

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)

Birgit Grohe

3 September 2008

0-0

- (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 ($D = \sum_{i \in \mathcal{I}} d_i$)

b_j = capacity of depot j —if it is opened

Constraints:

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

\Rightarrow replace (2) with (5)

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)

$$z_0^* = \min \sum_{i \in \mathcal{I}} \sum_{j \in \mathcal{J}} c_{ij} x_{ij} + \sum_{j \in \mathcal{J}} f_j y_j \quad (0)$$

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

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

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

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

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

$$\begin{aligned} z^* = \min \quad & \alpha \sum_{i \in \mathcal{I}} \sum_{j \in \mathcal{J}} c_{ij} x_{ij} + (1 - \alpha) \sum_{i \in \mathcal{I}} \sum_{j \in \mathcal{J}} c_{ij} w_{ij} + \sum_{j \in \mathcal{J}} f_j y_j \\ \text{s.t.} \quad & \sum_{j \in \mathcal{J}} w_{ij} = 1, \quad i \in \mathcal{I} \end{aligned} \quad (1)$$

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

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

$$w_{ij} - x_{ij} = 0, \quad i \in \mathcal{I}, j \in \mathcal{J} \quad (7)$$

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

$$w_{ij} \geq 0, \quad i \in \mathcal{I}, j \in \mathcal{J} \quad (8)$$

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

Capacitated facility location (CFL)

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

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

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

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

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

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

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

Subproblem in \mathbf{x} and \mathbf{y} :

$$\begin{aligned} q_{xy}(\lambda) = \min_{x,y} \quad & \sum_{i \in \mathcal{I}} \sum_{j \in \mathcal{J}} [\alpha c_{ij} - \lambda_{ij}] x_{ij} + \sum_{j \in \mathcal{J}} f_j y_j \\ \text{s.t.} \quad & \sum_{j \in \mathcal{J}} b_j y_j \geq D, \end{aligned} \quad (6)$$

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

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

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

For every \mathbf{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 $\sum_{i \in \mathcal{I}} d_i x_{ij} \leq b_j$

- Constraints (7) tie together (\mathbf{x}, \mathbf{y}) with \mathbf{w} .
- Lagrangian relax these with multipliers λ_{ij}

\Rightarrow Lagrange function

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

- 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 \mathbf{w} .

Solving the continuous knapsack problems [CKSP_j]

Greedy Algorithm:

- Sort $\frac{\alpha c_{ij} - \lambda_{ij}}{d_i} < 0$, $i \in \mathcal{I}$, in increasing order
- \Rightarrow 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 $x_{i_k j} = \min\{1; b_j - \sum_{s=1}^{k-1} d_i x_{i_s j}\}$ and let $k := k + 1$ until $\sum_{s=1}^k d_i x_{i_s j} = b_j$ or $k = m$.
 - Solution fulfills $\sum_{i \in \mathcal{I}} d_i x_{ij} = b_j$ and $x_{ij} \in [0, 1]$, $i \in \mathcal{I}$.
 - $v_j(\boldsymbol{\lambda}) = f_j + \min \sum_{k=1}^{|\mathcal{I}|} \sum_{j \in \mathcal{J}} [\alpha c_{i_k j} - \lambda_{i_k j}] x_{i_k j}$.

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

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

$$\begin{aligned} [\text{CKSP}_j] \quad v_j(\boldsymbol{\lambda}) = f_j + \min_x & \sum_{i \in \mathcal{I}} [\alpha c_{ij} - \lambda_{ij}] x_{ij} \\ \text{s.t.} & \sum_{i \in \mathcal{I}} d_i x_{ij} \leq b_j \\ & x_{ij} \in [0, 1], \quad i \in \mathcal{I} \end{aligned}$$

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

$$\begin{aligned} [0/1-\text{KSP}] \quad q_{xy}(\boldsymbol{\lambda}) = \min_y & \sum_{j \in \mathcal{J}} v_j(\boldsymbol{\lambda}) \cdot y_j \\ \text{s.t.} & \sum_{j \in \mathcal{J}} b_j y_j \geq D, \\ & y_j \in \{0, 1\}, \quad j \in \mathcal{J} \end{aligned}$$

Subproblem in w

($|\mathcal{I}|$ semi-assignment problems):

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

Solving semi-assignment problem i

- 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 $w_{i\ell_i}(\boldsymbol{\lambda}) = 1$, $w_{ij}(\boldsymbol{\lambda}) = 0$, $j \neq \ell_i$.

Solving 0/1 knapsack problem

Not polynomial. Solve with Dynammic Programming or Branch & Bound (CPLEX).

Solution:

$$y_j(\boldsymbol{\lambda}) \in \{0, 1\}, \quad j \in \mathcal{J}.$$

$$x_{ij}(\boldsymbol{\lambda}) = 0, \quad i \in \mathcal{I}, \text{ if } y_j(\boldsymbol{\lambda}) = 0.$$

$$x_{ij}(\boldsymbol{\lambda}) = x_{ij} \text{ by the above, } i \in \mathcal{I}, \text{ if } y_j(\boldsymbol{\lambda}) = 1.$$

How to find better value of λ_{ij} ?

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

- If $w_{ij}(\boldsymbol{\lambda}) > x_{ij}(\boldsymbol{\lambda}) \implies$ Increase value of λ_{ij} (more expensive to violate constraint)
- If $w_{ij}(\boldsymbol{\lambda}) < x_{ij}(\boldsymbol{\lambda}) \implies$ Decrease value of λ_{ij} (more expensive to violate constraint)
- Iterative method (subgradient algorithm) to find optimal penalties $\boldsymbol{\lambda}^*$:

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

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

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

Value of relaxed problem for fixed value of $\boldsymbol{\lambda}$

$$q(\boldsymbol{\lambda}) = \underbrace{q_{xy}(\boldsymbol{\lambda})}_{\text{difficult}} + \underbrace{q_w(\boldsymbol{\lambda})}_{\text{simple}}$$

- Can show that $q(\boldsymbol{\lambda}) \leq q^*$ for all $\boldsymbol{\lambda} \in \mathbb{R}^{|\mathcal{I}| \times |\mathcal{J}|}$ (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: $\max_{\boldsymbol{\lambda} \in \mathbb{R}^{|\mathcal{I}| \times |\mathcal{J}|}} q(\boldsymbol{\lambda}) \leq q^*$ (most often $\max_{\boldsymbol{\lambda} \in \mathbb{R}^{|\mathcal{I}| \times |\mathcal{J}|}} q(\boldsymbol{\lambda}) < z^*$, not strong duality)

$$\underbrace{\dots \Rightarrow \dots}_{\dots} q_{xy}(\boldsymbol{\lambda}) = \min \quad 5y_1 + 8.875y_2 + 18y_3$$

$$(\text{next page}) \quad \text{s.t.} \quad 12y_1 + 10y_2 + 13y_3 \geq 23, \quad \mathbf{y} \in \{0, 1\}^3$$

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

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

$$q_{xy}(\boldsymbol{\lambda}) = \min \quad \sum_{j=1}^3 v_j(\boldsymbol{\lambda}) \cdot y_j$$

$$\text{s.t.} \quad 12y_1 + 10y_2 + 13y_3 \geq 23$$

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

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

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

$$\mathbf{y}(\lambda^t) = (1, 0, 1)^T, \mathbf{x}(\lambda^t) = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}, q_{xy}(\lambda^t) = 5 + 0 + 18 = 23$$

w-problem separates into one for each customer i

$$q_w(\lambda^t) = \sum_{i=1}^4 q_w^i(\lambda^t), \quad \text{where} \quad (1 - \alpha = \frac{1}{2})$$

$$\begin{aligned} q_w^i(\lambda^t) &= \min \quad \sum_{j=1}^3 [(1 - \alpha)c_{ij} + \lambda_{ij}^t] w_{ij} \\ \text{s.t.} \quad &\sum_{j=1}^3 w_{ij} = 1, \quad w_{ij} \geq 0, \quad j = 1, 2, 3 \end{aligned}$$

$$\begin{aligned} v_1(\lambda^t) &= 11 + \min \quad -4x_{11} - 2x_{21} + 3x_{31} + 5x_{41} \\ \text{s.t.} \quad &6x_{11} + 4x_{21} + 8x_{31} + 5x_{41} \leq 12, \quad x_{.1} \in [0, 1]^4 \\ \implies &x_{11} = x_{21} = 1, \quad x_{31} = x_{41} = 0, \quad v_1(\lambda^t) = 5 \end{aligned}$$

$$\begin{aligned} v_2(\lambda^t) &= 16 + \min \quad x_{12} - 6x_{22} - x_{32} - x_{42} \\ \text{s.t.} \quad &6x_{12} + 4x_{22} + 8x_{32} + 5x_{42} \leq 10, \quad x_{.2} \in [0, 1]^4 \\ \implies &x_{22} = x_{42} = 1, \quad x_{32} = \frac{1}{8}, \quad x_{12} = 0, \quad v_2(\lambda^t) = 8.875 \end{aligned}$$

$$\begin{aligned} v_3(\lambda^t) &= 21 + \min \quad 2x_{13} + 0x_{23} + 3x_{33} - 3x_{43} \\ \text{s.t.} \quad &6x_{13} + 4x_{23} + 8x_{33} + 5x_{43} \leq 13, \quad x_{.3} \in [0, 1]^4 \\ \implies &x_{23} = x_{43} = 1, \quad x_{13} = x_{33} = 0, \quad v_3(\lambda^t) = 18 \end{aligned}$$

Solution to w problem

$$w(\lambda^t) = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} \\ 1 & 0 & 0 \end{bmatrix}, \quad q_w(\lambda^t) = 13,$$

$$q(\lambda^t) = q_{xy}(\lambda^t) + q_w(\lambda^t) = 35 \quad \text{New } \lambda \text{ vector (e.g., } \rho_t = 8\text{:)} \\ \implies z^* \geq 35$$

$$\begin{aligned} \lambda^{t+1} &= \lambda^t + \rho_t [\mathbf{w}(\lambda^t) - \mathbf{x}(\lambda^t)] \\ &= \begin{bmatrix} 7 - \rho_t & \rho_t & 0 \\ 3 & 10 & 2 - \rho_t \\ 5 & 2 + \frac{\rho_t}{2} & \frac{\rho_t}{2} \\ \rho_t & 7 & 5 - \rho_t \end{bmatrix} = \begin{bmatrix} -1 & 8 & 0 \\ 3 & 10 & -6 \\ 5 & 6 & 4 \\ 8 & 7 & -3 \end{bmatrix} \end{aligned}$$

$$\begin{aligned} q_w^1(\lambda^t) &= \min \quad 10w_{11} + w_{12} + 2w_{13} \\ \text{s.t.} \quad &w_{11} + w_{12} + w_{13} = 1, \quad w_{1j} \geq 0, \quad j = 1, 2, 3 \\ \implies &w_{12}(\lambda^t) = 1, \quad w_{11}(\lambda^t) = w_{13}(\lambda^t) = 0, \quad q_w^1(\lambda^t) = 1 \end{aligned}$$

$$\begin{aligned} q_w^2(\lambda^t) &= \min \quad 4w_{21} + 14w_{22} + 4w_{23} \\ \text{s.t.} \quad &w_{21} + w_{22} + w_{23} = 1, \quad w_{2j} \geq 0, \quad j = 1, 2, 3 \\ \implies &w_{21}(\lambda^t) = 1, \quad w_{22}(\lambda^t) = w_{23}(\lambda^t) = 0, \quad q_w^2(\lambda^t) = 4 \end{aligned}$$

$$\begin{aligned} q_w^3(\lambda^t) &= \min \quad 13w_{31} + 3w_{32} + 3w_{33} \\ \text{s.t.} \quad &w_{31} + w_{32} + w_{33} = 1, \quad w_{3j} \geq 0, \quad j = 1, 2, 3 \\ \implies &w_{32}(\lambda^t) = w_{33}(\lambda^t) = \frac{1}{2}, \quad w_{31}(\lambda^t) = 0, \quad q_w^3(\lambda^t) = 3 \end{aligned}$$

$$\begin{aligned} q_w^4(\lambda^t) &= \min \quad 5w_{41} + 13w_{42} + 7w_{43} \\ \text{s.t.} \quad &w_{41} + w_{42} + w_{43} = 1, \quad w_{4j} \geq 0, \quad j = 1, 2, 3 \\ \implies &w_{41}(\lambda^t) = 1, \quad w_{42}(\lambda^t) = w_{43}(\lambda^t) = 0, \quad q_w^4(\lambda^t) = 5 \end{aligned}$$

- 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, e.g. Bender's algorithm using Bender's subproblem (later).

Feasible solution $\iff x(\lambda^t) = w(\lambda^t)$? No \implies Feasibility heuristic

Idea: Open depots given by $y(\lambda^t) \implies y^H = y(\lambda^t) = (1, 0, 1)^T$.

Send only from open depots ($y_j^H = 0 \implies x_{ij}^H = 0, \forall i$).

Fulfill demand but do not violate capacity restrictions:

$$\text{Let } \mathbf{x}^H = \begin{bmatrix} \frac{1}{6} & 0 & \frac{5}{6} \\ 1 & 0 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} \implies$$

$$z^H = 6 \cdot \frac{1}{6} + 4 \cdot \frac{5}{6} + 2 + 6 + 10 + 11 + 21 = 52 + \frac{1}{3}$$

$$\implies z^* \in [35, 52 + \frac{1}{3}] = [q(\lambda^t), z^H] \quad (\text{not very good interval})$$