

Project course: Optimization TM
The solution of a difficult problem
(facility location)

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Location of facilities which serve customers

- Potential sites: $\mathcal{J} = \{1, \dots, n\}$ (geographical locations)
- Existing customers: $\mathcal{I} = \{1, \dots, m\}$ (geographical locations)
- f_j = fixed cost of using depot j
- c_{ij} = transportation cost when customer i 's demand is fulfilled entirely from depot j

Decision problem:

- Which depots to open?
- Which depots to serve which customers, and how much?
- **Goal:** minimize cost
- **Assumption:** depots have unlimited capacity (to be removed)

Decision variables:

$$y_j = \begin{cases} 1, & \text{if depot } j \text{ is set up} \\ 0, & \text{otherwise} \end{cases}$$

x_{ij} = portion of customer i 's demand to be delivered from depot j

Uncapacitated facility location (UFL)

$$z_0^* = \min \sum_{i \in \mathcal{I}} \sum_{j \in \mathcal{J}} c_{ij} x_{ij} + \sum_{j \in \mathcal{J}} f_j y_j \quad \text{s.t.}$$

$$(1) \quad \sum_{j \in \mathcal{J}} x_{ij} = 1, \quad i \in \mathcal{I}$$

$$(2) \quad x_{ij} - y_j \leq 0, \quad i \in \mathcal{I}, j \in \mathcal{J}$$

$$(3) \quad x_{ij} \in [0, 1], \quad i \in \mathcal{I}, j \in \mathcal{J}$$

$$(4) \quad y_j \in \{0, 1\}, \quad j \in \mathcal{J}$$

(0) Minimize cost

(1) Deliver precisely the demand

(2) Deliver only from open depots

(3) x is the portion of the demand

(4) Do not partially open a depot

Suppose depots have limited capacity

d_i = demand of customer i ($D = \sum_{i \in I} d_i$)

b_j = capacity of depot j —if it is opened

Constraints:

$$\sum_{i \in I} d_i x_{ij} \leq b_j y_j, \quad j \in J \quad (5) \quad \Longleftrightarrow \quad x_{ij} \leq y_j, \quad \forall i, j)$$

\Longleftrightarrow replace (2) with (5)

Capacitated facility location (CFL)

$$(0) \quad z^* = \min \sum_{i \in \mathcal{I}} \sum_{j \in \mathcal{J}} c_{ij} x_{ij} + \sum_{j \in \mathcal{J}} f_j y_j$$

$$(1) \quad \sum_{j \in \mathcal{J}} x_{ij} = 1, \quad i \in \mathcal{I} \quad \text{s.t.}$$

$$(5) \quad \sum_{j \in \mathcal{J}} d_i x_{ij} - b_j y_j \leq 0, \quad j \in \mathcal{J}$$

$$(3) \quad x_{ij} \in [0, 1], \quad i \in \mathcal{I}, j \in \mathcal{J}$$

$$(4) \quad y_j \in \{0, 1\}, \quad j \in \mathcal{J}$$

Observation: Total capacity of open depots must cover the entire demand \implies an additional (redundant) constraint:

$$(1), (5) \implies \underbrace{\sum_{j \in \mathcal{J}} b_j y_j}_{\text{capacity}} \geq \sum_{i \in \mathcal{I}} \sum_{j \in \mathcal{J}} d_i x_{ij} = \sum_{i \in \mathcal{I}} d_i \sum_{j \in \mathcal{J}} x_{ij} = \sum_{i \in \mathcal{I}} d_i \cdot 1 = \underbrace{D}_{\text{demand}}$$

Trick: Exchange x_{ij} for w_{ij} in constraint (1) and in “half” the objective, add the constraints $x_{ij} = w_{ij}$, and let $0 \leq \alpha \leq 1$.

$$\begin{aligned}
 z^* = \min \quad & \alpha \sum_{i \in I} \sum_{j \in J} c_{ij} x_{ij} + (1 - \alpha) \sum_{i \in I} \sum_{j \in J} c_{ij} w_{ij} + \sum_{j \in J} f_j y_j \\
 \text{s.t.} \quad & \sum_{j \in J} w_{ij} = 1, \quad i \in I \tag{1} \\
 & \sum_{j \in J} d_i x_{ij} - b_j y_j \leq 0, \quad j \in J \tag{5} \\
 & \sum_{j \in J} b_j y_j \geq D, \tag{6} \\
 & w_{ij} - x_{ij} = 0, \quad i \in I, j \in J \tag{7} \\
 & x_{ij} \in [0, 1], \quad i \in I, j \in J \tag{3} \\
 & w_{ij} \geq 0, \quad i \in I, j \in J \tag{8} \\
 & y_j \in \{0, 1\}, \quad j \in J \tag{4}
 \end{aligned}$$

- Constraints (7) tie together (\mathbf{x}, \mathbf{y}) with w .

- Lagrangian relax these with multipliers λ_{ij}

\Leftarrow Lagrange function

$$L(\mathbf{x}, w, \mathbf{y}, \boldsymbol{\lambda}) = \left[\sum_{i \in \mathcal{I}} \sum_{j \in \mathcal{J}} \alpha c_{ij} x_{ij} + (1 - \alpha) c_{ij} w_{ij} + \underbrace{\lambda_{ij} (w_{ij} - x_{ij})}_{\text{penalty}} \right] + \sum_{j \in \mathcal{J}} f_j y_j$$

$$= \sum_{i \in \mathcal{I}} \sum_{j \in \mathcal{J}} (\alpha c_{ij} - \lambda_{ij}) x_{ij} + \sum_{j \in \mathcal{J}} f_j y_j + \sum_{i \in \mathcal{I}} \sum_{j \in \mathcal{J}} [(1 - \alpha) c_{ij} + \lambda_{ij}] w_{ij}$$

- Subproblem (for fixed value of $\boldsymbol{\lambda}$):

Minimize the Lagrange function under constraints (1), (5), (6), (3), (8) & (4).

Separates into one in (\mathbf{x}, \mathbf{y}) and $|\mathcal{I}|$ in w .

Subproblem in x and y :

$$q^{xy}(\lambda) = \min_{x, y} \sum_{i \in \mathcal{I}} \sum_{j \in \mathcal{J}} [\alpha_{cij} - \lambda_{ij}] x_{ij} + \sum_{j \in \mathcal{J}} f_j y_j$$

s.t.

$$\sum_{j \in \mathcal{J}} b_j y_j \geq D, \quad (6)$$

$$\sum_{i \in \mathcal{I}} d_i x_{ij} \leq b_j y_j, \quad j \in \mathcal{J} \quad (5)$$

$$x_{ij} \in [0, 1], \quad i \in \mathcal{I}, j \in \mathcal{J} \quad (3)$$

$$y_j \in \{0, 1\}, \quad j \in \mathcal{J} \quad (4)$$

For every y -solution (such that $\sum_{j \in \mathcal{J}} b_j y_j \geq D$) we have:

• If $y_j = 0$ then $x_{ij} = 0, i \in \mathcal{I}$

• If $y_j = 1$ then $\sum_{i \in \mathcal{I}} d_i x_{ij} \leq b_j$

Value [in (x, y) -subproblem] of opening depot j

That is: letting $y_j = 1$ ($|\mathcal{J}|$ continuous knapsack problems)

$$[\text{CKSP}_j] \quad v_j(\lambda) = f_j + \min_x \sum_{i \in \mathcal{I}} [\alpha_{ij} - \lambda_{ij}] x_{ij}$$

$$\text{s.t.} \quad \sum_{i \in \mathcal{I}} d_i x_{ij} \leq b_j$$

$$x_{ij} \in [0, 1], \quad i \in \mathcal{I}$$

\Leftarrow Projection onto y -space (a 0/1 knapsack problem)

$$[0/1\text{-KSP}] \quad q^{xy}(\lambda) = \min_y \sum_{j \in \mathcal{J}} v_j(\lambda) \cdot y_j$$

$$\text{s.t.} \quad \sum_{j \in \mathcal{J}} b_j y_j \geq D,$$

$$y_j \in \{0, 1\}, \quad j \in \mathcal{J}$$

Solving the continuous knapsack problems [CKSP]^j

- Sort $\frac{\alpha_{c_{ij}} - \lambda_{ij}}{d_i} > 0, i \in \mathcal{I}$, in increasing order
 - \implies indices $\{i_1, i_2, \dots, i_m\}, m \leq |\mathcal{I}|$.
 - If $m = 0$ then $x_{ij} = 0, i \in \mathcal{I}$. Else, let $k = 1$ and:
 - Let $x_{ik_j} = \min\{1; b_j - \sum_{s=1}^{k-1} d_i x_{is_j}\}$ and let $k := k + 1$ until $\sum_{s=1}^k d_i x_{is_j} = b_j$ or $k = m$.
 - Solution fulfills $\sum_{i \in \mathcal{I}} d_i x_{ij} = b_j$ and $x_{ij} \in [0, 1], i \in \mathcal{I}$.
 - $v_j(\lambda) = f_j + \min_{|\mathcal{I}|} \sum_{k=1}^{|\mathcal{I}|} \sum_{j \in \mathcal{J}} [\alpha_{c_{ik_j}} - \lambda_{ik_j}] x_{ik_j}$.
- Solving 0/1 knapsack problems**
- Not polynomial. Solve with Branch & Bound (CPLEX).
- Solution:** $y_j(\lambda) \in \{0, 1\}, j \in \mathcal{J}$.
- $x_{ij}(\lambda) = 0, i \in \mathcal{I}, \text{ if } y_j(\lambda) = 0.$
 - $x_{ij}(\lambda) = x_{ij}$ by the above, $i \in \mathcal{I}, \text{ if } y_j(\lambda) = 1.$

Subproblem in w
 $(|\mathcal{I}| \text{ semi-assignment problems}):$

$$[\text{SAP}] \quad \left[\begin{array}{l} \min_w \sum_{j \in \mathcal{J}} [(1 - \alpha)c_{ij} + \lambda_{ij}] w_{ij} \\ \text{s.t.} \quad \sum_{j \in \mathcal{J}} w_{ij} = 1, \quad w_{ij} \geq 0, \quad j \in \mathcal{J} \\ \sum_{i \in \mathcal{I}} = (\lambda)^w \end{array} \right]$$

Solving semi-assignment problem ?

(special case of [CKSP]):

- Find ℓ_i such that $(1 - \alpha)c_{i\ell_i} + \lambda_{i\ell_i} = \min_{j \in \mathcal{J}} \{(1 - \alpha)c_{ij} + \lambda_{ij}\}$.
- Let $w_{i\ell_i}(\lambda) = 1, w_{ij}(\lambda) = 0, j \neq \ell_i$.

Value of relaxed problem for fixed value of λ

$$q(\lambda) = \underbrace{q^{xy}(\lambda)}_{\text{difficult}} + \underbrace{q^w(\lambda)}_{\text{simple}}$$

- Can show that $q(\lambda) \leq q^*$ for all $\lambda \in \mathbb{R}^{|\mathcal{I}| \times |\mathcal{J}|}$ (weak duality)
- λ_{ij} is the penalty for violating $w_{ij} = x_{ij}$
- Find best underestimate of $q^* \iff$ find “optimal” values of penalties λ_{ij}
- That is: $\max_{\lambda \in \mathbb{R}^{|\mathcal{I}| \times |\mathcal{J}|}} q(\lambda) \leq q^*$ (most often $\max_{\lambda \in \mathbb{R}^{|\mathcal{I}| \times |\mathcal{J}|}} q(\lambda) > z^*$, not strong duality)

How to find better value of λ_{ij} ?

$$\text{Penalty: } \min \dots \sum_{i \in I} \sum_{j \in J} \lambda_{ij} (w_{ij} - x_{ij})$$

- If $w_{ij}(\lambda) > x_{ij}$ (more expensive to violate constraint) \implies Increase value of λ_{ij}

- If $w_{ij}(\lambda) < x_{ij}$ (more expensive to violate constraint) \implies Decrease value of λ_{ij}

- Iterative method (subgradient algorithm) to find optimal penalties λ^* :

$$\lambda_{ij}^{t+1} = \lambda_{ij}^t + \rho^t [w_{ij}(\lambda^t) - x_{ij}(\lambda^t)], \quad t = 0, 1, \dots$$

where $\rho^t > 0$ is a step length, decreasing with t

- Use feasibility heuristic from every $[x(\lambda^t), w(\lambda^t), y(\lambda^t)]$ to yield a feasible solution to CFL (open more depots, send only from open depots, $x = w, \dots$). Example: Benders' subproblem!

Example: $|I| = 4$, $|J| = 3$, $\alpha = \frac{1}{2}$

$$(c_{ij}) = \begin{bmatrix} 6 & 2 & 4 \\ 16 & 2 & 4 \\ 10 & 12 & 4 \end{bmatrix}, (f_j) = \begin{bmatrix} 11 \\ 16 \\ 21 \end{bmatrix}, (d_i) = \begin{bmatrix} 6 \\ 4 \\ 8 \\ 5 \end{bmatrix}, (b_j) = \begin{bmatrix} 12 \\ 10 \\ 13 \end{bmatrix}$$

$$q_{xy}(\lambda) = \min \sum_{j=1}^3 v_j(\lambda) \cdot y_j \quad \text{s.t.} \quad 12y_1 + 10y_2 + 13y_3 \geq 23$$

$$y \in \{0, 1\}^3 \quad \left| \quad \text{Let } (\lambda_{ij}^t) = \begin{bmatrix} 7 & 3 & 0 \\ 0 & 10 & 5 \\ 0 & 2 & 7 \\ 0 & 0 & 5 \end{bmatrix}$$

Observe: implies that $y_3 = 1$ must hold.

$$\begin{aligned}
 & \underbrace{\dots \iff \dots}_{\text{(next page)}} \\
 & q^{xy}(\lambda) = \min \quad 5y_1 + 8.875y_2 + 18y_3 \\
 & \text{s.t.} \quad 12y_1 + 10y_2 + 13y_3 \geq 23, \quad \mathbf{y} \in \{0, 1\}^3
 \end{aligned}$$

$$\begin{aligned}
v_1(\lambda_t) = 11 + \min & -4x_{11} - 2x_{21} + 3x_{31} + 5x_{41} \\
\text{s.t.} & 6x_{11} + 4x_{21} + 8x_{31} + 5x_{41} \leq 12, \quad x_{\cdot 1} \in [0, 1]^4 \\
& x_{11} = x_{21} = 1, \quad x_{31} = x_{41} = 0, \quad v_1(\lambda_t) = 5 \quad \Longleftarrow \\
\\
v_2(\lambda_t) = 16 + \min & x_{12} - 6x_{22} - x_{32} - x_{42} \\
\text{s.t.} & 6x_{12} + 4x_{22} + 8x_{32} + 5x_{42} \leq 10, \quad x_{\cdot 2} \in [0, 1]^4 \\
& x_{22} = x_{42} = 1, \quad x_{32} = \frac{1}{8}, \quad x_{12} = 0, \quad v_2(\lambda_t) = 8.875 \quad \Longleftarrow \\
\\
v_3(\lambda_t) = 21 + \min & 2x_{13} + 0x_{23} + 3x_{33} - 3x_{43} \\
\text{s.t.} & 6x_{13} + 4x_{23} + 8x_{33} + 5x_{43} \leq 13, \quad x_{\cdot 3} \in [0, 1]^4 \\
& x_{23} = x_{43} = 1, \quad x_{13} = x_{33} = 0, \quad v_3(\lambda_t) = 18 \quad \Longleftarrow
\end{aligned}$$

Solution to (x, y) problem for $\lambda = \lambda_t$

$$h(\lambda_t) = (1, 0, 1)_{\top}, x = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix}, h^x(\lambda_t) = 5 + 0 + 18 = 23$$

w -problem separates into one for each customer i

$$q_i^w(\lambda_t) = \sum_{j=1}^3 q_i^j(\lambda_t), \text{ where } (1 - \alpha) = \frac{2}{1}$$

$$\min = q_i^w(\lambda_t) = \sum_{j=1}^3 [(1 - \alpha)c_{ij} + \lambda_t^j] w_{ij}$$

$$\text{s.t. } \sum_{j=1}^3 w_{ij} = 1, w_{ij} \geq 0, j = 1, 2, 3$$

$$\begin{aligned}
& \min 10w_{11} + w_{12} + 2w_{13} \\
& \text{s.t. } w_{11} + w_{12} + w_{13} = 1, \quad w_{1j} \geq 0, \quad j = 1, 2, 3 \\
& \Leftrightarrow w_{12}(\lambda_t) = 1, \quad w_{11}(\lambda_t) = w_{13}(\lambda_t) = 0, \quad q_1^m(\lambda_t) = 1 \\
& \min 4w_{21} + 14w_{22} + 4w_{23} \\
& \text{s.t. } w_{21} + w_{22} + w_{23} = 1, \quad w_{2j} \geq 0, \quad j = 1, 2, 3 \\
& \Leftrightarrow w_{21}(\lambda_t) = 1, \quad w_{22}(\lambda_t) = w_{23}(\lambda_t) = 0, \quad q_2^m(\lambda_t) = 4 \\
& \min 13w_{31} + 3w_{32} + 3w_{33} \\
& \text{s.t. } w_{31} + w_{32} + w_{33} = 1, \quad w_{3j} \geq 0, \quad j = 1, 2, 3 \\
& \Leftrightarrow w_{32}(\lambda_t) = w_{33}(\lambda_t) = \frac{1}{2}, \quad w_{31}(\lambda_t) = 0, \quad q_3^m(\lambda_t) = 3 \\
& \min 5w_{41} + 13w_{42} + 7w_{43} \\
& \text{s.t. } w_{41} + w_{42} + w_{43} = 1, \quad w_{4j} \geq 0, \quad j = 1, 2, 3 \\
& \Leftrightarrow w_{41}(\lambda_t) = 1, \quad w_{42}(\lambda_t) = w_{43}(\lambda_t) = 0, \quad q_4^m(\lambda_t) = 5
\end{aligned}$$

Solution to w problem

$$m(\lambda_t) = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & \frac{1}{2} & 0 \end{bmatrix}, \quad q^m(\lambda_t) = 13,$$

$$b(\lambda_t) = q^{xy}(\lambda_t) + q^m(\lambda_t) = 35$$

New λ vector (e.g., $p_t = 8$):

$$\leftarrow z^* \geq 35$$

$$\lambda_{t+1} = \lambda_t + p_t [\lambda_t x - \lambda_t m]$$

$$= \begin{bmatrix} 7 - p_t & 3 & 5 & p_t \\ p_t & 10 & 2 + \frac{2}{p_t} & 7 \\ 0 & 2 - p_t & \frac{2}{p_t} & 5 - p_t \end{bmatrix} = \begin{bmatrix} -1 & 3 & 5 & 8 \\ 8 & 10 & 6 & 7 \\ 0 & -6 & 4 & -3 \end{bmatrix}$$

Feasible solution $\iff x(\lambda_t) = w(\lambda_t)$? No \iff
Feasibility heuristic

Idea: Open depots given by $y(\lambda_t) \iff y_H = y(\lambda_t) = (1, 0, 1)^T$.
 Send only from open depots ($y_H^j = 0 \iff x_H^{ij} = 0, \forall i$).
 Fulfill demand but do not violate capacity restrictions:

$$\text{Let } x_H = \begin{bmatrix} \frac{7}{2} & 0 & \frac{12}{5} \\ 1 & 0 & 0 \\ 0 & 0 & 1 \\ \frac{1}{2} & 0 & \frac{1}{2} \end{bmatrix} \iff$$

$$z_H = 6 \cdot \frac{7}{2} + 4 \cdot \frac{12}{5} + 2 + 6 + 10 \cdot \frac{7}{2} + 4 \cdot \frac{7}{2} + 11 + 21 = 52 + \frac{6}{1} \iff z^* \in [35, 52 + \frac{6}{1}] = [q(\lambda_t), z_H] \text{ (not very good interval)}$$

- Choice of step lengths (ρ^t) later (subgradient optimization, convergence to an optimal value of λ)
- Feasibility heuristics can be made more or less sophisticated
- There are more ways in which to Lagrangian relax *continuous* constraints in an optimization problem
- E.g.: Lagrangian relax (1) or (5) (with multipliers $\mu_i \in \mathbb{R}$ resp. $\nu_j \in \mathbb{R}_+$) in the original formulation (CFL)

- There are also other methods for solving CFL. Consider for example the fact that for fixed \mathbf{y} , the remaining problem over \mathbf{x} is very simple (a transportation problem). Algorithms can be based on only adjusting \mathbf{y} , always optimizing over \mathbf{x} for each \mathbf{y} . (We say that we *project* the problem onto the y variables.)
- This is the Benders' subproblem (more on the Benders algorithm later).
- Solve Benders' subproblem at $\mathbf{y} = (1, 0, 1)^T$:

$$\mathbf{h}(\mathbf{y})\mathbf{x} = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \end{pmatrix} \cdot \mathbf{x}$$
- Total cost: 50 (32 + 18).

- Alternative solution: $(0, 1, 1)^T$. Benders' subproblem:

$$\mathbf{x}(\mathbf{y}) = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1/3 & 0 & 0 & 1 \\ 2/3 & 1 & 1 & 1 \end{pmatrix}.$$

- Total cost: 53 (37 + 16).

- $\mathbf{y}_* = (1, 0, 1)^T$; $z_* = 50$.

- Note that we have (probably) not solved the dual problem to optimality, so we do not know what the size of the duality gap is.