D	time	D	%gap	%reduction	time
MI I	480	5	354.00	105	476	26%	1%	201
MI II	785	5	449.28	330	686	35%	13%	2415
MI III	1102	6	382.88	334	822	53%	25%	938
MI IV	1067	5	721.00	839	951	24%	11%	2547
MI V	1002	8	290.76	288	565	49%	44%	2503
MI VI	898	5	415.13	655	816	49%	9%	618





Table: Results for the real-world instances of Rete Ferroviaria Italiana.

Questions



Thank you!

References

-  A. Caprara, L. Galli and P. Toth (2011). Solution of the Train Platforming Problem. *Tran. Sci.*, 45(2), 246–257.
-  A. Caprara, L. Galli and P. Toth (2012). Delay-Robust Event-Scheduling *Submitted*.
-  C. Liebchen, M. Lübbecke, R.H. Möhring, and S. Stiller (2009). The Concept of Recoverable Robustness, Linear Programming Recovery, and Railway Applications. In R.K. Ahuja, R.H. Möhring and C.D. Zaroliagis (eds.), *Robust and On-Line Large Scale Optimization*. Lecture Notes in Computer Science 5868, Springer-Verlag, 1–27.
-  S. Stiller (2008). *Extending Concepts of Reliability* PhD thesis, Institut für Mathematik, Technische Universität Berlin, Berlin, Germany.