

Operations Research

Applications of Integer Programming

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Road map

- ▶ **Facility location problems.**
- ▶ Machine scheduling problems.
- ▶ Vehicle routing problems.

Facility location problems

- ▶ One typical managerial decision is “where to build my facility?”
 - ▶ Where to open convenience stores?
 - ▶ Where to build warehouses or distribution centers?
 - ▶ Where to build factories?
 - ▶ Where to build power stations, fire stations, or police stations?
- ▶ A similar question is “where to locate a scarce resource?”
 - ▶ Where to put a limited number of fire engines or ambulances?
 - ▶ Where to put a limited number of police officers?
 - ▶ Where to put a limited number of ice cream machines?
- ▶ These problems are **facility location problems**.
 - ▶ In this lecture, we focus on **discrete** facility location problems: We choose a subset of locations from a set of finite locations.

Facility location problems

- ▶ In general, there are some **demand nodes** and some **potential locations**.
 - ▶ We build facilities at locations to serve demands.
 - ▶ E.g., build distribution centers to ship to retail stores.
 - ▶ E.g., build fire stations to cover cities, towns, and villages.
- ▶ Facility location problems are typically categorized based on their objective functions.
- ▶ In this lecture, we introduce three types of facility location problems:
 - ▶ **Set covering problems**: Build a minimum number of facilities to cover all demands.
 - ▶ **Maximum covering problems**: Build a given number of facilities to cover as many demands as possible.
 - ▶ **Fixed charge location problems**: Finding a balance between benefit of covering demands and cost of building facilities.

Set covering problems

- ▶ Consider a set of demands I and a set of locations J .
- ▶ The distance (or traveling time) between demand i and location j is $d_{ij} > 0$, $i \in I$, $j \in J$.
- ▶ A service level $s > 0$ is given: Demand i is said to be “covered” by location j if $d_{ij} < s$.
- ▶ Question: How to allocate as few facilities as possible to cover all demands?

Set covering problems

- ▶ Let's define the following parameter: $a_{ij} = 1$ if $d_{ij} < s$ or 0 otherwise, $i \in I$, $j \in J$.
- ▶ Let's define the following variables: $x_j = 1$ if a facility is built at location $j \in J$ or 0 otherwise.
- ▶ The complete formulation:

$$\begin{aligned} \min \quad & \sum_{j \in J} x_j \\ \text{s.t.} \quad & \sum_{j \in J} a_{ij} x_j \geq 1 \quad \forall i \in I \\ & x_j \in \{0, 1\} \quad \forall j \in J. \end{aligned}$$

- ▶ The weighted version: $\min \sum_{j \in J} w_j x_j$.

Maximum covering problems

- ▶ Consider a set of demands I and a set of locations J .
- ▶ The distances d_{ij} , service level s , and the covering coefficient a_{ij} are also given.
- ▶ We are restricted to build at most $p \in \mathbb{N}$ facilities.
- ▶ Question: How to allocate at most p facilities to cover as many demands as possible?

Maximum covering problems

- ▶ Still let $x_j = 1$ if a facility is built at location $j \in J$ or 0 otherwise.
- ▶ Also let $y_i = 1$ if demand $i \in I$ is covered by any facility or 0 otherwise.
- ▶ The complete formulation:

$$\begin{aligned} \max \quad & \sum_{i \in I} y_i \\ \text{s.t.} \quad & \sum_{j \in J} a_{ij} x_j \geq y_i \quad \forall i \in I \\ & \sum_{j \in J} x_j \leq p \quad \forall j \in J \\ & x_j \in \{0, 1\} \quad \forall j \in J \\ & y_i \in \{0, 1\} \quad \forall i \in I. \end{aligned}$$

- ▶ The weighted version: $\max \sum_{i \in I} w_i y_i$.

Fixed charge location problems

- ▶ Consider a set of demands I and a set of locations J .
- ▶ At demand i , the demand size is $h_i > 0$.
- ▶ The unit shipping cost from location j to demand i is $d_{ij} > 0$.
- ▶ The fixed construction cost at location j is $f_j > 0$.
- ▶ Question: How to allocate some facilities to minimize the total shipping and construction costs?

Fixed charge location problems

- ▶ We still need x_j : $x_j = 1$ if a facility is built at location $j \in J$ or 0 otherwise.
- ▶ We now need y_{ij} : $y_{ij} = 1$ if demand $i \in I$ is served by facility at location $j \in J$ or 0 otherwise.
- ▶ The complete formulation:

$$\begin{aligned} \min \quad & \sum_{i \in I} \sum_{j \in J} h_i d_{ij} y_{ij} + \sum_{j \in J} f_j x_j \\ \text{s.t.} \quad & y_{ij} \leq x_j \quad \forall i \in I, j \in J \\ & \sum_{j \in J} y_{ij} = 1 \quad \forall i \in I \\ & x_j \in \{0, 1\} \quad \forall j \in J \\ & y_i \in \{0, 1\} \quad \forall i \in I. \end{aligned}$$

Fixed charge location problems

- ▶ The previous model is the **uncapacitated** version.
 - ▶ A facility can serve any amount of demand.
- ▶ If facility at location j has a limited capacity $K_j > 0$, we may add the capacity constraint

$$\sum_{i \in I} h_i y_{ij} \leq K_j \quad \forall j \in J.$$

- ▶ The **capacitated** version is usually called the capacitated facility location problem (abbreviated as CFL). The uncapacitated one is abbreviated as UFL.

Remarks

- ▶ When to use set covering?
 - ▶ When we are required to take care of (almost) everyone.
 - ▶ E.g., fire stations and police stations.
- ▶ When to use maximum covering?
 - ▶ When budgets are limited.
 - ▶ E.g., cellular data networks.
- ▶ When to use fixed charge location?
 - ▶ When service costs depends on distances.
 - ▶ E.g., distribution centers.
- ▶ All the three models are **NP-hard**.
 - ▶ For large instances, it really takes time to obtain an optimal solution.
 - ▶ Many researchers look for effective heuristics for these problems.

Road map

- ▶ Facility location problems.
- ▶ **Machine scheduling problems.**
- ▶ Vehicle routing problems.

Machine scheduling problems

- ▶ In many cases, **jobs/tasks** must be assigned to **machines**.
- ▶ As an example, consider a factory producing one product for n customers.
 - ▶ Serial production: Only one job can be processed at one time.
 - ▶ Each job has its due date.
 - ▶ How to schedule the n jobs to minimize the total number of delayed jobs?
- ▶ In this example, scheduling is nothing but **sequencing**.
 - ▶ Splitting jobs is not helpful.
 - ▶ There are $n!$ ways to sequence the n jobs.
 - ▶ Is there a polynomial-time algorithm?
- ▶ The problems of scheduling jobs to machines are **machine scheduling problems**.

Machine scheduling problems

- ▶ Machine scheduling problems can be categorized in multiple ways:
 - ▶ Production mode:
 - ▶ Single machine serial production.
 - ▶ Multiple parallel machines.
 - ▶ Flow shop problems.
 - ▶ Job shop problems.
 - ▶ Job splitting:
 - ▶ Non-preemptive problems.
 - ▶ Preemptive problems.
 - ▶ Performance measurement:
 - ▶ Makespan (the time that all jobs are completed).
 - ▶ (Weighted) total completion time.
 - ▶ (Weighted) number of delayed jobs.
 - ▶ (Weighted) total lateness.
 - ▶ (Weighted) total tardiness.
 - ▶ And more.

Minimizing total tardiness on a single machine

- ▶ Consider scheduling n jobs on a single machine.
- ▶ Job $j \in J = \{1, 2, \dots, n\}$ has **processing time** p_j and due time d_j .
- ▶ Different schedules give these jobs different **completion times**. The completion time of job j is denoted as C_j .
- ▶ For job j , its **tardiness** is¹

$$T_j = \max\{C_j - d_j, 0\}.$$

- ▶ There is only one machine, which can process only one job at a time.
- ▶ How to schedule all the jobs to minimize the total tardiness $\sum_{j \in J} T_j$?
- ▶ While many researchers study specific properties and algorithms for specific problems, we will only try to formulate the problem as an integer program.

¹Its **lateness** is $L_j = C_j - d_j$, which may be negative.

Minimizing total tardiness on a single machine

- ▶ Let's use C_j to be our decision variables.
- ▶ Suppose we schedule jobs 1, 2, ..., and n in this order, we will have $C_1 = p_1$, $C_2 = p_1 + p_2$, ..., and $C_n = \sum_{i=1}^n p_i$.
- ▶ A **Gantt chart** is helpful to illustrate a schedule.

- ▶ Obviously, splitting jobs does not help for this problem. (Why?)
- ▶ Because the machine can start job 2 only after job 1 is completed, we have $C_2 \geq C_1 + p_2$ as a constraint. But what if job 2 should be scheduled before job 1?

Minimizing total tardiness on a single machine

- ▶ In a feasible schedule, job i is either before or after job j , for all $j \neq i$.
- ▶ Therefore, we need to satisfy at least one of the following two constraints:

$$C_j \geq C_i + p_j \quad \text{and} \quad C_i \geq C_j + p_i.$$

- ▶ Let $z_{ij} = 1$ if job j is before job i or 0 otherwise, $i \in J, j \in J, i < j$.
- ▶ The constraints we need:

$$\begin{aligned} C_i + p_j - C_j &\leq Mz_{ij} \\ C_j + p_i - C_i &\leq M(1 - z_{ij}) \end{aligned}$$

- ▶ What value of M works?
 - ▶ How about $M = \sum_{j \in J} p_j$?

Minimizing total tardiness on a single machine

- ▶ It remains to linearize the objective function

$$\min \sum_{j \in J} \max\{C_j - d_j, 0\}.$$

- ▶ The complete formulation:

$$\begin{aligned} \min \quad & \sum_{j \in J} T_j \\ \text{s.t.} \quad & T_j \geq C_j - d_j && \forall j \in J \\ & C_i + p_j - C_j \leq Mz_{ij} && \forall i \in J, j \in J, i < j \\ & C_j + p_i - C_i \leq M(1 - z_{ij}) && \forall i \in J, j \in J, i < j \\ & T_j \geq 0, C_j \geq 0 && \forall j \in J \\ & z_{ij} \in \{0, 1\} && \forall i \in J, j \in J, i < j. \end{aligned}$$

Minimizing makespan on parallel machines

- ▶ Consider scheduling n jobs on m **parallel** machines.
 - ▶ Job $j \in J = \{1, 2, \dots, n\}$ has processing time p_j .
 - ▶ The capacity of machine $i \in I = \{1, 2, \dots, m\}$ is unlimited.
 - ▶ A job can be processed at any machine. However, it can be processed only on one machine.
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- ▶ Different schedules give these jobs different completion times C_j s.
 - ▶ The **makespan** of a schedule is $\max_{j \in J} C_j$.
 - ▶ How may we minimize the makespan?

Minimizing makespan on parallel machines

- ▶ As long as some jobs are assigned to a machine, the sequence on that machine does not matter.
- ▶ The problem of minimizing makespan is just to **assign** jobs to machines.
- ▶ Let $x_{ij} = 1$ if job $j \in J$ is assigned to machine $i \in I$ or 0 otherwise.
- ▶ On machine $i \in I$, the last job is completed at

$$\sum_{j \in J} p_j x_{ij}.$$

- ▶ The objective is to

$$\min \max_{i \in I} \left\{ \sum_{j \in J} p_j x_{ij} \right\}.$$

How to linearize it?

Minimizing makespan on parallel machines

- ▶ The complete formulation is

$$\begin{aligned} \min \quad & M \\ \text{s.t.} \quad & M \geq \sum_{j \in J} p_j x_{ij} \quad \forall i \in I \\ & \sum_{i \in I} x_{ij} = 1 \quad \forall j \in J \\ & x_{ij} \in \{0, 1\} \quad \forall i \in I, j \in J. \end{aligned}$$

- ▶ Sometimes people want to maximize the completion time of the least-loaded machine (for, e.g., fairness):

$$\begin{aligned} \max \quad & M \\ \text{s.t.} \quad & M \leq \sum_{j \in J} p_j x_{ij} \quad \forall i \in I \\ & x_{ij} \in \{0, 1\} \quad \forall i \in I, j \in J. \end{aligned}$$

Road map

- ▶ Facility location problems.
- ▶ Machine scheduling problems.
- ▶ **Vehicle routing problems.**

Vehicle routing problems

- ▶ In many cases, we need to deliver/collect items to/from customers in the most efficient way.
- ▶ E.g., consider a post officer who needs to deliver to four addresses.
- ▶ The shortest path between any pair of two addresses can be obtained.
- ▶ This is a **routing** problem: To choose a route starting from the office, passing each address exactly once, and then returning to the office.
- ▶ This is a sequencing problem; in total there are $4! = 24$ feasible routes.
- ▶ Which route minimizes the total distance (or travel time)?

Vehicle routing problems

- ▶ The problem described above is the famous **traveling salesperson problem**.
 - ▶ It assumes that the truck has ample capacity.
- ▶ Consider the truck towing bicycles in NTU. It must start at the car pound, pass several locations in NTU, and then return to the origin.
 - ▶ However, the truck capacity is quite limited (because too many people violate the parking regulation).
 - ▶ The driver needs to find **multiple** routes to cover all the locations.
- ▶ The traveling salesperson problem (TSP) is a special case of **vehicle routing problems**.

Traveling salesperson problem

- ▶ How to formulate the TSP into an integer program?
- ▶ Let's consider a directed complete network $G = (V, E)$.
 - ▶ There are n nodes and $n(n - 1)$ arcs.
 - ▶ The arc weight for arc (i, j) is $d_{ij} > 0$.
- ▶ We select a few arcs in E to form a **tour**.
 - ▶ To form a tour, we need to select n arcs.
 - ▶ These n arcs should form a cycle passing all nodes.
- ▶ Let $x_{ij} = 1$ if arc $(i, j) \in E$ is selected or 0 otherwise.
 - ▶ The objective:

$$\min \sum_{(i,j) \in E} d_{ij} x_{ij}.$$

- ▶ How to ensure the routing requirement?
- ▶ Is $\sum_{(i,j) \in E} d_{ij} x_{ij} = n$ enough?

Traveling salesperson problem

- ▶ For node $k \in V$:
 - ▶ We must select exactly one incoming arc:

$$\sum_{i \in V, i \neq k} x_{ik} = 1.$$

- ▶ We must select exactly one outgoing arc:

$$\sum_{j \in V, j \neq k} x_{kj} = 1.$$

- ▶ Now each node is on a cycle.
- ▶ However, these are not enough to prevent **subtours**.

Eliminating subtours: alternative 1

- ▶ There are at least two ways to eliminate subtours.
- ▶ For each **subset of nodes** with at least two nodes, we limit the maximum number of arcs selected:

$$\sum_{i \in S, j \in S, i \neq j} x_{ij} \leq |S| - 1 \quad \forall S \subsetneq V, |S| \geq 2.$$

- ▶ When we have n nodes, we have $2^n - n - 1$ constraints.

Eliminating subtours: alternative 2

- ▶ Let u_i s represent the order of passing nodes. More precisely, $u_i = k$ if node i is the k th node to be passed in a tour.
- ▶ We add the following constraints:

$$u_1 = 1$$

$$2 \leq u_i \leq n \quad \forall i \in V \setminus \{1\}$$

$$u_i - u_j + 1 \leq (n - 1)(1 - x_{ij}) \quad \forall (i, j) \in E, i \neq 1, j \neq 1.$$

- ▶ If $x_{ij} = 0$, there is no constraint for u_i and u_j ; otherwise, u_j must be larger than u_i by at least 1.
- ▶ If a tour does not contain node 1, the last constraint pushes those u_i s to infinity and violates constraint 2.
- ▶ Note that only node 1 is not restricted by these constraints!
- ▶ When we have n nodes, we have n additional variables and $n + (n - 1)(n - 2)$ constraints.

The complete formulation

- ▶ The complete formulation is

$$\begin{aligned}
 \min \quad & \sum_{(i,j) \in E} d_{ij} x_{ij} \\
 \text{s.t.} \quad & \sum_{i \in V, i \neq k} x_{ik} = 1 \quad \forall k \in V \\
 & \sum_{j \in V, j \neq k} x_{kj} = 1 \quad \forall k \in V \\
 & x_{ij} \in \{0, 1\} \quad \forall (i, j) \in E.
 \end{aligned}$$

with either alternative 1 or alternative 2.

- ▶ Which alternative is better?