

INFRASTRUCTURE RAPID TRANSIT NETWORK DESIGN MODEL solved by BENDERS DECOMPOSITION

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1. Rapid Transit Design Problem

In recent years, new rail transit systems have been built, expanded or are being planned for construction.

Reasons:

- house spreading
- enlargement of urbanised areas
- traffic problem in the centres or in entrances of the cities
- reduction on average ground traffic speed

2. Infrastructure Network Design Model

Data:

- Let $N = \{i | i = 1, 2, \dots, n\}$ and $E = \{e_1, \dots, e_m\}$ two given sets of potential sites (stations) and edges (links), respectively.
- Each station i and edge e has an associated construction cost, $c_i, c_e \in \mathbb{R}^+$. Let C_{max} the available budget.
- Every feasible arc $a \in A = \{a_1, \dots, a_{2m}\} \subset N \times N$ has an associated length d_a .
- Travel patterns are given by the origin-destination matrix $G = (g^P)$, where g^P is the demand of the pair $p = (p^s, p^t) \in P = \{p_1, \dots, p_v\} \subset N \times N$ and P is the set of pairs of demand.
- Let u_{PRIV}^P the time needed to go from $p^s \in N$ to $p^t \in N$.

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Infrastructure Network Design Model

- Objective function: maximize trip coverage

$$\max \sum_{p \in P} g^p z^p$$

- Budget constraint

$$\sum_{e \in E} c_e x_e + \sum_{i \in N} c_i y_i \leq C_{max}$$

- Alignment location constraints

$$x_e \leq y_i, \quad e \in E, \quad i \in \{e^s, e^t\}$$

- Routing demand conservation constraints

$$\sum_{a \in \delta^+(p^s)} f_a^p = z^p, \quad p = (p^s, p^t) \in P$$

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$$\sum_{a \in \delta^+(p^t)} f_a^p = 0, \quad p = (p^s, p^t) \in P$$

$$\sum_{a \in \delta^-(k)} f_a^p - \sum_{a \in \delta^+(k)} f_a^p = 0, \quad k \notin \{p^s, p^t\}, \quad p = (p^s, p^t) \in P$$

- Location-Allocation constraints

$$f_a^p + f_{a'}^p \leq x_e, \quad p = (p^s, p^t) \in P, \quad e = a \text{ or } e = a'$$

- Splitting demand constraints

$$\sum_{a \in A} d_a f_a^p \leq u_{PRIV}^p z^p, \quad p = (p^s, p^t) \in P$$

- Binary constraints

$$x_e, \quad y_t, \quad f_a^p, \quad z^p \quad \in \{0, 1\}.$$

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Observation

Continuous Budget Constraint

$$\sum_{e \in E} c_e x_e + \sum_{i \in N} c_i y_i \leq C_{max}$$

can be replaced by the set of Discrete Budget Constraints:

$$\sum_{i \in N} y_i \leq N_{max}, \quad \sum_{e \in E} x_e \leq M_{max},$$

where N_{max} and M_{max} are de the maximum number of stations and edges that can be built. These polyhedrons are related by the equality:

$$C_{max} = \sum_{i=1}^{N_{max}} (L_{stations})_i + \sum_{j=1}^{M_{max}} (L_{edges})_j,$$

where $L_{stations}$ and L_{edges} are the costs lists in non-creasing order of the stations and edges sets respectively.

4. Benders Decomposition for INDM

Benders Decomposition for INDM

Master Problem:

$$\begin{aligned} \max \quad & \sum_{p \in P} g^p z^p + \mathbf{q}(x, y, z) \\ \text{s.t.} \quad & \sum_{e \in E} c_e x_e + \sum_{i \in N} c_i y_i \leq C_{max}, \quad e \in E, \quad i \in N \quad (1) \\ & x_e \leq y_i, \quad e \in E, \quad i \in \{e^s, e^t\} \\ & x_e, \quad y_i, \quad z^p \in \{0, 1\} \end{aligned}$$

Benders Decomposition for INDM

Subproblem:

$$\mathbf{q}(x, y, z) = \max \quad 1$$

$$\text{s.t.} \quad \sum_{a \in \delta^+(p^s)} f_a^p = \bar{z}^p, \quad p = (p^s, p^t) \in P$$

$$\sum_{a \in \delta^-(p^s)} f_a^p = 0, \quad p = (p^s, p^t) \in P$$

$$\sum_{a \in \delta^+(p^t)} f_a^p = 0, \quad p = (p^s, p^t) \in P$$

$$\sum_{a \in \delta^-(k)} f_a^p - \sum_{a \in \delta^+(k)} f_a^p = 0, \quad k \notin \{p^s, p^t\}, \quad p = (p^s, p^t) \in P$$

$$f_a^p + f_{a'}^p \leq \bar{x}_e, \quad p = (p^s, p^t) \in P, \quad e = a \text{ or } e = a'$$

$$\sum_{a \in A} d_a f_a^p \leq u_{PRIV}^p \bar{z}^p, \quad p = (p^s, p^t) \in P$$

$$f_a^p \in [0, 1]$$

(2)

Benders Decomposition for INDM

Benders Master Problem:

$$\max \quad \phi + \sum_{p \in P} g^p z^p$$

$$\text{s.t.} \quad \phi \leq \sum_{p \in P} \left(z^p \left(\alpha_{p^s}^p + U_{PRIV}^p \sum_{a \in A} \omega_a^p \right) + \sum_{e \in E} x_e \beta_e^p + \sum_{a \in A} \gamma_a^p \right)$$

$$0 \leq \sum_{p \in P} \left(z^p \left(\alpha_{p^s}^p + U_{PRIV}^p \sum_{a \in A} \omega_a^p \right) + \sum_{e \in E} x_e \beta_e^p + \sum_{a \in A} \gamma_a^p \right)$$

$$\sum_{e \in E} c_e x_e + \sum_{i \in N} c_i y_i \leq C_{max}, \quad e \in E, \quad i \in N$$

$$x_e \leq y_i, \quad e \in E, \quad i \in \{e^s, e^t\}$$

$$x_e, \quad y_i, \quad z^p \in \{0, 1\}$$

(3)

Disaggregated Benders Decomposition for INDM

Subproblem for each commodity p :

$$\begin{aligned} \mathbf{q}_p(x, y, z) = \max \quad & 1 \\ \text{s.t.} \quad & \sum_{a \in \delta^+(p^s)} f_a^p = \bar{z}^p, \\ & \sum_{a \in \delta^-(p^s)} f_a^p = 0, \\ & \sum_{a \in \delta^+(p^t)} f_a^p = 0, \\ & \sum_{a \in \delta^-(k)} f_a^p - \sum_{a \in \delta^+(k)} f_a^p = 0, \quad k \notin \{p^s, p^t\}, \\ & f_a^p + f_{a'}^p \leq \bar{x}_e, \quad e = a \text{ or } e = a' \\ & \sum_{a \in A} d_a f_a^p \leq u_{PRIV}^p \bar{z}^p, \\ & f_a^p \in [0, 1] \end{aligned} \tag{4}$$

Disaggregated Benders Decomposition for INDM

Benders Master Problem:

$$\begin{aligned} \max \quad & \sum_{p \in P} \phi^p + \sum_{p \in P} g^p z^p \\ \text{s.t.} \quad & \phi^p \leq z^p \left(\alpha_{p^s}^p + U_{PRIV}^p \sum_{a \in A} \omega_a^p \right) + \sum_{e \in E} x_e \beta_e^p + \sum_{a \in A} \gamma_a^p, \quad p = (p^s, p^t) \in P \\ & 0 \leq z^p \left(\alpha_{p^s}^p + U_{PRIV}^p \sum_{a \in A} \omega_a^p \right) + \sum_{e \in E} x_e \beta_e^p + \sum_{a \in A} \gamma_a^p, \quad p = (p^s, p^t) \in P \\ & \sum_{e \in E} c_e x_e + \sum_{i \in N} c_i y_i \leq C_{max}, \quad e \in E, \quad i \in N \\ & x_e \leq y_i, \quad e \in E, \quad i \in \{e^s, e^t\} \\ & x_e, \quad y_i, \quad z^p \in \{0, 1\} \end{aligned} \tag{5}$$

4. Preliminary computational results

Preliminary computational results

| Network | Nmax | Mmax | ObjValue | BB | BD | DBD |
|---------|------|------|----------|----------|-----------|-----------|
| N20-65 | 10 | 20 | 2386 | 2079.12 | 169.509 | 165.307 |
| | 15 | 30 | 4734 | 512.24 | 98.437 | 89.462 |
| | 18 | 40 | 6482 | 106.63 | 52.377 | 51.068 |
| | 20 | 65 | 7698 | 1.13 | 2.840 | 2.305 |
| N25-116 | 13 | 35 | 4131 | *1d | 1473.71 | 1407 |
| | 19 | 53 | 8022 | 82251.03 | 484.461 | 486.83 |
| | 23 | 71 | 11029 | 6446.85 | 371.05 | 703.96 |
| | 25 | 116 | 12557 | 6.66 | 33.68 | 32.86 |
| N30-166 | 15 | 50 | 5446 | *4d | 145340.58 | 143785.65 |
| | 23 | 76 | 11673 | *4d | 5079.85 | 4791.26 |
| | 27 | 101 | 15391 | *3d | 1409.71 | 1616.58 |
| | 30 | 166 | 18448 | 13.60 | 45.42 | 45.49 |
| N35-202 | 18 | 60 | 7571 | *5d | *5d | 354550.95 |
| | 26 | 93 | 14902 | *3d | 86966.48 | 81941.39 |
| | 32 | 123 | 21645 | *3d | 11365.28 | 9374.81 |
| | 35 | 202 | 24913 | 16.33 | 378.02 | 352.93 |
| N40-250 | 20 | 75 | - | *5d | *5d | *5d |
| | 30 | 115 | - | *3d | *3d | *3d |
| | 36 | 153 | 18448 | *3d | 261.27 | 239.42 |
| | 40 | 250 | 32426 | 24.82 | 487.36 | 500.67 |

- The more adjusted the upper bound, the bigger is the difference between the computational time of the three algorithms.
- In the majority of the cases, DBD is faster than BD but there are weird cases.
- DB constraints accelerates the resolution process.

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- Use Benders Decomposition to solve others similar problems
 - Design and locate the lines
 - Take into account transfer lines and other parameters

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Thanks for your attention!