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Minimax objective functions

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The non-linear objective function

 $\min \max(x_1, x_2, \ldots, x_n)$

can be linearised by introducing a decision variable y and n linear constraints

 $\begin{array}{l} \min y \\ y \ge x_1 \\ y \ge x_2 \\ \vdots \\ y \ge x_n \end{array}$

You can see that y has to be no less than each of the x's, and the minimisation objective will force it down to take the value of the largest x. Similarly for the maxmin case. Unfortunately the same trick cannot be applied where we want to min min (x_1, x_2, \ldots, x_n) or max max (x_1, x_2, \ldots, x_n)

Absolute value

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Since

$$|\mathbf{x}| = egin{cases} \mathbf{x} & ext{if } \mathbf{x} \geq \mathbf{0} \ -\mathbf{x} & ext{if } \mathbf{x} \leq \mathbf{0} \end{cases}$$

it follows that $|\mathbf{x}| = \max\{\mathbf{x}, -\mathbf{x}\}$. Then min $|\mathbf{x}|$ can be also written as

min y $y \ge x$ $y \ge -x$

Linearisation and other modelling issues Ratio objective function

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The non-linear problem

$$\max f(x) = \frac{cx + c_0}{dx + d_0}$$
$$Ax \le b$$
$$x \ge 0$$

can be linearised by introducing $t = \frac{1}{dx+d_0}$. If we assume $dx + d_0 > 0$ (and hence t > 0) - if t < 0 the inequality signs must be reversed when we multiply by t we can write the non-linear problem

 $egin{aligned} \max f(x,t) &= ctx + c_0t \ Atx &\leq bt \ dtx + d_0t = 1 \ x,t \geq 0, \end{aligned}$

which becomes linear by introducing y = tx

$$egin{aligned} \max f(y,t) &= cy + c_0 t \ Ay - bt \leq 0 \ dy + d_0 t = 1 \ y,t \geq 0, \end{aligned}$$

Logical relations - NOT 4 | 14

The use of **binary variables** allows us to model logical relations as linear constraints

NOT

•



It can be easily modelled by constraint

 $x_A = 1 - x_B$, where $x_A, x_B \in \{0, 1\}$

Logical relations - OR

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Α	В	C
0	0	0
0	1	1
1	0	1
1	1	1

It can be modelled by constraints

 $x_{C} \ge x_{A}$ $x_{C} \ge x_{B}$ $x_{C} \le x_{A} + x_{B}$ $x_{A}, x_{B}, x_{C} \in \{0, 1\}$

Logical relations - AND

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А	В	С
0	0	0
0	1	0
1	0	0
1	1	1

It can be modelled by constraints

 $x_C \leq x_A$ $x_C \leq x_B$ $x_C \geq x_A + x_B - 1$ $x_A, x_B, x_C \in \{0, 1\}$

Logical relations - XOR

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В	C
0	0
1	1
0	1
1	0
	B 0 1 0 1

It can be modelled by constraints

$$x_C \ge x_A - x_B$$
$$x_C \ge x_B - x_A$$
$$x_C \le x_A + x_B$$
$$x_C \le 2 - x_A - x_B$$
$$x_A, x_B, x_C \in \{0, 1\}$$

The Bin packing problem 8 | 14

In the bin packing problem, items of different volumes must be packed into a finite number of bins or containers each of a fixed given volume in a way that minimises the number of bins used.

If there are m items (i = 1, ..., m), each of volume d_i , and n bins (j = 1, ..., n), each of capacity K_j , some obvious decision variables are

 $\mathbf{x}_{ij} = \begin{cases} \mathbf{1} & \text{if item } \mathbf{i} \text{ is inserted in bin } \mathbf{j} \\ \mathbf{0} & \text{otherwise.} \end{cases}$

There are two alternative formulations

Linearisation and other modelling issues The Bin packing problem

$$\min \sum_{y=1}^{n} y_{j}$$

$$\sum_{j=1}^{n} x_{ij} = 1 \text{ for } i = 1, \dots, m$$

$$\sum_{i=1}^{m} d_{i} x_{ij} \leq K_{j} y_{j} \text{ for } j = 1, \dots, m$$

$$x_{ij} \in \{0, 1\}, y_{j} \in \{0, 1\}$$

 $y_j = 1$ if the *j*th bin used, 0 otherwise.

min w

$$w \ge \sum_{j=1}^{n} j x_{ij}$$
 for $i = 1, \dots, m$
 $\sum_{j=1}^{n} x_{ij} = 1$ for $i = 1, \dots, m$
 $\sum_{i=1}^{m} d_i x_{ij} \le K_j$ for $j = 1, \dots, n$
 $x_{ij} \in \{0, 1\}, w \ge 0$

The value of w must be at least as large as the highest index number of the bins that are used.

Eliminating sub-tours

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As we might remember, in the TSP formulation sub-tours can be eliminated only by adding an exponential number of constraints. A alternative reasoning follows.

- Let arbitrarily fix one node as node 1.
- If there is a sub-tour that does not contain node **1**, there must be another sub-tour that contains node **1**.
- The idea is not to allow the existence of sub-tours that do not contain node **1**.
- In this way, we only allow complete tours.

Linearisation and other modelling issues **Eliminating sub-tours**





Eliminating sub-tours 12 | 14

For a direct graph $\mathcal{G} = (N, A)$, let's introduce variable u_i , where $u_1 = 1$ and $u_i \ge 1$ for each $i \in N$ and such that $u_i = k$ if i is the kth node visited in the cycle. If $x_{ij} = 1$ if arc (i, j) belongs to the Hamiltonian tour and 0 otherwise, a formulation is

$$\begin{split} \min \sum_{(i,j) \in A} c_{ij} x_{ij} \\ \sum_{i \in N \setminus \{j\}} x_{ij} = 1 & \forall j \in N \\ \sum_{j \in N \setminus \{i\}} x_{ij} = 1 & \forall i \in N \\ |N| x_{ij} + u_i - u_j \leq |N| - 1 & \forall (i,j) \in A, j \neq 1 \\ x_{ij} \in \{0,1\} & \forall (i,j) \in A \\ u_1 = 1 & \\ 2 \leq u_i \leq |N| & \forall i \in N, i \neq 1 \end{split}$$

Linearisation and other modelling issues **Eliminating sub-tours**



Why the following inequalities allow to eliminate sub-tours?

 $|N|x_{ij} + u_i - u_j \leq |N| - 1 \quad \forall (i,j) \in A, j \neq 1$

- If x satisfy the assignment constraints but it is not a Hamiltonian tour, then there are some sub-tours and one of them does not contain node 1. In these case the above inequalities are not satisfied.
- Instead, the constraints are satisfied by each Hamiltonian tour.

These constraints are on the order of $|\mathbf{N}|^2$, and are not exponential in number($2^{|\mathbf{N}|}$).

Linearisation and other modelling issues **Eliminating sub-tours**



