

Randomized Rounding for Uncapacitated Facility Location¹

- In the *uncapacitated facility location* (UFL) problem, we are given a set of clients C , a set of facilities F , and a distance function d on $C \cup F$ that satisfies the triangle inequality. We are also given *facility opening costs* $\{f_i\}_{i \in F}$. The task is to “open” a subset $S \subseteq F$ of facilities that minimizes the sum of the *facility cost* $\text{cost}_F(S) = \sum_{i \in S} f_i$ and the *connection cost* $\text{cost}_C(S) = \sum_{j \in C} d(j, S)$, where $d(j, S)$ is the smallest distance from j to any point in S . This problem is NP-hard, and we look at an approximation algorithm for it that uses the primal and the dual linear programs for it, as follows. In the primal, variables y_i denote whether $i \in S$, and x_{ij} denotes whether client j connects to facility i .

Primal:	Dual:
$\min \sum_{i \in F} f_i y_i + \sum_{j \in C} \sum_{i \in F} x_{ij} d(i, j)$	$\max \sum_{j \in C} v_j$
$\sum_{i \in F} x_{ij} \geq 1 \quad \forall j \in C$	$\sum_{j \in C} w_{ij} \leq f_i \quad \forall i \in F$
$y_i - x_{ij} \geq 0 \quad \forall j \in C, i \in F$	$v_j - w_{ij} \leq d(i, j) \quad \forall i \in F, j \in C$
$x_{ij} \geq 0, y_i \geq 0 \quad \forall j \in C, i \in F$	$v_j \geq 0, w_{ij} \geq 0 \quad \forall i \in F, j \in C$

We study a randomized algorithm that rounds a solution to the primal with the help of a solution to the dual. We first obtain an expected approximation ratio of $(1 + 3/e) < 2.11$, and then improve it to $(1 + 2/e) < 1.74$.

- We solve the primal LP to get (x, y) . We use some convenient notation:

$$\text{lp}_F := \sum_{i \in F} f_i y_i ; \quad \text{lp}_C := \sum_{j \in C} \sum_{i \in F} x_{ij} d(i, j) ; \quad C_j := \sum_{i \in F} d(i, j) x_{ij} \text{ for a client } j \in C$$

so that $\text{lp} = \text{lp}_F + \text{lp}_C$ and $\text{lp}_C = \sum_{j \in C} C_j$. We also solve the dual to get (v, w) . By complementary slackness², for any $i \in F, j \in C, x_{ij} > 0 \implies v_j - w_{ij} = d(i, j) \implies d(i, j) \leq v_j$. That is, in (x, y) , each client j only uses facilities within distance v_j of itself. We call this property *v-closeness*.

Remark: We can avoid solving the dual. We only need that (x, y) is *v-close*, and $\sum_{j \in C} v_j \leq \text{lp}$. For this, it actually suffices to set, for each $j \in C, v_j := \max \{d(i, j) \mid x_{ij} > 0\}$.

- **A deterministic start.** We first describe a simple deterministic rounding from the *v-close* fractional solution (x, y) to a $3v$ -close integral solution, i.e. $S \subseteq F$ such that $d(j, S) \leq 3v_j, \forall j \in C$. This is a “filtering” step, where we find a subset $D \subseteq C$ of clients that are far from each other. To do this, we pick the client j_0 with minimum v_{j_0} into D , and put any clients that share facilities with j_0 into the *child set* of j_0 , called Chld_{j_0} , which we then discard. We repeat this until we run out of clients. To complete the algorithm, we open the cheapest facility in each $N(j_0) : j_0 \in D$.

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²Assuming the reader knows this terminology from the course’s lecture notes.

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1: procedure FILTERING FOR UFL( $F \cup C, \{f_i\}_{i \in F}, (x, y)$ ):
2:    $U \leftarrow C, D \leftarrow \emptyset \triangleright U$  is the set of clients for which we have not decided
3:    $\forall j \in C, N(j) \leftarrow \{i \in F \mid x_{ij} > 0\}$ 
4:   while  $U \neq \emptyset$  do
5:     Pick  $j_0 \in U$  with minimum  $v_{j_0}$ 
6:      $D \leftarrow D + j_0$ 
7:      $\text{Chld}_{j_0} \leftarrow \{j \in C \mid N(j) \cap N(j_0) \neq \emptyset\}$ 
8:      $U \leftarrow U \setminus \text{Chld}_{j_0}$ 
9:   return  $D$ 
10: For  $D$  constructed above,  $S \leftarrow \{i_0 = \operatorname{argmin}_{i \in N(j_0)} f_i \mid j_0 \in D\}$ 
11: return  $S$ 

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Analysis. Consider a client $j \in \text{Chld}_{j_0}, j_0 \in D$. Let i_0 be the cheapest facility in $N(j_0)$. Then

$$d(j, S) \leq d(j, i_0) \leq d(j, j_0) + d(j_0, i_0) \leq d(j, i) + d(i, j_0) + d(j_0, i_0)$$

for some $i \in N(j) \cap N(j_0)$. By v -closeness of (x, y) , $d(j, i) \leq v_j$, and $d(i, j_0), d(j_0, i_0) \leq v_{j_0}$. Also by the filtering algorithm, since $j \in \text{Chld}_{j_0}, v_{j_0} \leq v_j$. So $d(j, S) \leq 3v_j$. We can show that we already have a 4-approximation.

Exercise: Show that $\sum_{i \in S} f_i \leq \text{lp}_F$, and hence conclude that $\text{cost}_F(S) + \text{cost}_C(S) \leq 4\text{lp}$.

When we randomize, we use this 4-approximation as our worst-case “backup”. That is, in most cases, we get $\mathbf{Exp}[d(j, S)] \leq C_j$ but, in a few bad cases, we rely on the deterministic bound $d(j, S) \leq 3v_j$.

- **Randomization.** Our randomized algorithm starts with the filtering process and obtains D . But instead of opening the cheapest facility in each $N(j_0) : j_0 \in D$, we open *one* i_0 in each $N(j_0)$ with probability proportional to x_{ij_0} . We also consider facilities outside all $N(j_0)$'s, i.e. $R := F \setminus \cup_{j_0 \in D} N(j_0)$. We independently open each $i \in R$ with probability y_i .

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1: procedure RANDOMIZED ROUNDING FOR UFL( $D \subseteq C$  from filtering, and  $(x, y)$ ):
2:    $S \leftarrow \emptyset$ 
3:   for  $j_0 \in D$  do
4:     Pick one  $i_0 \in N(j_0)$  as per distribution  $(x_{ij_0})_{i \in N(j_0)} \triangleright \sum_{i \in F} x_{ij_0} = 1$ 
5:      $S \leftarrow S + i_0$ 
6:   for  $i \in R$  do
7:     With probability  $y_i, S \leftarrow S + i$ 
8:   return  $S$ 

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Analysis of Randomized Rounding. For facility costs,

$$\mathbf{Exp}[\text{cost}_F(S)] \leq \sum_{j_0 \in D} \mathbf{Exp}[f_{i_0}] + \sum_{i \in R} f_i \Pr[i \in S] = \sum_{j_0 \in D} \sum_{i \in N(j_0)} f_i x_{ij_0} + \sum_{i \in R} f_i y_i \leq \text{lp}_F \quad (1)$$

To analyze the connection costs, fix a client j such that $j \in \text{Chld}_{j_0}, j_0 \in D$. Our algorithm always opens some $i_0 \in N(j_0)$, so by the same argument as the deterministic algorithm, $d(j, S)$ is at most $3v_j$. However, we hope that some facility in $N(j)$ is also open, allowing us to bound $d(j, S)$ by v_j instead.

Using the pairwise disjointness of the sets $\{N(\ell)\}_{\ell \in D} \cup \{R\}$, we partition $N(j)$ into sets $\{S_\ell := N(j) \cap N_\ell\}_{\ell \in D} \cup \{R_j := N(j) \cap R\}$. For $\ell \in D$, E_ℓ be the event that we open some facility in S_ℓ , and $p_\ell := \Pr[E_\ell] = \sum_{i \in S_\ell} x_{i\ell}$. Also, for each $i \in R_j$, E_i be the event that i is open, and $p_i := \Pr[E_i] = y_i$. Observe that all the events $\{E_\ell\}_{\ell \in D} \cup \{E_i\}_{i \in R_j}$ are pairwise independent. Let B_j be the bad event that *none* of these events occur. Dropping the subscript for a fixed j ,

$$\Pr[B] = \left(\prod_{\ell \in D} (1 - p_\ell) \right) \left(\prod_{i \in R_j} (1 - p_i) \right) \leq e^{-\left(\sum_{\ell \in D} p_\ell + \sum_{i \in R_j} p_i\right)} \quad (2)$$

For ease of analysis, we make the following ‘‘completeness’’ assumption: $\forall j \in C, i \in N(j), x_{ij} = y_i$, i.e. clients use fractional facilities either as much as possible or not at all. This assumption can be made true by creating an equivalent instance where it holds.

Exercise: Given an LP solution (x, y) on the instance \mathcal{I} , give a polynomial time algorithm to construct a new instance \mathcal{I}' and an LP solution (x', y') on \mathcal{I}' so that $\text{cost}_{\mathcal{I}}(x, y) = \text{cost}_{\mathcal{I}'}(x', y')$, and the latter is complete; and also if (x, y) is v -close then so is (x', y') .

Under this assumption, for $\ell \in D, i \in S_\ell, x_{i\ell} = y_i = x_{ij}$; and for $i \in R_j, y_i = x_{ij}$. So we get

$$\sum_{\ell \in D} p_\ell + \sum_{i \in R_j} p_i = \sum_{\ell \in D} \sum_{i \in S_\ell} x_{i\ell} + \sum_{i \in R_j} y_i = \sum_{i \in N(j)} x_{ij} = 1$$

So (2) gives $\Pr[B] \leq 1/e$. Appealing to the backup cost when B occurs, we get

$$\mathbf{Exp}[d(j, S)] \leq \sum_{\ell \in D} \mathbf{Exp}[d(j, S_\ell) \mid E_\ell] p_\ell + \sum_{i \in R_j} \mathbf{Exp}[d(j, i) \mid E_i] p_i + 3v_j/e \quad (3)$$

For $\ell \in D$, conditioned on E_ℓ occurring, $i \in S_\ell$ is open with probability $x_{i\ell}/p_\ell$. So the first term above becomes

$$\sum_{\ell \in D} \mathbf{Exp}[d(j, S_\ell) \mid E_\ell] p_\ell = \sum_{\ell \in D} \left(\sum_{i \in S_\ell} d(j, i) x_{i\ell}/p_\ell \right) p_\ell = \sum_{\ell \in D} \sum_{i \in S_\ell} x_{ij} d(j, i)$$

where the last step is by our completeness assumption. Also, for $i \in R_j, \mathbf{Exp}[d(j, i) \mid E_i] \leq d(j, i)$ and $p_i = y_i = x_{ij}$. So the RHS in (3) becomes at most

$$\sum_{\ell \in D} \sum_{i \in S_\ell} d(j, i) x_{ij} + \sum_{i \in R_j} d(j, i) x_{ij} + 3v_j/e \leq \sum_{i \in N(j)} d(j, i) x_{ij} + 3v_j/e = C_j + 3v_j/e$$

by definition of $N(j)$. So summing over all $j \in C$,

$$\mathbf{Exp}[\text{cost}_C(S)] \leq \sum_{j \in C} C_j + (3/e) \sum_{j \in C} v_j = \text{lp}_C + 3\text{lp}/e \quad (4)$$

Adding facility costs from (1), $\mathbf{Exp}[\text{cost}_F(S) + \text{cost}_C(S)] \leq \text{lp}_F + \text{lp}_C + 3\text{lp}/e = (1 + 3/e)\text{lp}$.

- In fact, one can do a little better. We can write, for a $j \in C$, $\mathbf{Exp}[d(j, S)] \leq \mathbf{Exp}[d(j, S) \mid \overline{B_j}](1 - 1/e) + \mathbf{Exp}[d(j, S) \mid B_j]/e$. So, intuitively, the lp_C term in (4) should be safe to multiply by $(1 - 1/e)$. To formalize this, we would need to show

Claim 1. $\forall j \in C, \mathbf{Exp}[d(j, S) \mid \overline{B_j}] \leq C_j$

This is **not** trivial to prove but, since it is intuitively believable, we believe it for the purpose of this note. Let us rewrite (4) accordingly.

$$\mathbf{Exp}[\text{cost}_S(C)] \leq (1 - 1/e)\text{lp}_C + \sum_{j \in C} \mathbf{Exp}[d(j, S) \mid B_j]/e \leq (1 + 3/e)\text{lp} - \text{lp}_C/e \quad (4')$$

We now pursue better bounds for $\mathbf{Exp}[d(j, S) \mid B_j]$ and use them with (4').

Exercise: Prove [Claim 1](#), or read and understand its proof from [1].

Improved Filtering. Notice that our bounds on $\mathbf{Exp}[d(j, S) \mid \overline{B_j}]$ and $\mathbf{Exp}[\text{cost}_F(S)]$ are ensured by the randomized rounding, as long as $\{N(\ell)\}_{\ell \in D}$ is pairwise disjoint. Contrarily, the bounds on $\mathbf{Exp}[d(j, S) \mid B_j]$ depend on the construction of D itself. So we alter our filtering procedure to get a new D that still has the above disjointness, but also gives better bounds on $\mathbf{Exp}[d(j, S) \mid B_j]$.

Observe that a $j_0 \in D$ never relies on the backup of $3v_{j_0}$; rather, because the randomized rounding opens some $i_0 \in N(j_0)$ with probability x_{ij_0} , $\mathbf{Exp}[d(j_0, S)] = C_{j_0}$, which could be much smaller than v_{j_0} . Connection costs of the form $C_{j_0} : j_0 \in D$ appear many times in our bounds; so perhaps we should pick vertices into D by minimum C_j 's, rather than by minimum v_j 's? But, we do not want to abandon the benefits of v -closeness. So we strike a natural compromise: we pick vertices into D by minimum $v_j + C_j$. That is, we replace Line 5 of the filtering algorithm with the following:

Pick $j_0 \in U$ with minimum $v_{j_0} + C_{j_0}$

Analysis with Improved Filtering. Our proof relies on the key lemma

Lemma 1. For any $j \in C$, $\mathbf{Exp}[d(j, S) \mid B_j] \leq 2v_j + C_j$.

By the lemma, $\sum_{j \in C} \mathbf{Exp}[d(j, S) \mid B_j] \leq \sum_{j \in C} (2v_j + C_j) = 2\text{lp} + \text{lp}_C$. So from (4') and (1), $\mathbf{Exp}[\text{cost}_F(S) + \text{cost}_C(S)] \leq \text{lp}_F + (1 - 1/e)\text{lp}_C + (2\text{lp} + \text{lp}_C)/e = (1 + 2/e)\text{lp}$.

Proof of Lemma 1. Fix $j \in \text{Chld}_{j_0}, j_0 \in D$. Drop the subscript from B_j . By the improved filtering, $v_{j_0} + C_{j_0} \leq v_j + C_j$. Also let i_0 be the facility in $N(j_0)$ that is opened during randomized rounding. Since $j \in \text{Chld}_{j_0}, N(j) \cap N(j_0) \neq \emptyset$, so for some $i \in N(j) \cap N(j_0)$, $d(j, S) \leq d(j, i_0) \leq d(j, i) + d(i, j_0) + d(j_0, i_0)$. Since our target quantity is $2v_j + C_j$, we want to use v -closeness on two of these terms, and bound the remaining term using some connection cost. We proceed via two cases.

Case 1. This is the case where $\exists i \in N(j) \cap N(j_0), d(i, j_0) \leq C_{j_0}$. Choosing this i and using v -closeness, $d(j, i) + d(i, j_0) + d(j_0, i_0) \leq v_j + (C_{j_0} + v_{j_0}) \leq v_j + (C_j + v_j) = 2v_j + C_j$.

Case 2. This is the case where $\forall i \in N(j) \cap N(j_0), d(i, j) > C_{j_0}$. In this case, we show that

$$\mathbf{Exp}[d(j_0, i_0) \mid B] \leq C_{j_0} \quad (5)$$

which gives, by linearity of expectation,

$$\mathbf{Exp}[d(j, i) + d(i, j_0) + d(j_0, i_0)] \leq v_j + (v_{j_0} + C_{j_0}) \leq 2v_j + C_j$$

We now prove (5). We know that $\mathbf{Exp}[d(j_0, i_0)] = C_{j_0}$. We are in the case where all facilities in $N(j) \cap N(j_0)$ are farther away from j than this unconditioned expected cost. So conditioned on B i.e. when they are all closed, the expected cost should not get worse. Formalizing this:

$$\begin{aligned} C_{j_0} = \mathbf{Exp}[d(j_0, i_0)] &= \sum_{i \in N(j_0)} d(i, j_0) x_{ij_0} \\ &> \text{(Case 2)} \quad C_{j_0} \underbrace{\sum_{i \in N(j_0) \cap N(j)} x_{ij_0}}_{=:p} + (1-p) \sum_{i \in N(j_0) \setminus N(j)} d(i, j_0) x_{ij_0} / (1-p) \\ &= C_{j_0} p + (1-p) \sum_{i \in N(j_0) \setminus N(j)} d(i, j_0) \mathbf{Pr}[i_0 = i \mid N(j) \cap S = \emptyset] \\ &= C_{j_0} p + (1-p) \mathbf{Exp}[d(j_0, i_0) \mid B] \end{aligned}$$

Rearranging, we get

$$(1-p)C_{j_0} > (1-p) \mathbf{Exp}[d(j_0, i_0) \mid B] \implies \mathbf{Exp}[d(j_0, i_0) \mid B] \leq C_{j_0}$$

□

Ponder This: *What happens if we filter by picking clients with minimum C_j ?*

Notes

This algorithm appears in the paper [1] by Chudak and Shmoys. The initial deterministic idea, that yields a 4-approximation, earlier appeared in a paper [2] by Shmoys, Tardos, and Aardal, and this is detailed in lecture note 5 of the course. [2] also used a different randomization to improve the 4 to 3.16, and gave a simple derandomization for it.

References

- [1] F. A. Chudak and D. B. Shmoys. Improved approximation algorithms for the uncapacitated facility location problem. *SIAM Journal on Computing*, 33(1):1–25, 2003.
- [2] D. B. Shmoys, É. Tardos, and K. Aardal. Approximation algorithms for facility location problems. In *Proceedings of the twenty-ninth annual ACM symposium on Theory of computing*, pages 265–274, 1997.