Automation possibilities in metro-rail operations planning

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This talk will soon be available at www.ee.iitb.ac.in/%7Ebelur/talks www.ee.iitb.ac.in/%7Ebelur/railways/workshop www.ee.iitb.ac.in/%7Ebelur/railways

Group: Narayan Rangaraj and others from IIT Bombay narayan.rangaraj@iitb.ac.in

I-Metro-Workshop, Mar 2023, I-Metro workshop

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Automation in operations planning

- Need for software tools in operations planning
- Data availability/digitization: machine learning/artificial intelligence
- Tools recently developed by our group in railway operations
- Other tools (non-railway) developed by our group
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Vehicles flow models: flow rate (throughput) vs density

(Partly relevant for trains)



Source: Traffic Flow Theory, D.L. Gerlough and M.J. Huber (1975) Throughput peaks at an optimum density

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Automation in operations planning



Hard constraints:

- Headway: safety
- Frequency of service: demands: origin/destination services
- Traversal times
- Turn-around constraints: 'rake linking'
- Platform dwell/occupancy constraints

Soft constraints:

• Spacing between consecutive 'similar' services



Mumbai harbour-line

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Constraints tightest at CSTM ("VT")

Platform vacating constraints:

An exit (from platform 2) hinders consecutive two entries

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- Ideal: 'Scissor' crossing
- Fastest exit and least 'hindrance' to following entries
- Curves are on only a small distance: hence smaller distance of low-speed constraints
- Pretty expensive (but not exorbitant)





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Higher hindrance layout: at CSTM (due to curvature constraints: not displayed)

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- Lower layout: curve after bottleneck portion: slower exit

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Automation in operations planning

Inputs: different infrastructural parameters:

- Stations, network layout
- Passenger demands, traversal times
- Turnaround times at terminals

Output: timetable meeting all constraints with least number of rakes

Current Status: Mumbai Harbour line network

- 12 major stations
- 49 rakes in service in peak time 38 for Harbour and 11 for Trans Harbour

For a 3 hour peak timetable, about:

- 750 Departure-Arrival Events
- 40500 Headway Variables (Integer or Binary)
- 10000 Linkage Variables (Integer or Binary)
- 6500 Platform Variables (Integer or Binary)
- 3200 Precedence Variables for Platform Allocation (Binary)

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Case Study: Platform Allocation at CST

	ODD-Arrival-dep-events			EVEN-Arrival-dep-events	
Prev-Departure	Next-odd-Arrival	OccupiedFor	Prev-Departure	Next-even-Arrival	OccupiedFor
5:00	5:04	4.0 minutes	5:03	5:07	4.0 minutes
5:07	5:10	3.0 minutes	5:10	5:14	4.0 minutes
5:13	5:18	5.0 minutes	5:17	5:24	7.0 minutes
5:21	5:27	6.0 minutes	5:27	5:33.5	6.5 minutes
5:30	5:39	9.0 minutes	5:38	5:42	4.0 minutes
5:42	5:45	3.0 minutes	5:45	5:49	4.0 minutes
5:48.5	5:53	4.5 minutes	5:52	5:59	7.0 minutes
5:58	6:02	4.0 minutes	6:04	6:08	4.0 minutes
6:07	6:11	4.0 minutes	6:11	6:15	4.0 minutes
6:14	6:18	4.0 minutes	6:18	6:22	4.0 minutes
6:21	6:28	7.0 minutes	6:27	6:34	7.0 minutes
6:33	6:40	7.0 minutes	6:39	6:44	5.0 minutes
6:45	6:48	3.0 minutes	6:48	6:53	5.0 minutes
6:53	6:56	3.0 minutes	6:57	7:01	4.0 minutes

Case Study: Rakes in Operation vs time



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Case Study: Rake Cycles vs time



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Constrained programming based solvers:

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 - Helps to have spreadsheet based validators/checkers

(Though solvers have ensured satisfaction of constraints, one can validate by introducing 'test-flaws')

Periodic and Aperiodic Versions

Periodic Version:

- Tough integer programming problem: but fewer variables (just one period)
- Needs identical situations at end of every time period
- Very compact description of timetable
- Makes crew-scheduling easier

Aperiodic Version:

- Allows customization of services (peak 3-hour timetable need not have all service-counts as multiple of 3)
- Many more event arrival/departure variables

Have tried both versions in Mumbai suburban, but detailed case study is the aperiodic version

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Delhi Metro Train Frequency

Line 1 (Dilshad Garden to Rithala) Peak Hours: 3 min Lean Hours : Upto 12 min Line 2 (Jahangirpuri to HUDA City Cent) Peak Hours: 3 min Lean Hours : Upto 12 min Line 3 (Noida City Cent to Dwarka Sec. 21) Peak Hours: 3 min Lean Hours : Upto 12 min Line 4 (Yamuna Bank - Anand Vihar(ISBT)) Peak Hours : 6 min. Lean Hours : Upto 12 min. (Central Secretariat - Badarpur) Peak Hours : 4 min. Lean Hours : Upto 12 min. (Inderlok - Mundka) Frequency : Upto 10 min.

Last trains from Inder lok Towards Dilshad Garden 23:18 hrs Towards Rithala 23:27 hrs Towards Jahangirpuri 23:33 hrs **Towards Huda City Centre** 22:51 hrs Towards Anand Vihar 22:27 hrs Towards Vishwavidyalaya 23:33 hrs Towards Noida City Centre 22:39 hrs Towards Mundka 23:30 hrs Towards Dwarka 22:51 hrs Towards Sarita Vihar 22:51 hrs

Delhi metro timetable: example

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- However, spreadsheet:
 - insufficient for complex allocations/constraint validation
 - has limited ability as a 'solver'
 - no 'while' loop, no jumping from one solution to another
 - not OK for automation, nor for large data
 - semi-automatic, at best

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In my understanding

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Machine Learning/Artificial-intelligence

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 - optimize crew allotment
 - choose start-timings to get 'better and better' grouping at congested section

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Recent railway projects completed/ongoing

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- Use of section simulator for ALD study: for RITES Ltd: 2018
- Use of section simulator for Niti-Aayog: short study: 2017
- Ahmedabad junction: simulation based congestion study: 2019
- WR/CR: crew allotment and harbour line timetable: 2019 (but)
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- PhD-students portal: database/scheduling: semi-automation (in EE, IITB)

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- Commercial software: dependence: cannot install on many computers

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- Gurobi as the solver: Gurobi's limited version is free
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- Commercial software: dependence: cannot install on many computers
- Software made in FOSS can be modified to yield valuable statistics for analysis

Ready-made/proprietary vs 'home-grown' software

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- Problem specifications: from our own 'shop-floor': operations personnel
- Railways/Academia/Software-agency: combination inevitable

Zero-base TT efforts elsewhere: example

The New Dutch Timetable: The OR Revolution, 2009 paper in *Interfaces* Paper abstract

In December 2006, Netherlands Railways introduced a completely new timetable. Its objective was to facilitate the growth of passenger and freight transport on a highly utilized railway network, and improve the robustness of the timetable resulting in less train delays in the operation. Further adjusting the existing timetable constructed in 1970 was not option anymore, because further growth would then require significant investments in the rail infrastructure.

Constructing a railway timetable from scratch for about 5,500 daily trains was a complex problem. To support this process, we generated several timetables using sophisticated operations research techniques, and finally selected and implemented one of these timetables. Furthermore, because rolling-stock and crew costs are principal components of the cost of a passenger railway operator, we used innovative operations research tools to devise efficient schedules for these two resources.

The new resource schedules and the increased number of passengers resulted

Madhu Belur (EE, IITB)

The New Dutch Timetable: The OR Revolution, 2009 paper in *Interfaces* Authors: Kroon, Huisman, Abbink, Fioole, Ybema, Maroti, Schrijver, Steenbeek, Fischetti

Affiliations: railway personnel/software-firm/academia

- Department of Logistics, Netherlands Railways: Kroon, Huisman, Abbink, Fioole, Ybema
- Rotterdam School of Management, Erasmus University: Kroon, Maroti
- Econometric Institute, Erasmus University Rotterdam: Huisman
- CWI and University of Amsterdam: Schrijver
- Safiro Software Solutions: Steenbeek
- University of Padova Italy: Fischetti

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Similar efforts in Germany: Narayan Rangaraj's collaborators

- Need to shift to modern tools for
 - self-growth (and ourselves remaining relevant over next few decades)
 - system productivity/efficiency
- Tools that are 'home-grown' and in FOSS allow complete flexibility/independence and customization
- Describing/formulating the specs of the tool:
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 - from anybody interested in the area

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- This talk will soon be available at www.ee.iitb.ac.in/%7Ebelur/talks www.ee.iitb.ac.in/%7Ebelur/railways/workshop www.ee.iitb.ac.in/%7Ebelur/railways
- Narayan Rangaraj

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Automation in operations planning

Mar'23