

Outline

- Crew Scheduling
- Work Rules and Policies
- Manual Scheduling Process
- Model Formulation
- Automated Scheduling

Input

- A set of vehicle blocks each starting with a pull-out and ending with a pull-in at the depot
- Crew work rule constraints and pay provisions

Objective

- Define crew duties (i.e. runs, days, or shifts) covering all vehicle block time so as to minimize crew costs

Constraints

- Work rules (hard constraints)
- Policies (preferences or soft constraints)
- Crews available
 - in the short run, the number of crews available are known

Variations

- different crew types
 - full-time
 - part-time
- mix restrictions
 - constraints on maximum number of part-timers

Three-stage sequential approach

1. Cutting long vehicle blocks into pieces of work
2. Combining pieces to form alternative runs
 - with meal breaks, etc.
3. Selection of minimum cost set of runs

- manual process includes only steps 1 and 2
 - also accomplished with automatic heuristics
- optimization process also involves step 3

MIT Typical Crew Scheduling Approach

1 - Cutting Blocks

- each block consists of a sequence of vehicle revenue trips and non-revenue activities
- blocks can be cut only at relief points where one crew can replace another.
- relief points are typically at terminals which are accessible
 - but en-route timing points are an option
- avoid cuts within peak period
 - don't complicate an already stressed part of operations
- resulting pieces
 - have minimum and maximum lengths
 - meal break
 - maximum spread
 - should be combinable to form legal runs

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MIT Vehicle Block Partitions

Definition a *partition* of a block is the selection of a set of cuts each representing a relief.

Key problems

- very hard to evaluate a partition before forming runs
- many partitions are possible for any vehicle block

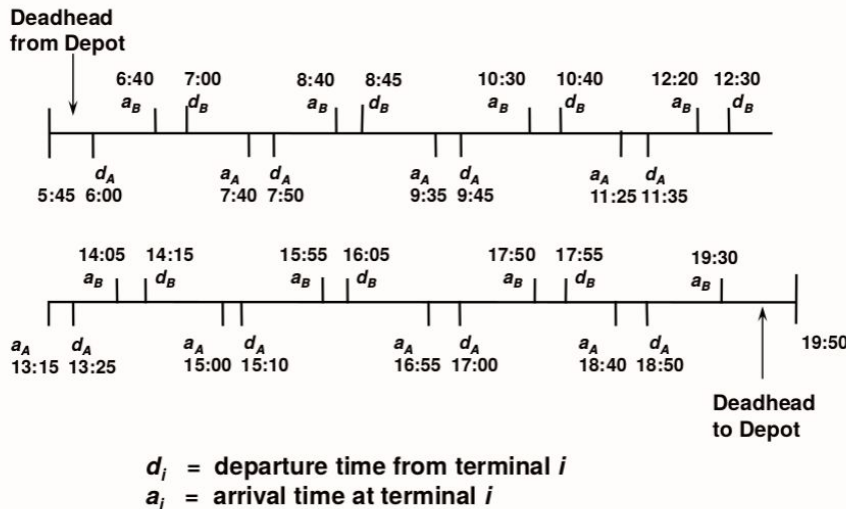
Possible Approaches

- generate only one partition for each vehicle block
- generate multiple partitions for each vehicle block
- generate all possible partitions for each vehicle block

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MIT A Vehicle Block on Route AB



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MIT Combining Pieces of Work to Form Runs

2 - Combining Pieces

- Large number of feasible runs by combining pieces of work
- Work rules are complex and constraining
 - maximum work hours: e.g. 8 hrs 15 min
 - minimum paid hours (guarantee time): e.g. 8 hrs
 - overtime constraints and pay premiums: e.g. 50% pay premium
- Spread constraints and pay premiums: time between first report and last release for duty, e.g.,



has a spread of 12 hours.

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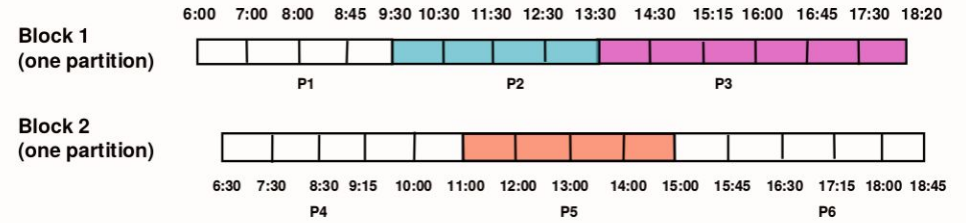
Combining Pieces of Work to Form Runs

- Swing pay premiums associated with runs with pieces which start and end at different locations, e.g.,

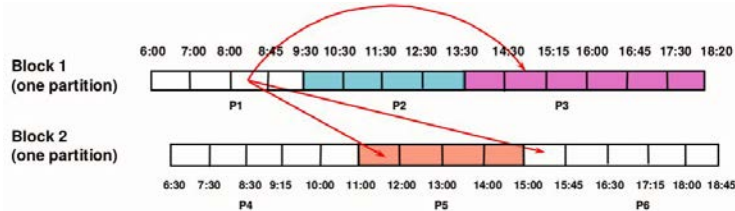


- Different types of duties
 - straight: a continuous run
 - split: a two-piece run
 - trippers: a short run, usually worked on overtime
- Approach: generate and cost out each feasible run
 - infeasible runs are not generated

Combining Pieces of Work to Form Runs



Combining Pieces of Work to Form Runs



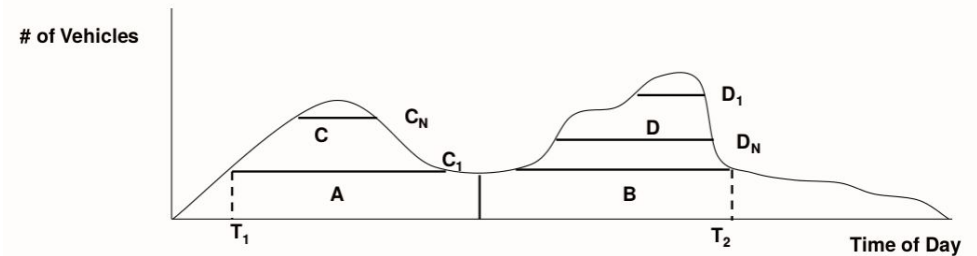
Run #	1st piece	2nd piece	Spread Time	Work Time	Cost
1	P1	P2	07:30	07:30	C1
2	P1	P3	12:20	08:20	C2
3	P1	P5	09:00	07:30	C3
4	P1	P6	12:45	07:15	C4
5	P2	P3	08:50	08:50	C5
6	P2	P6	09:15	07:45	C6
7	P4	P3	11:50	09:20	C7
8	P4	P5	08:30	08:30	C8
9	P4	P6	12:15	08:15	C9
10	P5	P6	07:45	07:45	C10

Possible Runs from pieces P1-P6

Illegal run: Max work time violation

Crew Scheduling: Manual Techniques

3- Minimizing Cost



- T_1 is earliest AM pullout which can still serve PM peak
- T_2 is latest PM pullback which can still serve AM peak
- A are AM straights (or short split runs)
- B are PM straights (or short split runs)
- C and D are long split or part time runs

MIT Typical Sequence

1. Based on total vehicle hours, estimate total operators required
2. Determine # operators required in AM and PM peaks
3. Determine B based on: # of pull-ins after time T_2 .
4. Determine # split runs: (# of PM Peak Vehicles - B)
5. Determine A based on: # of AM Peak Vehicles - split runs
6. Combine earliest pullouts in C with earliest pull-ins in D to produce minimum spread split runs C_1D_1 . Iterate until all split runs are matched C_ND_N .

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MIT Examples (1/2)

Time Period	# Vehicles	Period Length	# Vehicle Hours
AM Peak	8	3	24
Base	4	6	24
PM Peak	8	3	24
Evening	4	6	24
			96 12 FTOs

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MIT Examples (1/2)

Time Period	# Vehicles	Period Length	# Vehicle Hours
AM Peak	8	3	24
Base	4	6	24
PM Peak	8	3	24
			PM + Evening runs = 4 (Straight)
Evening	4	6	24
			PM + Evening runs = 4 (Straight)
			96 12 FTOs

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			PM + Evening runs = 4 (Straight) AM + PM runs = 4 (Split)
Evening	4	6	24
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			96 12 FTOs

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MIT Examples (2/2)

Time Period	# Vehicles	Period Length	# Vehicle Hours
AM Peak	8	3	24
Base	6	6	36
PM Peak	8	3	24
Evening	3	6	18
			102 13 FTOs

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MIT Examples (2/2)

Time Period	# Vehicles	Period Length	# Vehicle Hours
AM Peak	8	3	24
Base	6	6	36
PM Peak	8	3	24
			PM + Evening runs = 3 (Straight)
Evening	3	6	18
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			102 13 FTOs

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MIT Examples (2/2)

Time Period	# Vehicles	Period Length	# Vehicle Hours
AM Peak	8	3	24
			AM + PM runs = 5 (Split)
Base	6	6	36
PM Peak	8	3	24
			PM + Evening runs = 3 (Straight) AM + PM runs = 5 (Split)
Evening	3	6	18
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MIT Examples (2/2)

Time Period	# Vehicles	Period Length	# Vehicle Hours
AM Peak	8	3	24
			AM + PM runs = 5 (Split) AM + Base runs = 3 (Straight)
Base	6	6	36
			AM + Base runs = 3 (Straight) Base only runs = 3 (Part-time)
PM Peak	8	3	24
			PM + Evening runs = 3 (Straight) AM + PM runs = 5 (Split)
Evening	3	6	18
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MIT Examples (2/2)- Start from AM

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MIT Examples (2/2)- Start from AM

Time Period	# Vehicles	Period Length	# Vehicle Hours
AM Peak	8	3	24
			AM + Base runs = 6 (Straight) AM + PM runs = 2 (Split)
Base	6	6	36
			AM + Base runs = 6 (Straight)
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MIT Selection of Minimum Cost Set of Runs

- Usually built around mathematical programming formulation
- **Problem** Given a set of m trips and a set of n feasible driver runs, find a subset of the n runs which cover all trips at minimum cost

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Basic Model Set Partitioning Problem

• Notation

- P = set of trips to be covered
- R = set of feasible runs
- c_j = cost of run j
- δ_i^j = binary parameter
 - 1 - trip (or set of trips) i is included in run j
 - 0 - not included
- x_j = binary decision variable
 - 1 - run j is selected
 - 0 - not selected

$$\begin{aligned} \text{Min} \quad & \sum_{j \in R} c_j x_j \\ \text{Subject to:} \quad & \sum_{j \in R} x_j \delta_i^j = 1 \quad \forall i \in P \\ & x_j \in \{0,1\}, \quad \forall j \in R \end{aligned}$$

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• Problem size

- R decision variables (likely to be in millions)
- P constraints (likely to be in thousands)
- much more difficult than the MDVSP
 - complex work rules
 - many valid duties

• Problem size reduction strategy

- replace individual trips with compound trips consisting of a sequence of vehicle trips which will always be served by a single crew
- sometimes the first constraint is relaxed to simplify computation
 - allow more than one driver per trip
 - usually leads to few cases of overcovering, which can be eliminated afterwards with heuristics

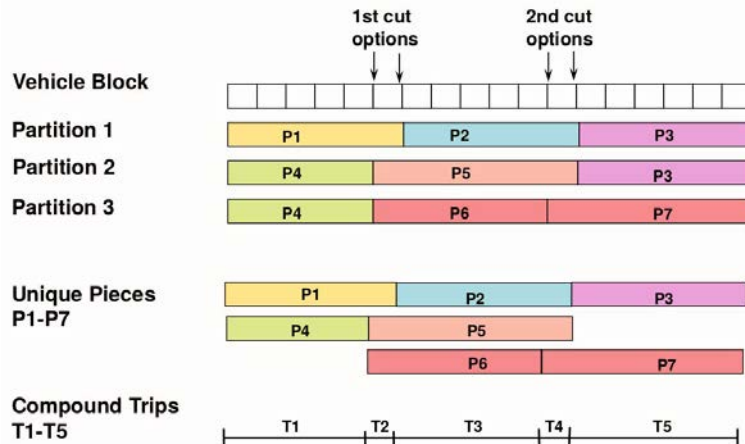
• Optimization methods

- column generation
- branch and price
- heuristics, e.g. genetic algorithms

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MIT Partitions of Vehicle Block, Pieces of Work, and Compound Trip



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MIT Variations of Set Partitioning Problem

1. Set R consists of all feasible runs given all feasible partitions for all vehicle blocks
 - size of model explodes with problem size
 - only possible for small problems
2. Set R consists of a subset of all feasible runs
 - not guaranteed to find an optimal solution
 - effectiveness will depend on quantity and quality of runs included
3. Column generation based on starting with a subset of runs and generating additional runs which will improve the solution as part of the model solution process.

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MIT Model with Side Constraints

Often the number (or mix) of crew types is constrained in various ways which can be formulated as side constraints

Example: Suppose total tripper hours are constrained to be less than 25% of timetable time.

Let WT = total timetable time

Let WT = total timetable time

R^T = set of tripper runs

t_j = work time for tripper run j

Then the additional constraint is
$$\sum_{j \in R^T} t_j x_j \leq 0.25 WT$$

MIT Automated Crew Scheduling Systems

- Virtually universally used in medium and large operators worldwide
- Two most widely used commercial packages are HASTUS (by GIRO Inc in Montreal) and Trapeze (by Trapeze Software Inc in Toronto/Phoenix), each with over 200 customers worldwide
- Typical cost ranges from \$100K to \$2 M for the software
- Key benefits of automated scheduling are:
 - scheduling process time reductions
 - improved accuracy
 - modest improvements in efficiency (typically 0-3%)
 - provides a key database for many other applications

MIT Automated Crew Scheduling Systems

- Evolution of software has been from “black box” optimization/heuristics to highly interactive and graphical tools
- Current systems allow much greater ability to “shape” the solution to the needs of specific agencies
- One implication however is a profusion of these “soft” parameters which means greater complexity and it is very hard to get full value out of systems.

MIT Rostering

Problem Given duties for a week or a month, combine duties to form rosters

Approaches

- Optimization
 - Minimize required drivers
 - Distribute work evenly - most common in Europe
 - Improve likeability
 - consecutive days off
 - Monday-Friday with weekend off
 - Very large problem: exact OR formulations are seldom used
- Cafeteria pick
 - Most common in North America
 - Pick in order of seniority

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