

Uncapacitated Metric Facility Location Problem (UMFL).

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1 Uncapacitated Metric Facility Location Problem (UMFL).

- Let \mathcal{L} be a set of locations.
- Let $\mathcal{F} \subset \mathcal{L}$ be a set of potential facility locations.
- Let $\mathcal{C} \subset \mathcal{L}$ be a set of clients (cities).
- Let $c : \mathcal{L} \times \mathcal{L} \rightarrow \mathbb{R}^+ \forall i, j \in \mathcal{L}$ be a distance function. Alternatively think of the function as describing the cost of assigning city j to facility i .
- Let $f : \mathcal{F} \rightarrow \mathbb{R}^+ \forall i \in \mathcal{F}$ be a cost function describing the cost of opening a facility at i .
- Let $\phi : \mathcal{C} \rightarrow \mathcal{F} \forall j \in \mathcal{C}$ be an assignment function. I.e. $\phi(j) = i$ if city j is assigned to facility i .

Problem: Determine a set of facilities to open and an assignment of all cities to the open facilities that minimizes the total opening and distance cost.

Notation 1.1. $f(i) = f_i$. $c(i, j) = c_{ij}$

The problem is uncapacitated as there is no bound on how many cities an open facility can serve. It is metric as c is defining a metric. This especially means:

$$\begin{aligned} c_{ij} &= c_{ji} & \forall i, j \in L \\ c_{ij} &\leq c_{ik} + c_{kj} & \forall i, j, k \in L \quad (\text{triangle inequality}) \end{aligned}$$

2 IP Formulation.

$$\begin{aligned} \text{Let } y_i &= \begin{cases} 1 & \text{if a facility is opened at } i \in \mathcal{F} \\ 0 & \text{otherwise} \end{cases} \\ \text{Let } x_{ij} &= \begin{cases} 1 & \text{if } \phi(j) = i \\ 0 & \text{otherwise} \end{cases} \end{aligned}$$

$$\text{Minimize } Z(x, y) = F(x, y) + C(x, y) = \sum_{i \in \mathcal{F}} f_i \cdot y_i + \sum_{\substack{j \in \mathcal{C} \\ i \in \mathcal{F}}} c_{ij} \cdot x_{ij}$$

s.t.

$$\sum_{i \in \mathcal{F}} x_{ij} = 1 \quad \forall j \in \mathcal{C} \quad (1)$$

$$x_{ij} \leq y_i \quad \forall i \in \mathcal{F}, j \in \mathcal{C} \quad (2)$$

$$x_{ij}, y_i \in \{0, 1\} \quad \forall i \in \mathcal{F}, j \in \mathcal{C} \quad (3)$$

The constraint (1) ensures that all cities gets assigned to a facility. The constraint (2) ensures that the assigned facilities are open. The LP-relaxation is obtained by changing (3) into:

$$x_{ij}, y_i \geq 0 \quad \forall i \in \mathcal{F}, j \in \mathcal{C}$$

the upper bound is unnecessary.

3 Approximation Algorithm for UMFL.

Algorithm Idea.

The algorithm will take an optimal solution for the LP-relaxation (x^*, y^*) and change this into a feasible solution for the IP problem. This will consist of two operations:

$$(x^*, y^*) \xrightarrow{\text{filter}} (x, y) \xrightarrow{\text{round}} (\hat{x}, \hat{y})$$

where (x, y) and (\hat{x}, \hat{y}) are feasible solutions to the LP- and IP-problem respectively. A high-level description of the algorithm steps are:

1. Greedily chose the city, c_{min} which "is cheapest" i.e has the lowest overall distance cost.
2. Chose the cheapest facility location α among those "fractionally opened" locations which the city is "fractionally assigned" to.
3. Open a facility f_α at α completely.
4. Assign the city completely to the f_α facility (and only to this facility).
5. Assign all cities which are "fractionally assigned" to some facility locations in the "neighbourhood" of c_{min} completely to the the f_α facility.
6. Update collection of unassigned cities and repeat from step 1)

The reason for the filtering is step 5). If some city "far away" is assigned with a very small $x_{ij} > 0$ value to a neighbouring facility, it will be very costly to assign this city to the newly opened facility. This is avoided by ensuring $x_{ij} = 0$ if c_{ij} is "large".

Filtering.

Definition 3.1. Let

$$\Delta_j = \sum_{i \in \mathcal{F}} c_{ij} \cdot x_{ij} \quad \forall j \in \mathcal{C}$$

Definition 3.2. $\forall j \in \mathcal{C}$ let $B_j = \{i \in \mathcal{F} \mid c_{ij} < 2\Delta_j\}$ This describes a neighbourhood or "ball" around each city containing facility locations with "small" distances.

Lemma 3.3. *Given a solution (x', y') of the LP-problem, there exists a feasible solution to the LP-problem (x, y) such that:*

i) $x_{ij} > 0 \Rightarrow c_{ij} < 2\Delta_j$ (I.e. c_{ij} "is small")

ii) $Z(x, y) \leq 2 Z(x', y')$

Proof. $\forall i \in \mathcal{F}, j \in \mathcal{C}$ let

$$x_{ij} = \begin{cases} \frac{x'_{ij}}{\sum_{i \in B_j} x'_{ij}} & \text{if } i \in B_j \\ 0 & \text{otherwise} \end{cases}$$

$$y_i = \min\{1, 2y'_i\}$$

Observation 3.4. $x_{ij} \leq 1 \quad \forall i \in \mathcal{F}, j \in \mathcal{C}$

Claim 3.5. (x, y) is a feasible solution fulfilling Lemma 3.3 i).

Check for constraint (1)

$$\sum_{i \in \mathcal{F}} x_{ij} = \sum_{i \in B_j} x_{ij} + \sum_{i \notin B_j} x_{ij} = \sum_{i \in B_j} \frac{x'_{ij}}{\sum_{i \in B_j} x'_{ij}} + \sum_{i \notin B_j} 0 = 1 + 0 = 1$$

Check for constraint (2)

Case 1. $y_i = 1$ follows from observation 3.4

Case 2. $y_i = 2y'_i$. We have $\sum_{i \in \mathcal{F}} x'_{ij} = 1$ as (x'_{ij}, y'_i) is a solution to the LP-problem. Interpret x'_{ij} as a probability distribution for "assigning j to i " and c_{ij} as a "distance" random variable.

Theorem 3.6 (Markov Inequality). *Let X be a positive, random variable. Let $a > 0$ then*

$$Pr [X \geq a] \leq \frac{\mathbb{E}[X]}{a}$$

Using the Markov Inequality we get:

$$\begin{aligned}
\sum_{i \notin B_j} x'_{ij} &= Pr [c_{ij} \geq 2\Delta_j] \leq \frac{\Delta_j}{2\Delta_j} = \frac{1}{2} \\
&\Downarrow \\
\sum_{i \in B_j} x'_{ij} &\geq \frac{1}{2} \\
&\Downarrow \\
x_{ij} &\leq 2x'_{ij} \leq 2y'_i = y_i
\end{aligned}$$

The last line following from the definition of x_{ij} and from (x', y') being a feasible solution. This proves claim 3.5 and per construction part i) of Lemma 3.3

$$\begin{aligned}
Z(x, y) &= F(x, y) + C(x, y) \\
F(x, y) &= \sum_{i \in \mathcal{F}} f_i \cdot y_i \leq \sum_{i \in \mathcal{F}} f_i \cdot 2y'_i = 2F(x', y') \\
C(x, y) &= \sum_{\substack{j \in \mathcal{C} \\ i \in \mathcal{F}}} c_{ij} \cdot x_{ij} \leq \sum_{\substack{j \in \mathcal{C} \\ i \in \mathcal{F}}} c_{ij} \cdot 2x'_{ij} = 2C(x', y') \\
&\Downarrow \\
Z(x, y) &\leq 2Z(x', y')
\end{aligned}$$

This proves part ii) of Lemma 3.3

□

Algorithm.

Let (x', y') denote the constructed solution to the IP-problem.

Step 1. Solve the relaxed LP-problem getting optimal solution (x^*, y^*) .

Step 2. Filter $(x^*, y^*) \rightarrow (x, y)$.

Step 3. Define $\Delta_j = \sum_{i \in \mathcal{F}} c_{ij} x_{ij}$ and $B_j = \{i \in \mathcal{F} \mid c_{ij} < \Delta_j\}$.

Observation 3.7. No factor 2 in definition of B_j and $\Delta_j \leq 2 \cdot \Delta_j^*, \forall j \in \mathcal{C}$

Step 4. While $\mathcal{C} \neq \emptyset$ do

- Chose minimal overall cost city:

$$j \leftarrow \min_j \Delta_j$$

- Consider neighbourhood B_j . Let α be the facility location $i \in B_j$ with smallest opening cost (f_α is minimum.)
 - Open facility at α ($y'_\alpha = 1$).
 - Assign city j to α ($\phi(j) = \alpha$, $x'_{ij} = 1$ for $i = \alpha$ and $x'_{ij} = 0$ for $i \neq \alpha$)
 - Update $\mathcal{C} \leftarrow \mathcal{C} \setminus \{j\}$.
- Consider all other neighbourhoods $\overline{B_j}$ for which $B_j \cap \overline{B_j} \neq \emptyset \Rightarrow \exists \bar{i} \in \mathcal{F} : \bar{i} \in B_j$ and $\bar{i} \in \overline{B_j}$
 - Assign city \bar{j} to α ($\phi(\bar{j}) = \alpha$, $x'_{i\bar{j}} = 1$ for $i = \alpha$ and $x'_{i\bar{j}} = 0$ for $i \neq \alpha$)
 - Update $\mathcal{C} \leftarrow \mathcal{C} \setminus \{\bar{j}\}$.

Step 5. Output $\{\alpha \mid y'_\alpha = 1\}$ and ϕ .

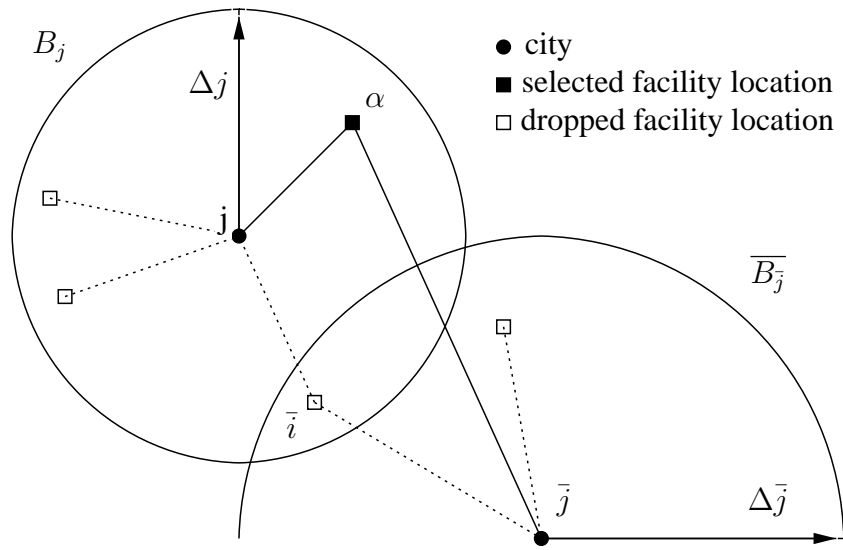


Figure 1: Assigning facility location to cities

Algorithm Analysis.

Claim 3.8. *The algorithm is a 6-approximation.*

Proof.

Termination and Feasibility: The number of cities is final and in each iteration at least one city is removed from the set of unassigned cities. The algorithm returns a feasible solution as each city has been assigned to an open facility location.

Opening Cost: Consider a round of the algorithm choosing city j .

Using the choice of α and that (x, y) is a filtered solution we have for all facility locations in B_j :

$$\sum_{i \in B_j} f_i \cdot y_i \geq \sum_{i \in B_j} f_\alpha \cdot y_i = f_\alpha \cdot \sum_{i \in B_j} y_i \geq f_\alpha \cdot \sum_{i \in B_j} x_{ij} = f_\alpha = \text{opening cost of algorithm.}$$

Let $\{\overline{B}_1, \overline{B}_2, \dots, \overline{B}_n\}$ be all the \overline{B}_j sets intersecting with B_j . Define a union of disjoint sets:

$$\overline{B} = \bigcup_{i,k \in 1..n} (\overline{B}_i \setminus \bigcup_{k < i} \overline{B}_k)$$

We have for the facility locations in $\overline{B} \setminus B_j$:

$$\sum_{i \in \overline{B} \setminus B_j} f_i \cdot y_i \geq 0 = \text{opening cost of algorithm.}$$

The algorithm "touches" each facility location exactly once, either selecting or dropping it \Rightarrow summing over all algorithm rounds and using Lemma 3.3 we get:

$$\text{Summed opening cost of algorithm} \leq \sum_{i \in \mathcal{F}} f_i \cdot y_i = F(x, y) \leq 2 F(x^*, y^*) \quad (4)$$

Connection Cost: For all cities we either have

- a) The city, j is assigned to a facility in its own neighbourhood: \Rightarrow connection cost for $j \leq \Delta j$
- b) The city, \bar{j} is assigned to a facility in the neighbourhood of another city, $j \Rightarrow$ connection cost for $\bar{j} \leq \underbrace{\Delta \bar{j}}_{\text{to get to } B_j} + \underbrace{\Delta j}_{\text{to get to } j} + \underbrace{\Delta j}_{\text{to get to location } \phi(\bar{j})} \leq 3\Delta \bar{j}$ (see Figure 1 on the previous page)

Using a) and b) and observation 3.7 we get:

$$C(x' y') = \sum_{\substack{j \in \mathcal{C} \\ i \in \mathcal{F}}} c_{ij} \cdot x'_{ij} \leq \sum_{j \in \mathcal{C}} 3\Delta j \leq \sum_{j \in \mathcal{C}} 6\Delta j^* = 6 \cdot \sum_{\substack{j \in \mathcal{C} \\ i \in \mathcal{F}}} c_{ij} \cdot x^*_{ij} = 6 C(x^*, y^*) \quad (5)$$

(4) and (5) gives: Algorithm cost $\leq 2 F(x^*, y^*) + 6 C(x^*, y^*) \leq 6 Z(x^*, y^*) \leq 6 OPT_{UMFL}$ \square