

MS-C2105 - Introduction to Optimization

Lecture 6

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Outline of this lecture

Sensitivity analysis

- Economic interpretation

- Changes in the independent term (b)

- Changes in the objective function coefficients (c)

Integer programming problems

- The assignment problem

- The knapsack problem

- The set covering problem

- Travelling salesman problem

Reading: Taha: Chapter 4; Winston: Chapter 9

Economic interpretation

Duality can be used for obtaining practical insights. Consider the **paint production** problem (Lecture 2) and its dual:

$$\max. \quad z = 5x_1 + 4x_2$$

$$\text{s.t.: } 6x_1 + 4x_2 \leq 24$$

$$x_1 + 2x_2 \leq 6$$

$$x_2 - x_1 \leq 1$$

$$x_2 \leq 2$$

$$x_1, x_2 \geq 0$$

$$\min. \quad z = 24y_1 + 6y_2 + y_3 + 2y_4$$

$$\text{s.t.: } 6y_1 + y_2 - y_3 \geq 5$$

$$4y_1 + 2y_2 + y_3 + y_4 \geq 4$$

$$y_1, y_2, y_3, y_4 \geq 0$$

We have that $x^* = (3, 1.5)$ and $y^* = (0.75, 0.5, 0, 0)$. Notice that:

$$21 = 5x_1^* + 4x_2^* = 24y_1^* + 6y_2^* + y_3^* + 2y_4^* = 21$$

1. y can be seen as the **marginal values** each resource has for the optimal solution.
2. Only **active constraints** have marginal value (implied also by complementarity, i.e., slack $x_i \geq 0 \Rightarrow y_i = 0, i = 1, \dots, m$)

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3. Increasing (reducing) one unit of b_i will improve (worsen) the objective function value **in y_i^* units** while B remains feasible.
4. y^* can be seen as the "fair price" for the resource.

Sensitivity analysis

Notice that y^* corresponds to the **marginal values** obtained graphically (see slides 19-23 in Lecture 2).

One can use the **optimality conditions**

$r_N = (c_N^\top - c_B^\top B^{-1}N) \leq 0$ and **feasibility conditions**

$x_B = B^{-1}\bar{b} \geq 0$ to analyse the stability of solutions against:

1. Changes in availability of resources. Term b is changed by Δb . Let $\bar{b} = b + \Delta b$. and $x^* = [x_B^* \ x_N^*]$ be the optimal solution with basis B for the original LP.

- ▶ Optimality conditions $r_N = c_N - c_B^\top B^{-1}N \leq 0$ are not affected, since r_N **does not depend on b** .
- ▶ Feasibility conditions $x_B = B^{-1}\bar{b} \geq 0$ **are affected**. Changes can only be such that $B^{-1}(b + \Delta b) \geq 0$ **remain true**.
- ▶ Notice that $\bar{z} = c_B^\top B^{-1}\bar{b} = y^{*\top}(b + \Delta b) = z + y^{*\top} \Delta b$.

Sensitivity analysis

Example: Variations in b_1 . To analyse variations $b_1 + \Delta b_1$, we include an extra element to capture how the basis is altered.

	x_1	x_2	s_1	s_2	s_3	s_4	Sol.	Δb
z	-5	-4	0	0	0	0	0	0
s_1	6	4	1	0	0	0	24	1
s_2	1	2	0	1	0	0	6	0
s_3	-1	1	0	0	1	0	1	0
s_4	0	1	0	0	0	1	2	0

The optimal tableau is below. Notice how the columns s_1 and Δb remain identical, a consequence of performing only row operations.

	x_1	x_2	s_1	s_2	s_3	s_4	Sol.	Δb
z	0	0	3/4	1/2	0	0	21	3/4
x_1	1	0	1/4	-1/2	0	0	3	1/4
x_2	0	1	-1/8	3/4	0	0	3/2	-1/8
s_3	0	0	3/8	-5/4	1	0	5/2	3/8
s_4	0	0	1/8	-3/4	0	1	1/2	1/8

Sensitivity analysis

Example: Variations in b_1 . In the optimal tableau, $B^{-1}b$ is in the column 'Sol.' and thus $B^{-1}\Delta b$ is in the column ' Δb '.

	x_1	x_2	s_1	s_2	s_3	s_4	Sol.	Δb
z	0	0	3/4	1/2	0	0	21	3/4
x_1	1	0	1/4	-1/2	0	0	3	1/4
x_2	0	1	-1/8	3/4	0	0	3/2	-1/8
s_3	0	0	3/8	-5/4	1	0	5/2	3/8
s_4	0	0	1/8	-3/4	0	1	1/2	1/8

Therefore, $B^{-1}(b + \Delta b) \geq 0$ implies that

$$3 + (1/4)\Delta b_1 \geq 0 \quad \Rightarrow \Delta b_1 \geq -12$$

$$3/2 + (-1/8)\Delta b_1 \geq 0 \quad \Rightarrow \Delta b_1 \leq 12$$

$$5/2 + (3/8)\Delta b_1 \geq 0 \quad \Rightarrow \Delta b_1 \geq -6.666$$

$$1/2 + (1/8)\Delta b_1 \geq 0 \quad \Rightarrow \Delta b_1 \geq -4$$

► $-4 \leq \Delta b_1 \leq 12$

and therefore

$$b_1 \in [20, 36].$$

► $z =$

$$21 + (3/4)\Delta b_1.$$

Sensitivity analysis

2. Changes in coefficients of objective function. Term c is changed by Δc . Let $\bar{c} = c + \Delta c = \begin{bmatrix} c_B + \Delta c_B \\ c_N + \Delta c_N \end{bmatrix}$ and $x^* = [x_B^*, x_N^*]^\top$ be the optimal solution with basis B for the original LP.

- ▶ Feasibility condition $x_B = B^{-1}b \geq 0$ are not affected.
- ▶ Optimality conditions $r_N = c_N - c_B^\top B^{-1}N \leq 0$ **are affected**. Changes must be such that $\bar{c}_N - \bar{c}_B^\top B^{-1}N \leq 0$ **still holds**.
- ▶ Two cases can occur:
 1. Change in coefficients of **basic variables**:

$$\begin{aligned} c_N^\top - \bar{c}_B^\top B^{-1}N &= c_N^\top - (c_B + \Delta c_B)^\top B^{-1}N \\ &= c_N^\top - c_B^\top B^{-1}N - \Delta c_B^\top B^{-1}N \\ &= r_N - \Delta c_B^\top B^{-1}N \leq 0 \end{aligned}$$

or equivalently: $r_N \leq \Delta c_B^\top B^{-1}N$.

Sensitivity analysis

2. Changes in coefficients of objective function. Term c is changed by Δc . Let $\bar{c} = c + \Delta c = \begin{bmatrix} c_B + \Delta c_B \\ c_N + \Delta c_N \end{bmatrix}$ and $x^* = [x_B^*, x_N^*]^\top$ be the optimal solution with basis B for the original LP.

- ▶ Feasibility condition $x_B = B^{-1}b \geq 0$ are not affected.
- ▶ Optimality conditions $r_N = c_N - c_B^\top B^{-1}N \leq 0$ are affected. Changes must be such that $\bar{c}_N - \bar{c}_B^\top B^{-1}N \leq 0$ still holds.
- ▶ Two cases can occur:

2. Change in coefficients of **nonbasic variables**:

$$\begin{aligned}\bar{c}_N^\top - \bar{c}_B^\top B^{-1}N &= (c_N + \Delta c_N)^\top - c_B^\top B^{-1}N \\ &= c_N^\top - c_B^\top B^{-1}N + \Delta c_N^\top \\ &= r_N + \Delta c_N^\top \leq 0\end{aligned}$$

or equivalently: $r_N \leq -\Delta c_N^\top$.

Sensitivity analysis

Example: Variations in c_1 . The optimal tableau is perturbed by Δc_1 . Being x_1 a **basic variable**, the tableau needs to be corrected.

	x_1	x_2	s_1	s_2	s_3	s_4	Sol.
z	$-\Delta c_1$	0	$3/4$	$1/2$	0	0	21
x_1	1	0	$1/4$	$-1/2$	0	0	3
x_2	0	1	$-1/8$	$3/4$	0	0	$3/2$
s_3	0	0	$3/8$	$-5/4$	1	0	$5/2$
s_4	0	0	$1/8$	$-3/4$	0	1	$1/2$

To do so, we multiply the x_1 -row by Δc_1 and add it to the z -row.

	x_1	x_2	s_1	s_2	s_3	s_4	Sol.
z	0	0	$\frac{3}{4} + \frac{1}{4}(\Delta c_1)$	$\frac{1}{2} - \frac{1}{2}(\Delta c_1)$	0	0	$21 + 3\Delta c_1$
x_1	1	0	$1/4$	$-1/2$	0	0	3
x_2	0	1	$-1/8$	$3/4$	0	0	$3/2$
s_3	0	0	$3/8$	$-5/4$	1	0	$5/2$
s_4	0	0	$1/8$	$-3/4$	0	1	$1/2$

Sensitivity analysis

Example: Variations in c_1 . Requiring that $r_N \leq \Delta c_B^T B^{-1} N$ means requiring the elements in the z - row to be non-negative,

	x_1	x_2	s_1	s_2	s_3	s_4	Sol.
z	0	0	$\frac{3}{4} + \frac{1}{4}(\Delta c_1)$	$\frac{1}{2} - \frac{1}{2}(\Delta c_1)$	0	0	$21 + 3\Delta c_1$
x_1	1	0	$1/4$	$-1/2$	0	0	3
x_2	0	1	$-1/8$	$3/4$	0	0	$3/2$
s_3	0	0	$3/8$	$-5/4$	1	0	$5/2$
s_4	0	0	$1/8$	$-3/4$	0	1	$1/2$

which leads to the following intervals:

$$(1/4)\Delta c_1 \geq -(3/4) \quad \Rightarrow \Delta c_1 \geq -3$$

$$-(1/2)\Delta c_1 \geq -(1/2) \quad \Rightarrow \Delta c_1 \leq 1$$

Therefore,

► $-3 \leq \Delta c_1 \leq 1$
and thus $c_1 \in [2, 6]$.

► $z = 21 + 3\Delta c_1$.

Types of integer programming problems

Our starting point is a **linear programming problem**:

$$\begin{aligned} \text{(LP)} : \min. \quad & c^\top x \\ \text{s.t.:} \quad & Ax \leq b \\ & x \geq 0, \end{aligned}$$

where $A \in \mathbb{R}^{m \times n}$, $c \in \mathbb{R}^n$, $b \in \mathbb{R}^m$, and x a vector of n decision variables.

Integer programming (IP) problems have additional constraints on the **domain of x** .

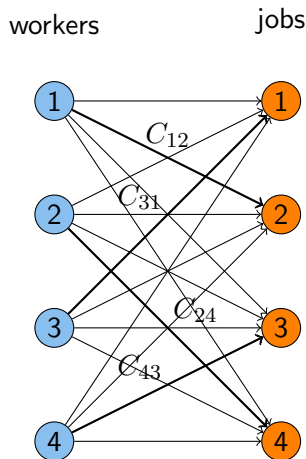
- ▶ **Integer Programming (IP)**: x must take integer values:
 $x \in \mathbb{Z}^n$;
- ▶ **Binary Integer Programming (BIP)**: x must be 0 or 1:
 $x \in \{0, 1\}$;
- ▶ **Mixed Integer Programming (MIP)**: some of the variables must take integer values: $x \in \mathbb{R}^q \times \mathbb{Z}^{n-q}$ or
 $x \in \mathbb{R}^q \times \{0, 1\}^{n-q}$.

The assignment problem

Problem statement:

- ▶ assign n jobs to n workers;
- ▶ one job associated to one worker;
- ▶ one worker associated to one job;
- ▶ it costs C_{ij} for worker i to execute job j .

Objective: find minimum cost assignment.



The assignment problem

Let $x_{ij} = 1$, if worker i is assigned to job j ; 0, otherwise, and $N = \{1, \dots, n\}$.

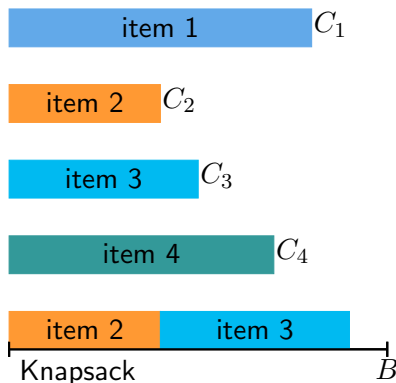
$$\begin{aligned} \text{(AP) : min. } & \sum_{i \in I} \sum_{j \in J} C_{ij} x_{ij} \\ \text{s.t.: } & \sum_{j \in N} x_{ij} = 1, \forall i \in M \\ & \sum_{i \in M} x_{ij} = 1, \forall j \in N \\ & x_{ij} \in \{0, 1\}, \forall i, \forall j \in N. \end{aligned}$$

The 0-1 knapsack problem

Problem statement:

- ▶ n items available for selection;
- ▶ it costs A_i to select i .
- ▶ each item i has value C_i ;
- ▶ The available budget is B .

Objective: find **maximum-valued selection** of items that does not exceed **budget**.



The 0-1 knapsack problem

Let $x_i = 1$, if item i is selected; 0, otherwise, and $N = \{1, \dots, n\}$.

$$\begin{aligned} \text{(0-1 KP)} : \max. \quad & \sum_{i=1}^n C_i x_i \\ \text{s.t.:} \quad & \sum_{i=1}^n A_i x_i \leq B \\ & x_i \in \{0, 1\}, \forall i \in N. \end{aligned}$$

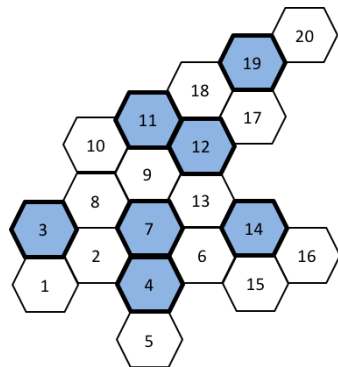
The set covering problem

Problem statement:

- ▶ A set of $M = \{1, \dots, m\}$ **regions must be served** by opening *service centres* (e.g., hospitals, schools, police stations);
- ▶ A centre can be opened at $N = \{1, \dots, n\}$ possible locations;
- ▶ If a centre is opened at location $j \in N$, then it serves a subset $S_j \subseteq M$ of regions and has opening cost C_j .

Objective: decide **where** to open the facilities so that **all regions are served** and the total opening cost is minimised.

The set covering problem: covering example



- ▶ Each **location** represents a candidate place for a **centre**;
- ▶ Once opened, the centre can only serve **immediate neighbours**.
- ▶ We have $M = \{1, \dots, 20\}$ and $N = \{3, 4, 7, 11, 12, 14, 19\}$.

In this case: $S_3 = \{1, 2, 3, 8\}$, $S_4 = \{2, 4, 5, 6, 7\}$, \dots

The set covering problem

To model the SCP as a BIP, we need a **0 – 1 incidence matrix** $A = [A_{ij}]_{m \times n}$ where $A_{ij} = 1$ if $i \in S_j$, $A_{ij} = 0$ otherwise.

Let $x_j = 1$ if facility is opened at location j ; $x_j = 0$, otherwise, and let $M = \{1, \dots, m\}$ and $N = \{1, \dots, n\}$.

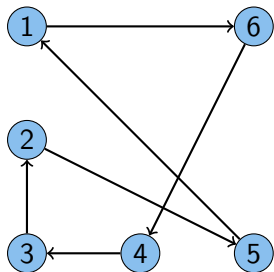
$$\begin{aligned} (\text{SCP}) : \min. \quad & \sum_{j \in N} C_j x_j \\ \text{s.t.:} \quad & \sum_{j \in N} A_{ij} x_j \geq 1, \forall i \in M \\ & x_j \in \{0, 1\}, \forall j \in N. \end{aligned}$$

Travelling salesman problem

Problem statement:

- ▶ A salesman must visit each of n cities **exactly once** and return to the starting city;
- ▶ It costs C_{ij} to travel from city i to city j ;

Objective: find a **least-cost tour**, i.e., an order in which the cities must be visited.



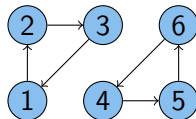
Travelling salesman problem

Let $x_{ij} = 1$ if city j is visited directly after city i , $x_{ij} = 0$ otherwise. Let $N = \{1, \dots, n\}$. We assume that x_{ii} is not defined for $i \in N$.

A naive model for the TSP could be:

$$\begin{aligned} (\text{TSP}) : \min. \quad & \sum_{i \in N} \sum_{j \in N} C_{ij} x_{ij} \\ \text{s.t.:} \quad & \sum_{j \in N \setminus \{i\}} x_{ij} = 1, \forall i \in N \\ & \sum_{i \in N \setminus \{j\}} x_{ij} = 1, \forall j \in N \\ & x_{ij} \in \{0, 1\}, \forall i, \forall j \in N : i \neq j \end{aligned}$$

- ▶ This is **exactly** the assignment problem.
- ▶ Also, solutions do not prevent **subtours**.



Travelling salesman problem

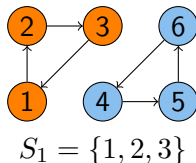
Preventing subtours: constraints that ensure full connectivity.

- ▶ Cutset constraints:

$$\sum_{i \in S} \sum_{j \in N \setminus S} x_{ij} \geq 1, \forall S \subset N, S \neq \emptyset$$

- ▶ Subtour elimination constraints:

$$\sum_{i \in S} \sum_{j \in S} x_{ij} \leq |S| - 1, \forall S \subset N, 2 \leq |S| \leq n - 1$$



Example for $S_1 = \{1, 2, 3\}$:

Cutset:

$$x_{14} + x_{24} + x_{34} + x_{15} + x_{25} + x_{35} + x_{16} + x_{26} + x_{36} \geq 1$$

Subtour elim.: $x_{12} + x_{23} + x_{31} \leq 2$