

ISE321 Notes on a Rotation Model

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1 Basic Mathematical Notation

We define some mathematical notation that are commonly used in a mathematical optimization model.

- Set: a collection of elements – e.g., $I = \{1, 3, 4, 6, 7\}$
- “ \in ” means “in” – e.g., if $I = \{1, 3, 4, 6, 7\}$, we have $3 \in I$ but $2 \notin I$
- “ \forall ” means “for all” – e.g., “ $\forall i \in I$ ” means for all elements i in the set I
- “ \sum_i ” means “sum over i ” – e.g., $\sum_{i=1}^5 a_i = a_1 + a_2 + \dots + a_5$
- “ $I \times J$ ” for sets I and J means “cross over” – e.g., if $I = \{1, 4\}$ and $J = \{2, 6\}$, then $I \times J = \{(1, 2), (1, 6), (4, 2), (4, 6)\}$

2 Rotation Model

A full set of notation we used is given in Table 1. In the rotation problem, we consider the assignment of a set of residents to different rotations and blocks. We use P to denote the set of residents, R to denote the set of rotations, and B to denote the set of blocks. In the rotation model, we decide the assignment of the residents to rotations and blocks. Thus, we define the *decision variables* of the model as

$$x_{p,r,b} = \begin{cases} 1 & \text{if resident } p \text{ is assigned to rotation } r \text{ and block } b, \\ 0 & \text{otherwise,} \end{cases}$$

for all $p \in P$, $r \in R$, and $b \in B$. Note that the (mathematical) decision variables $x_{p,r,b}$ for all combinations $(p, r, b) \in P \times R \times B$ correspond to a schedule in the real life.

To ensure that a schedule is *implementable*, we need to impose some *constraints*. We can derive mathematical constraints that correspond to the physical constraints in the real world.

Example: Constraints Set I

In practice, we may require that for **each** resident $p \in P$ and **each** block $b \in B$, the **number of rotations that one can perform at a time is exact one**. We can formulate the

following constraint:

$$\sum_{r \in R} x_{p,r,b} = 1, \quad \forall p \in P, b \in B. \quad (2.1)$$

First, the “**each**” components are represented using “ $\forall p \in P, b \in B$ ”, or equivalently, “ $\forall (p, b) \in P \times B$ ”. Now, **for a fixed $p \in P$ and $b \in B$** , the constraint reads $\sum_{r \in R} x_{p,r,b} = 1$. Recall that $x_{p,r,b} \in \{0, 1\}$ (i.e., taking value 0 or 1 only), and it takes value 1 only when resident p is assigned to rotation r and block b . Therefore, we can interpret the summation $\sum_{r \in R} x_{p,r,b}$ as the **number of rotations** that resident p at block b performs. Therefore, the mathematical constraints (2.1) correspond to **our requirements in the physical world**.

Example: Constraints Set II

Next, suppose that **for each** resident $p \in P$ and **each** rotation $r \in R$, the number of blocks assigned to him/her should be at least $c_{p,r}^{\min}$ but at most $c_{p,r}^{\max}$. We can formulate the following constraint:

$$c_{p,r}^{\min} \leq \sum_{b \in B} x_{p,r,b} \leq c_{p,r}^{\max}, \quad \forall p \in P, r \in R. \quad (2.2)$$

Again, the “**each**” components are represented using “ $\forall p \in P, r \in R$ ”, or equivalently, “ $\forall (p, r) \in P \times R$ ”. Now, **for a fixed $p \in P$ and $r \in R$** , the constraint reads $c_{p,r}^{\min} \leq \sum_{b \in B} x_{p,r,b} \leq c_{p,r}^{\max}$. Using a similar logic in the previous example, we can interpret the summation $\sum_{b \in B} x_{p,r,b}$ as the **number of blocks** that resident p performs rotation r . Therefore, the mathematical constraints (2.2) correspond to **our requirements in the physical world**.

Exercise I

Consider the following physical constraint. **For each** rotation $r \in R$ and **each** block $b \in B$, the number of residents assigned should be at least p_r^{\min} but at most p_r^{\max} . (Note that, in our case, the minimum and maximum number of residents depend on the rotation only). **Formulate the corresponding mathematical constraints; see model (2.4) for the solution.**

Exercise II

Consider the following physical constraint (e.g., due to some educational requirement). Let A_{pri} be the set of combinations $(p, r, b) \in P \times R \times B$ that we call **priority assignments**. These assignments **must be satisfied** in the output schedule. **Formulate the corresponding mathematical constraints.**

Despite the fact that we have formulated different constraints, we still need something to *optimize*. That is, we want to construct the *objective function* of our rotation model. In the current formulation, we can consider the following objective. Let A_{pref} be the set of combinations $(p, r, b) \in P \times R \times B$ that we call preference assignments. It is good that the output schedule can fulfill these assignments (or part of them), but they are *not necessary*. Thus, we may want to *maximize* the total number of preference assignments while keeping all other constraints. This gives rise to the following objective of the rotation model:

$$\text{maximize} \quad \sum_{(p,r,b) \in A_{\text{pref}}} x_{p,r,b}. \quad (2.3)$$

Note that the summation $\sum_{(p,r,b) \in A_{\text{pref}}} x_{p,r,b}$ counts **the number of preference assignments** in the schedule. Therefore, in the rotation model, we maximize this summation, which is exactly our goal.

Finally, we can combine our objective function (2.3) with constraints (2.1)–(2.2) and several other **additional** constraints. Define the following sets.

- A_{pri} : set of priority assignments $(p, r, b) \in P \times R \times B$ that must be satisfied
- A_{imp} : set of impossible assignments $(p, r, b) \in P \times R \times B$ that must be excluded
- B_{gone} : set of combinations $(p, b) \in B_{\text{gone}}$ that resident p has conflict with block b , i.e., no assignments are possible with such a combination of (p, b)
- C_{imp}^1 : set of combinations $(p, r) \in C_{\text{imp}}^1$ that resident p cannot perform rotation r
- C_{imp}^2 : set of combinations $(r, b) \in C_{\text{imp}}^2$ that rotation r cannot be performed in block b
- P_{all} : set of all-year residents
- R_{must} : set of must-do rotations (i.e., at least once) for all-year residents

With this notation, we can formulate our first rotation model as:

$$\text{maximize}_x \quad \sum_{(p,r,b) \in A_{\text{pref}}} x_{p,r,b} \quad (2.4a)$$

$$\text{subject to} \quad \sum_{r \in R} x_{p,r,b} = 1, \quad \forall p \in P, b \in B, \quad (2.4b)$$

$$c_{p,r}^{\min} \leq \sum_{b \in B} x_{p,r,b} \leq c_{p,r}^{\max}, \quad \forall p \in P, r \in R, \quad (2.4c)$$

$$p_r^{\min} \leq \sum_{p \in P} x_{p,r,b} \leq p_r^{\max}, \quad \forall r \in R, b \in B, \quad (\text{Exercise I}) \quad (2.4d)$$

$$x_{p,r,b} = 1, \quad \forall (p, r, b) \in A_{\text{pri}}, \quad (\text{Exercise II}) \quad (2.4e)$$

$$x_{p,r,b} = 0, \quad \forall (p, r, b) \in A_{\text{imp}}, \quad (2.4f)$$

$$x_{p,r,b} = 0, \quad \forall r \in R, (p, b) \in B_{\text{gone}}, \quad (2.4g)$$

$$x_{p,r,b} = 0, \quad \forall b \in B, (p, r) \in C_{\text{imp}}^1, \quad (2.4h)$$

$$x_{p,r,b} = 0, \quad \forall p \in P, (r, b) \in C_{\text{imp}}^2, \quad (2.4i)$$

$$\sum_{b \in B} x_{p,r,b} \geq 1, \quad \forall p \in P_{\text{all}}, r \in R_{\text{must}}, \quad (2.4j)$$

$$x_{p,r,b} \in \{0, 1\}, \quad \forall p \in P, r \in R, b \in B. \quad (2.4k)$$

Exercise III

Give the interpretations/physical meanings of constraints (2.4f)–(2.4j).

3 Food for Thought

- a. **Busy rotations.** Let R_{busy} be the set of busy rotations (e.g., heavy workloads). Consider the following physical constraints.
 - i. B_{vac} is a set of combinations $(p, b) \in B_{\text{vac}}$ that resident p is on vacations on block b . Hospital cannot assign residents to busy rotations if residents are on vacation. How can we add constraint(s) to satisfy this requirement?
 - ii. It is *not preferable* that a resident is assigned to two busy rotations in consecutive blocks. How can we modify (2.4) to include such a consideration?
- b. **Interface with solvers.** Suppose we can solve the optimization by some back-end solver. What kinds of interface we can use for effective communication between users and the optimization model (both inputs and output)?
- c. **Other practical constraints.** Can you think of any other reasonable/desirable constraints for the rotation model? If yes, how can you incorporate them into model (2.4)?

A Notation Summary

Table 1: Notation

Indices	
p	index of residents, $p \in P$
r	index of rotations, $r \in R$
b	index of blocks, $b \in B$
Parameters and sets	
P	set of residents
R	set of rotations
B	set of blocks
A_{pri}	set of priority assignments
A_{pref}	set of preference assignments
A_{imp}	set of impossible assignments
B_{gone}	set of (p, b) that a resident p is in conflict with block b
C_{imp}^1	set of (p, r) that resident p cannot perform rotation r
C_{imp}^2	set of (r, b) that rotation r cannot be performed in block b
P_{all}	set of all-year residents
R_{must}	set of rotations that all-year residents must do
p_r^{\min}	minimum number of residents in rotation r
p_r^{\max}	maximum number of residents in rotation r
$c_{p,r}^{\min}$	minimum number of blocks that resident p performs rotation r
$c_{p,r}^{\max}$	maximum number of blocks that resident p performs rotation r
Variables	
$x_{p,r,b}$	1 if resident p is assigned to rotation r and block b , and 0 otherwise