

DMP204 SCHEDULING, TIMETABLING AND ROUTING

Lecture 23 **Workforce Scheduling**

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1. Transportation Timetabling

2. Workforce Scheduling

Crew Scheduling and Rostering

Employee Timetabling

Shift Scheduling

Nurse Scheduling

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Outline

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1. Transportation Timetabling

2. Workforce Scheduling

Crew Scheduling and Rostering

Employee Timetabling

Shift Scheduling

Nurse Scheduling

A note on terminology

Shift: consecutive working hours

Roster: shift and rest day patterns over a fixed period of time (a week or a month)

Two main approaches:

- coordinate the design of the rosters and the assignment of the shifts to the employees, and solve it as a single problem.
- consider the scheduling of the actual employees only after the rosters are designed, solve two problems in series.

Features to consider: rest periods, days off, preferences, availabilities, skills.

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Workforce Scheduling Overview

Workforce Scheduling Overview

Workforce Scheduling:

1. Crew Scheduling and Rostering
2. Employee Timetabling

1. **Crew Scheduling and Rostering** is workforce scheduling applied in the **transportation and logistics sector** for enterprises such as airlines, railways, mass transit companies and bus companies (pilots, attendants, ground staff, guards, drivers, etc.)

The peculiarity is finding logistically feasible assignments.

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2. **Employee timetabling** (aka labor scheduling) is the operation of assigning **employees** to **tasks** in a **set of shifts** during a **fixed period of time**, typically a week.

Examples of employee timetabling problems include:

- assignment of nurses to shifts in a hospital,
- assignment of workers to cash registers in a large store
- assignment of phone operators to shifts and stations in a service-oriented call-center

Differences with Crew scheduling:

- no need to travel to perform tasks in locations
- start and finish time not predetermined

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Input:

- A set of flight legs (departure, arrival, duration)
- A set of crews

Output: A subset of flights feasible for each crew

How do we solve it?

Set partitioning or set covering??

Often treated as set covering because:

- its linear programming relaxation is numerically more stable and thus easier to solve
- it is trivial to construct a feasible integer solution from a solution to the linear programming relaxation
- it makes possible to restrict to only rosters of maximal length

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(k, m) -cyclic Staffing Problem

Assign persons to an m -period cyclic schedule so that:

- requirements b_i are met
- each person works a shift of k consecutive periods and is free for the other $m - k$ periods. (periods 1 and m are consecutive)

and the cost of the assignment is minimized.

min cx

$$st \quad \begin{bmatrix} 1 & 0 & 0 & 1 & 1 & 1 & 1 \\ 1 & 1 & 0 & 0 & 1 & 1 & 1 \\ 1 & 1 & 1 & 0 & 0 & 1 & 1 \\ 1 & 1 & 1 & 1 & 0 & 0 & 1 \\ 1 & 1 & 1 & 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 1 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 & 1 & 1 & 1 \end{bmatrix} x \geq b \quad (P)$$

$x \geq 0$ and integer

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Creating daily shifts:

- roster made of m time intervals not necessarily identical
- during each period, b_i personnel is required
- n different shift patterns (columns of matrix A)

min $c^T x$

st $Ax \geq b$

$x \geq 0$ and integer

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Recall: **Totally Unimodular Matrices**

Definition: A matrix A is **totally unimodular** (TU) if every square submatrix of A has determinant $+1$, -1 or 0 .

Proposition 1: The linear program $\max\{cx : Ax \leq b, x \in \mathbf{R}_+^m\}$ has an integral optimal solution for all integer vectors b for which it has a finite optimal value if and only if A is **totally unimodular**

Recognizing total unimodularity can be done in polynomial time (see [Schrijver, 1986])

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Basic examples:

Theorem

The $V \times E$ -incidence matrix of a graph $G = (V, E)$ is totally unimodular if and only if G is bipartite

Theorem

The $V \times A$ -incidence matrix of a directed graph $D = (V, A)$ is totally unimodular

Theorem

Let $D = (V, A)$ be a directed graph and let $T = (V, A_0)$ be a directed tree on V . Let M be the $A_0 \times A$ matrix defined by, for $a = (v, w) \in A$ and $a' \in A_0$

$$M_{a',a} := \begin{cases} +1 & \text{if the unique } v-w\text{-path in } T \text{ passes through } a' \text{ forwardly;} \\ -1 & \text{if the unique } v-w\text{-path in } T \text{ passes through } a' \text{ backwardly;} \\ 0 & \text{if the unique } v-w\text{-path in } T \text{ does not pass through } a' \end{cases}$$

M is called network matrix and is totally unimodular.

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What about this matrix?

$$\begin{bmatrix} 1 & 0 & 0 & 1 & 1 & 1 & 1 \\ 1 & 1 & 0 & 0 & 1 & 1 & 1 \\ 1 & 1 & 1 & 0 & 0 & 1 & 1 \\ 1 & 1 & 1 & 1 & 0 & 0 & 1 \\ 1 & 1 & 1 & 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 1 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$

Definition A $(0, 1)$ -matrix B has the **circular 1's property for rows (resp. for columns)** if the columns of B can be permuted so that the 1's in each row are circular, that is, appear in a circularly consecutive fashion

The circular 1's property for **columns** does not imply circular 1's property for **rows**.

Whether a matrix has the **circular 1's property for rows (resp. columns)** can be determined in $O(m^2n)$ time [A. Tucker, Matrix characterizations of circular-arc graphs. (1971) Pacific J. Math. 39(2) 535-545]

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All totally unimodular matrices arise by certain compositions from network matrices and from certain 5×5 matrices [Seymour, 1980]. This decomposition can be tested in polynomial time.

Definition

A $(0, 1)$ -matrix B has the **consecutive 1's property** if for any column j , $b_{ij} = b_{i'j} = 1$ with $i < i'$ implies $b_{lj} = 1$ for $i < l < i'$. That is, if there is a permutation of the rows such that the 1's in each column appear consecutively.

Whether a matrix has the **consecutive 1's property** can be determined in polynomial time [D. R. Fulkerson and O. A. Gross; Incidence matrices and interval graphs. 1965 Pacific J. Math. 15(3) 835-855.]

A matrix with **consecutive 1's property** is called an interval matrix and they can be shown to be network matrices by taking a directed path for the directed tree T

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Integer programs where the constraint matrix A have the **circular 1's property for rows** can be solved efficiently as follows:

Step 1 Solve the linear relaxation of (P) to obtain x'_1, \dots, x'_n . If x'_1, \dots, x'_n are integer, then it is optimal for (P) and STOP. Otherwise go to Step 2.

Step 2 Form two linear programs LP1 and LP2 from the relaxation of the original problem by adding respectively the constraints

$$x_1 + \dots + x_n = \lfloor x'_1 + \dots + x'_n \rfloor \tag{LP1}$$

and

$$x_1 + \dots + x_n = \lceil x'_1 + \dots + x'_n \rceil \tag{LP2}$$

From LP1 and LP2 an integral solution certainly arises (P)

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Cyclic Staffing with Overtime

- Hourly requirements b_i
- Basic work shift 8 hours
- Overtime of up to additional 8 hours possible

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minimize    cx
subject to
07 [ 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1
08 [ 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1
09 [ 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1
10 [ 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1
11 [ 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1
12 [ 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1
13 [ 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1
14 [ 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1
15 [ 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
16 [ 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
17 [ 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
18 [ 0 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
19 [ 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
20 [ 0 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
21 [ 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
22 [ 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
23 [ 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
24 [ 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
01 [ 0 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
02 [ 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
03 [ 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
04 [ 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
05 [ 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
06 [ 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1

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$x \geq b$

$x \geq 0$ and integer.

Days-Off Scheduling

- Guarantee two days-off each week, including every other weekend.

IP with matrix A :

first week	1 1 1 1 1 1 1	1 1 1 1 1 1 0
	1 1 1 1 1 1 1	1 1 1 1 0 0 0
	1 1 1 1 1 1 1	1 1 1 0 0 1 1
	1 1 1 1 1 1 1	1 0 0 1 1 1 1
	0 0 0 0 0 0 0	0 0 1 1 1 1 1
	0 0 0 0 0 0 0	0 1 1 1 1 1 1
second week	0 1 1 1 1 1 1	1 1 1 1 1 1 1
	0 0 1 1 1 1 1	1 1 1 1 1 1 1
	1 0 0 1 1 1 1	1 1 1 1 1 1 1
	1 1 0 0 1 1 1	1 1 1 1 1 1 1
	1 1 1 0 0 1 1	1 1 1 1 1 1 1
	1 1 1 1 0 0 0	0 0 0 0 0 0 0
	1 1 1 1 1 1 0	0 0 0 0 0 0 0

Cyclic Staffing with Part-Time Workers

- Columns of A describe the work-shifts
- Part-time employees can be hired for each time period i at cost c'_i per worker

$$\begin{aligned} \min \quad & cx + c'x' \\ \text{st} \quad & Ax + Ix' \geq b \\ & x, x' \geq 0 \text{ and integer} \end{aligned}$$

Cyclic Staffing with Linear Penalties for Understaffing and Overstaffing

- demands are not rigid
- a cost c'_i for understaffing and a cost c''_i for overstaffing

$$\begin{aligned} \min \quad & cx + c'x' + c''(b - Ax - x') \\ \text{st} \quad & Ax + Ix' \geq b \\ & x, x' \geq 0 \text{ and integer} \end{aligned}$$

- Hospital: head nurses on duty seven days a week 24 hours a day
- Three 8 hours shifts per day (1: daytime, 2: evening, 3: night)
- In a day each shift must be staffed by a different nurse
- The schedule must be the same every week
- Four nurses are available (A,B,C,D) and must work at least 5 days a week.
- No shift should be staffed by more than two different nurses during the week
- No employee is asked to work different shifts on two consecutive days
- An employee that works shifts 2 and 3 must do so at least two days in a row.

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Variables w_{sd} nurse assigned to shift s on day d and y_{id} the shift assigned for each day

$$w_{sd} \in \{A, B, C, D\} \quad y_{id} \in \{0, 1, 2, 3\}$$

Three different nurses are scheduled each day

$$\text{alldiff}(w_{.d}) \quad \forall d$$

Every nurse is assigned to at least 5 days of work

$$\text{cardinality}(w_{..} \mid (A, B, C, D), (5, 5, 5, 5), (6, 6, 6, 6))$$

At most two nurses work any given shift

$$\text{nvalues}(w_{s.} \mid 1, 2) \quad \forall s$$

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Mainly a feasibility problem

A CP approach

Two solution representations

	Sun	Mon	Tue	Wed	Thu	Fri	Sat
Shift 1	A	B	A	A	A	A	A
Shift 2	C	C	C	B	B	B	B
Shift 3	D	D	D	D	C	C	D

	Sun	Mon	Tue	Wed	Thu	Fri	Sat
Worker A	1	0	1	1	1	1	1
Worker B	0	1	0	2	2	2	2
Worker C	2	2	2	0	3	3	0
Worker D	3	3	3	3	0	0	3

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All shifts assigned for each day

$$\text{alldiff}(y_{.d}) \quad \forall d$$

Maximal sequence of consecutive variables that take the same values

$$\text{stretch-cycle}(y_{i.} \mid (2, 3), (2, 2), (6, 6), P) \\ \forall i, P = \{(s, 0), (0, s) \mid s = 1, 2, 3\}$$

Channeling constraints between the two representations:

on any day, the nurse assigned to the shift to which nurse i is assigned must be nurse i

$$w_{y_{id},d} = i \quad \forall i, d$$

$$y_{w_{sd},d} = s \quad \forall s, d$$

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The complete CP model

Alldiff: $\left\{ \begin{array}{l} (w..d) \\ (y..d) \end{array} \right\}$, all d

Cardinality: $(w.. | (A, B, C, D), (5, 5, 5, 5), (6, 6, 6, 6))$

Nvalues: $(w_s. | 1, 2)$, all s

Stretch-cycle: $(y_i. | (2, 3), (2, 2), (6, 6), P)$, all i

Linear: $\left\{ \begin{array}{l} w_{y_id} = i, \text{ all } i \\ y_{w_sd} = s, \text{ all } s \end{array} \right\}$, all d

Domains: $\left\{ \begin{array}{l} w_{sd} \in \{A, B, C, D\}, s = 1, 2, 3 \\ y_{id} \in \{0, 1, 2, 3\}, i = A, B, C, D \end{array} \right\}$, all d

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Constraint Propagation:

- alldiff: matching
- nvalues: max flow
- stretch: poly-time dynamic programming
- index expressions w_{y_id} replaced by z and constraint:
element(y, x, z): z be equal to y -th variable in list x_1, \dots, x_m

Search:

- branching by splitting domains with more than one element
- first fail branching
- symmetry breaking:
 - employees are indistinguishable
 - shifts 2 and 3 are indistinguishable
 - days can be rotated

Eg: fix A, B, C to work 1, 2, 3 resp. on Sunday

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Local search methods and metaheuristics are used if the problem has large scale. Procedures very similar to what we saw for employee timetabling.