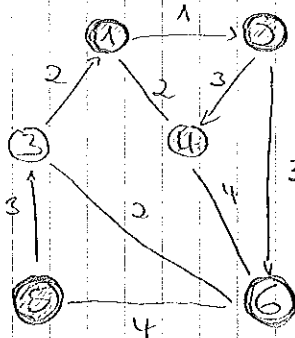


### 3. The Uncapacitated Facility Location Problem

3.1. The general Uncapacitated Fac. Loc. Prob.

3.1.1. (Sibel & Schmidt (1985)) & Def. (Uncapacitated Fac. Loc. Problem):



$G = (V, E, A)$  Network  
 $I = V$  customers (clients)  
 $J \subseteq V$  potential facilities  
 $d_{ij} \in \mathbb{R}_{\geq 0}, i, j \in A$  transportation costs  $i \rightarrow j$   
 $f_i \in \mathbb{R}_{\geq 0}, i \in I$  setup costs  
 $A \subseteq I \times J$  Uncapacitated Facility Loc. Prob.

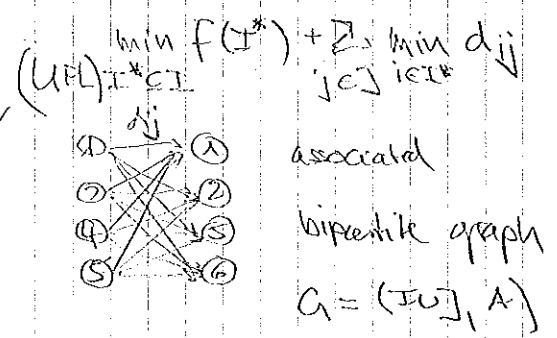
$I = \{1, 2, 5, 6\}$

$J = \{1, 2, 4, 5\}$

$D = (d_{ij}) =$

	1	2	5	6	$\Sigma$	$p =$	$p <$
1	0	1	8	4	13	18	15
2	5	0	7	3	15	20	
4	2	3	8	4	17	22	
5	6	3	0	4	13	18	

$f = (f_i) = (5, 5, 5, 5)$



3.1.2. Prop. (IP-formulation for the UFL, Paliuska (1965))

$y_i \in \{0, 1\}, i \in I$  facility setup variables  
 $x_{ij} \in \{0, 1\}, j \in J$  client assignment variables

(UFL)  $\min \sum_{i \in I} f_i y_i + \sum_{j \in A} d_{ij} x_{ij}$   
 (i)  $\sum_{i \in I} x_{ij} \geq 1 \quad \forall j \in J$   
 (ii)  $y_i \geq x_{ij} \quad \forall j \in A$   
 (iii)  $y_i \in \mathbb{Z}_+ \quad \forall i \in I$   
 (iv)  $x_{ij} \in \mathbb{Z}_+ \quad \forall j \in A$

- a) (UFL) has an optimal 0/1-solution.
- b) (UFL)  $\Leftrightarrow$  (UFL)(i),(ii),(iii),  $x_{ij} \geq 0 \quad \forall j \in A$ , i.e., the client assignment variables are automatically integral.

relax Comp  $\leftarrow$  Proof: ex. sheet 11.  $\square$

3.1.3. Prop. (Set Covering Model for the UFL):

$J(i) := \{j \in J : j \in A\}$  clients covered by fac.  $i$   
 $\mathcal{C} := \{C_i, J^i\} : i \in I, \emptyset \neq J^i \subseteq J(i)$

$$c_{(i,j)} := f_i + \sum_{j \in J} d_{ij}$$

$$(SCP) \min \sum_{(i,j) \in X} c_{(i,j)} z_{(i,j)}$$

$$\sum_{j \in J} z_{(i,j)} \geq 1 \quad \forall i \in I$$

$$z_{(i,j)} \in \{0,1\} \quad \forall (i,j) \in X$$

There exists a one-to-one correspondence between optimal solutions of (SCP) and (UFL).

Proof: Ex. sheet 11.  $\square$

Cor. 3.13: a) The greedy algorithm for UFL is  $H(\max\{f_i\})$ -approximate.

b) UFL is APX-hard.

Proof: a) follows from Thm. 2.6.8

b) SCP is APX-hard. Given all SCP with  $\max_{(i,j) \in X} c_{(i,j)} \geq 1$ ,  $c_{(i,j)} \in \{0,1\}^n$ , construct an UFL <sup>(in polynomial time)</sup>

$$I = \{1, \dots, n\}, J = \{1, \dots, n\}, A = \{(i,j) : c_{ij} = 1\}, f_i = c_{ij}$$

then (UFL)  $\Leftrightarrow$  (SCP) and the equivalence is approximation preserving.  $\square$

Obs 3.14 (Relation to the p-median problem):

$$a) (UFL) \Leftrightarrow |I|/|I| \cdot |d_{ij}|/Z$$

$$b) p/|V| \cdot |d_{ij}|/Z \Leftrightarrow (UFL), (\forall) \sum_{i \in I} y_i \leq p$$

3.2. The Metric Uncapacitated Facility Location Problem

3.2.1 Def. (Metric UFL): A UFL is

i)  $I, J \subseteq \mathbb{R}^k$ ,  $I, J$  are embedded in  $\mathbb{R}^k$

ii)  $X = I \times J$ , all requests are possible

iii)  $d_{ij} = \|i, j\|$ , distances with norm

is a metric UFL (MUFL).

3.2.3. Alg. (LP bounding):

Input: (MUFL)  $I, J \subseteq \mathbb{R}^k$ ,  $d_{ij} = \|i, j\| \forall i, j \in I, J$ ,  $f_i \in \mathbb{R}_{\geq 0} \forall i \in I$

Output:  $(x, y)$  feasible for (MUFL)

D. LP Solving:  $(x, y) = \text{argmin (MUFL)}_{LP}$ , where

$$(MUFL)_{LP} \min \sum_{i \in I} f_i y_i + \sum_{(i,j) \in X} d_{ij} x_{ij}$$

$$\sum_{i \in I} x_{ij} \geq 1 \quad \forall j \in J$$

$$x_{ij} \geq 0 \quad \forall i \in I, j \in J$$

$$y_j \geq 0 \quad \forall j \in J$$

1. Relaxing

$$D_j \leftarrow \sum_{i \in I} d_{ij} y_j \quad \text{"total fractional distance"}$$

$$N_j \leftarrow \{i \in I : x_{ij} > 0\} \quad \left. \begin{array}{l} x_{ij} > 0 \wedge d_{ij} \leq 2D_j \\ \text{"fractionally assigned"} \end{array} \right\} \text{"near frac. assigned fac"}$$

2.0th Case:  $\sum_{i \in I} x_{ij} = 1 \quad \forall j \in J$

$$x_{ij} \in \{0, 1\} \quad \forall i \in I, j \in J$$

$$y_j \leftarrow \max_{i \in I} x_{ij}$$

3.0th Obs:  $(x^*, y^*)$  is feasible for (MFL) if and

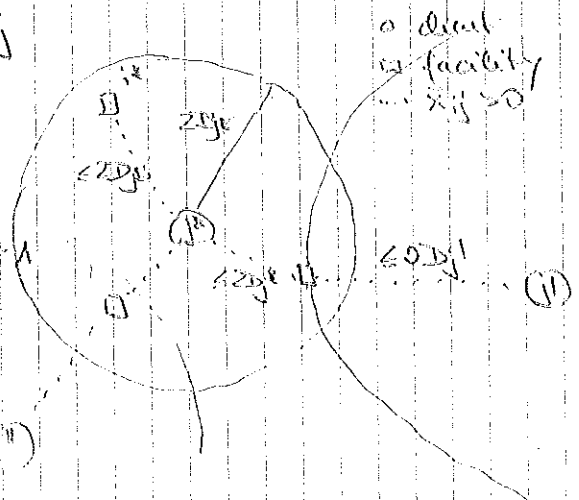
- i)  $x_{ij} > 0 \Rightarrow d_{ij} \leq 2D_j$
- ii)  $(x^*, y^*) \in Z(x^*, y^*)$

2. Rounding:

2a.  $J^k = \{j \in J : (x^*, y^*) \in Z(x^*, y^*)\}$

2b.  $J^k \leftarrow \text{argmin}_{j \in J^k} D_j$

$i^k \leftarrow \text{argmin}_{i \in N_{j^k}} d_{ij^k}$



$EW_{j^k} \leftarrow \{i \in I : i \in N_{j^k}\}$  "extended neighborhood of client  $j^k$ "

$$y_{j^k} \leftarrow \begin{cases} 1 & i = i^k \\ 0 & i \in N_{j^k} \setminus \{i^k\} \\ y_{j^k} & \text{else} \end{cases} \quad \text{select } i^k, \text{ drop } N_{j^k} \setminus \{i^k\}$$

$$x_{ij^k} \leftarrow \begin{cases} 1 & i = i^k, j^k \in EW_{j^k} \\ 0 & i \in N_{j^k} \setminus \{i^k\}, j^k \in EW_{j^k} \\ 0 & i \in N_{j^k} \setminus \{i^k\}, j^k \notin EW_{j^k} \\ x_{ij} & \text{else} \end{cases} \quad \begin{array}{l} \text{assign all unass. clients in} \\ EW_{j^k} \text{ to } i^k \end{array}$$

$J^k \leftarrow J^k \cup EW_{j^k}$

2c.  $(x^{n+1}, y^{n+1}) \leftarrow (x^n, y^n)$ , output  $(x^n, y^n)$ .

3.2.6 Claim:

i)  $j \in E \cup J^* \Rightarrow |d_j| \leq 2 \cdot 2D_j + 2D_j \leq 6D_j$

ii)  $f_i^* = \sum_{j \in E} x_j^* y_j^* f_j$

iii)  $\sum_{i \in I} f_i x_i^{n+1} + \sum_{j \in J} d_j x_j^{n+1} \leq 6 \sum_{i \in I} f_i y_i^n + \sum_{j \in J} d_j x_j^n$

$\sum_{i \in I} f_i y_i^n \leq 2 \sum_{i \in I} f_i y_i^n$

$\sum_{j \in J} d_j x_j^n \leq 6 \sum_{j \in J} x_j^n d_j$

3.2.5(i)

(Always Terminate Algorithm (1997))

3.2.7 Theorem:  $\forall \epsilon > 0$  an approximate.

3.2.8 Lemma (generalization of Claim 3.2.4): for  $\beta > 0$  let

$N_J(\beta) := \{i \in I : x_{ij} > 0 \wedge d_j \leq \beta D_j\}$ .

then  $\sum_{i \in N_J(\beta)} x_{ij} \leq 1 - \frac{\beta}{D_j} \leq \frac{\beta-1}{\beta}$ .

Proof: let

$F_J(\beta) := \{i \in I : x_{ij} > 0 \wedge d_j > \beta D_j\}$

$D_j := \sum_{i \in I} x_{ij} \leq 1$

Suppose

$D_j = \sum_{i \in N_J(\beta)} x_{ij} + \sum_{i \in F_J(\beta)} x_{ij} \leq \sum_{i \in N_J(\beta)} d_j x_{ij} + \sum_{i \in F_J(\beta)} d_j x_{ij}$

$\leq \sum_{i \in N_J(\beta)} \beta D_j x_{ij} + \sum_{i \in F_J(\beta)} D_j x_{ij}$

$\leq \beta D_j \sum_{i \in N_J(\beta)} x_{ij} + D_j \sum_{i \in F_J(\beta)} x_{ij}$

$\leq \beta D_j \sum_{i \in N_J(\beta)} x_{ij} + D_j (1 - \sum_{i \in N_J(\beta)} x_{ij})$

$\leq D_j \leq \beta D_j \sum_{i \in N_J(\beta)} x_{ij} + D_j (1 - \sum_{i \in N_J(\beta)} x_{ij})$

$\Rightarrow \sum_{i \in N_J(\beta)} x_{ij} \leq \frac{\beta-1}{\beta}$

$\sum_{i \in N_J(\beta)} x_{ij} \leq \frac{\beta-1}{\beta} \Rightarrow \sum_{i \in N_J(\beta)} x_{ij} \leq 1 - \frac{1}{\beta} \Rightarrow 1 - \frac{1}{\beta} \leq 1 - \frac{1}{\beta}$

25.06.12  
(17)

# Metric Uncapacitated Facility Location

$I$ : set of facilities

$J$ : set of clients

$f_i$ : opening cost of facility  $i$

$d_{ij}$ : cost of assigning client  $j$  to facility  $i$

Goal: Open a subset of facilities and assign each client to an open facility while incurring in a minimum cost

## Primal-Dual Approximation Algorithm (Jain & Vazirani, 2001)

$$(LP): \min \sum_{i \in I} f_i y_i + \sum_{i \in I} \sum_{j \in J} d_{ij} x_{ij}$$

(1)

$$\text{s.t. } \sum_{i \in I} x_{ij} \geq 1 \quad \forall j \in J \rightarrow \alpha_j$$
$$x_{ij} \leq y_i \quad \forall i \in I, j \in J \rightarrow \beta_{ij}$$
$$x_{ij}, y_i \geq 0 \quad \forall i \in I, j \in J$$

$$(D): \max \sum_{j \in J} \alpha_j$$
$$\text{s.t. } \sum_{j \in J} \beta_{ij} \leq f_i \quad \forall i \in I \rightarrow y_i$$
$$\alpha_j - \beta_{ij} \leq d_{ij} \quad \forall i \in I, j \in J \rightarrow x_{ij}$$
$$\alpha_j, \beta_{ij} \geq 0 \quad \forall i \in I, j \in J$$

(1)  
Interpretation of dual:  $\alpha_j$ : Total cost paid by  $j$  to get assigned to an open facility.  
 $\beta_{ij}$ : Cost paid by  $j$  to open facility  $i$ .

Why? Suppose  $(x, y)$  is integral and optimal for (LP). Let  $(\alpha, \beta)$  be a corresponding dual solution.

By complementary slackness:

•  $y_i > 0 \Rightarrow \sum_{j \in J} \beta_{ij} = f_i$  (To open  $i$ ,  $f_i$  needs to be paid for by all clients)

•  $x_{ij} > 0 \Rightarrow \alpha_j = d_{ij} + \beta_{ij}$  (If  $j$  is connected to  $i$ ,  $j$  pays the assignment cost and its share for opening  $i$ )

•  $\alpha_j > 0 \Rightarrow \sum_{i \in I} x_{ij} = 1$  (Abt interesting)

•  $\beta_{ij} > 0 \Rightarrow x_{ij} = y_i$  (Client  $j$  only pays its share of  $f_i$  if it uses facility  $i$ )

### Algorithm: Phase 1

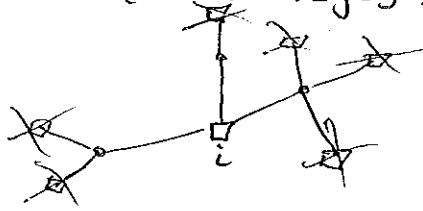
$x, y, \alpha, \beta := 0, T := \{i \in I \mid \sum_{j \in J} \beta_{ij} = f_i\}$  (temporarily open facilities)

Definitions:  
 $j$  neighbors  $i$   
 if  $\alpha_j \geq d_{ij}$   
 $j$  contributes to  $i$   
 if  $\beta_{ij} > 0$

- while  $\exists$  a client  $j$  not neighboring any facility in  $T$ .  
 Raise  $\alpha_j$  uniformly for all such  $j$ .
- if  $j$  becomes neighbor of some  $i \in T \Rightarrow$  Increase  $\beta_{ij}$  at the same rate to maintain feasibility.
- if  $j$  becomes neighbor of some  $i \in T \Rightarrow$  Freeze variable  $\alpha_j$ .

### Phase 2 (Cleanup)

- while  $\exists$  a client contributing to at least two facilities in  $T$ :  
 - Pick an arbitrary such  $i$   
 -  $T := T - \{h \in T \mid \exists j \in J \text{ s.t. } \beta_{ij} > 0 \text{ and } \beta_{hj} > 0\}$



### Phase 3

Set  $y_i = 1 \quad \forall i \in T$

For all  $j \in J$ :

- Case 1:  $j$  contributes to a unique  $i \in T \Rightarrow x_{ij} = 1$
- Case 2:  $j$  contributes to no  $i \in T$  but neighbors some  $i \in T$   
 $\Rightarrow x_{ij} := 1$  for arbitrary such  $i$ .
- Case 3:  $j$  has no neighbors in  $T \Rightarrow$  Let  $i$  be closest facility in  $T, x_{ij} = 1$ .

### Analysis:

Lemma: If at the end of Phase 2 there is a client  $j$  with no neighbors in  $T$ , there exists  $i \in T$  s.t.  $d_{ij} \leq 3\alpha_j$ .

Proof: Later

Theorem: The algorithm gives a 3-approximation.

Proof: For every  $i$ , let  $A(i) = \{\text{neighbours of } i \text{ s.t. } x_{ij} = 1\}$

$$\sum_{i \in T} (f_i + \sum_{j \in A(i)} d_{ij}) = \sum_{i \in T} \sum_{j \in A(i)} (\beta_{ij} + d_{ij}) = \sum_{i \in T} \sum_{j \in A(i)} \alpha_j$$

Let  $Z := J \setminus \bigcup_{i \in T} A(i)$  (clients with no neighbors in  $T$ )

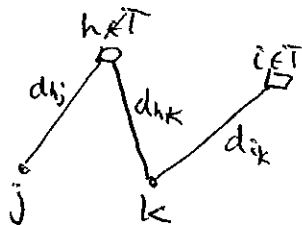
$$\sum_{i \in T} \sum_{j \in Z} d_{ij} x_{ij} = \sum_{j \in Z} \sum_{i \in T} d_{ij} x_{ij} \leq \sum_{j \in Z} 3\alpha_j$$

Lemma

$$\Rightarrow \sum_{i \in T} (f_i + \sum_{j \in J} d_{ij} x_{ij}) = \sum_{i \in T} \sum_{j \in A(i)} \alpha_j + \sum_{j \in Z} 3\alpha_j \leq 3 \sum_{j \in J} \alpha_j \leq 3 \text{OPT}_D \leq 3 \text{OPT}_{LP} \quad \square$$

Proof of Lemma:

Claim 1: Given such  $j$ , there exists  $h, k$ , s.t.:



Why?

- We stopped increasing  $\alpha_j$  when  $j$  neighbored some  $h \in T$ .
- $h$  was removed from  $T$  in Phase 2 because  $k$  was contributing both to it and to  $i$  (which remains in  $T$ ).

Claim 2:  $d_{ij} \leq 3\alpha_j$

Pf: By triangle inequality.

1.  $d_{hj} \leq \alpha_j$  because  $h$  and  $j$  are neighbours.
2.  $k$  contributes to  $h \Rightarrow \beta_{hk} > 0 \Rightarrow \alpha_k > d_{hk}$ .  
When  $\alpha_j$  stops growing,  $h$  is in  $T$ .  $k$  contributed to  $h \Rightarrow$  At that time,  $k$  already neighbors  $h$   
 $\Rightarrow \alpha_k$  doesn't keep growing after  $\alpha_j$  stops  $\Rightarrow \alpha_j \geq \alpha_k \Rightarrow \alpha_j \geq d_{hk}$ .
3.  $\beta_{ik} > 0 \Rightarrow \alpha_k \geq d_{ik} \Rightarrow \alpha_j \geq d_{ik}$ .

So,  $d_{ij} \leq d_{hj} + d_{hk} + d_{ik} \leq 3\alpha_j \quad \square$

Best approximation known: 1.52-approx. by Mahdian, Ye, Zhang.

Hardness: No approximation better than 1.46 possible (unless  $P=NP$ )

References: • Jain, Vazirani: Approx. algorithms for metric facility location and  $k$ -median problems (2001)

• Shmoys, Williamson: The Design of Approximation Algorithms (2010)

• Vazirani: Approximation Algorithms (2001)

3.2.9 Alg. (Modified Greedy Algorithm for the metric UFL):

Input:  $I, J \subseteq \mathbb{R}^k$ ,  $d_{ij} = \|i - j\| \forall i, j \in I \cup J$ ,  $f_i \in \mathbb{R} \geq 0 \forall i \in I$

Output:  $I^* \subseteq I$  opened facilities

1.  $I^* \leftarrow \emptyset$
2. While  $J \neq \emptyset$  do
  - 2a.  $(i \in I, A) \leftarrow \text{argmin} \frac{f_i + \sum_{j \in A} d_{ij}}{|A|}$
  - 2b.  $I^* \leftarrow I^* \cup \{i\}$
  - 2c.  $f_i \leftarrow 0$  (close open facility for free)
  - 2d.  $J \leftarrow J \setminus A$
3. Output  $I^*$

3.2.10 Remark: The difference between the standard and the modified greedy is step 2c, which allows an open facility to be closed at 0 cost.

3.2.11 Remark (Primal and Dual LP for UFL):

$$\begin{aligned}
 \text{(P)} \quad \min \quad & \sum_{i \in I} f_i x_i - \sum_{j \in J} d_{ij} x_j \\
 \text{s.t.} \quad & \sum_{i \in I} x_i \geq 1 \\
 & x_i - x_j \geq 0 \\
 & x_i \geq 0 \\
 & x_j \geq 0
 \end{aligned}
 \quad \stackrel{(\text{D})}{=} \quad
 \begin{aligned}
 \max \quad & \sum_{j \in J} \alpha_j \\
 \text{s.t.} \quad & \sum_{i \in I} \beta_{ij} \leq f_i \\
 & \alpha_j - \beta_{ij} \leq d_{ij} \\
 & \alpha_j \geq 0 \\
 & \beta_{ij} \geq 0
 \end{aligned}$$

$\alpha_j \triangleq$  transport contribution of client  $j$

$\beta_{ij} \triangleq$  contribution of client  $j$  to facility  $i$

3.2.12 Alg. (Primal-Dual Alg. for the metric UFL):

Input:  $I, J \subseteq \mathbb{R}^k$ ,  $d_{ij} = \|i - j\| \forall i, j \in I \cup J$ ,  $f_i \in \mathbb{R} \geq 0 \forall i \in I$

Output:  $I^* \subseteq I$ ,  $A_i \subseteq J$ ,  $i \in I \cup \{0\}$

Data Structures:

- $I^* \subseteq I$ : opened facilities
- $J' \subseteq J$ : unassigned clients (with active duals)
- $(\text{P}, \text{D}) \geq 0$ : infeasible solution for (P)
- $A_i \subseteq J$ : clients assigned to facility  $i \in I$  when  $i$  is opened
- $A_0$ : " " " opened facility



1.  $I^* \neq \emptyset, J \neq \emptyset, d = 0, k = 0, A_i = \emptyset \forall i \in I \cup J$

2. while  $J \neq \emptyset$  do

uniformly raise

$$x_j \quad \forall j \in J$$

$$p_j \quad \forall j \in J, i \in I \setminus I^*; \quad x_j \geq d_{ij}$$

until

a)  $x_j = d_{ij} \quad \forall j \in J, i \in I^*$ ; then

$$J' \leftarrow J \setminus \{j\} \quad (\text{assign } j \text{ to opened facility } i)$$

$$\beta_{ij} \leftarrow 0 \quad \forall i \in I$$

b)  $\sum_{j \in J} \beta_{ij} = f_i \quad \forall i \in I^*$ ; then

$$A_i \leftarrow \{j \in J' : \beta_{ij} > 0\}$$

$\beta_{ij} = 0 \quad \forall i \in I \setminus \{i\}, j \in A_i$  (open facility  $i$  and assign clients in  $A$  to  $i$ )

$$J' \leftarrow J' \setminus A_i$$

3.2.13 Prop: Alg. 32.12 computes  $I^*, A_i, i \in I \cup J, d, k$  such that

$$f(I^*) + \sum_{i \in I^*} d(i, A_i) = d(J)$$

i.e., the objective values of the primal version  $(P)$  and the dual (infeasible) version  $(D)$  coincide.

Proof:

(1)  $|\{i \in I : \beta_{ij} > 0\}| \leq 1 \quad \forall j \in J, \text{ i.e. } J = \bigcup A_i$

Subst:  $|\cdot| = \begin{cases} 0 & \text{if } j \text{ is assigned to } i \text{ in } \{2a\} \\ 1 & \text{if } j \text{ is assigned to } i \text{ in } \{2b\} \end{cases}$

(2)  $f_i = \sum_{j \in A_i} \beta_{ij} = \sum_{j \in A_i} (x_j - d_{ij}) \quad \forall i \in I^*$

Subst: By construction.

(3)  $x_j = d_{ij} \quad \forall j \in A_0$

Subst: By construction.

Thus

$$\begin{aligned} \sum_{i \in I^*} f_i + \sum_{j \in J} d(I^*, j) &= \sum_{i \in I^*} \sum_{j \in A_i} (x_j - d_{ij}) + \sum_{j \in J} d(I^*, j) \\ &= \sum_{i \in I^*} \sum_{j \in A_i} x_j + \sum_{j \in A_0} d(I^*, j) \end{aligned}$$

$$\sum_{j \in A_i} \alpha_j + \sum_{j \in A_0} \alpha_j$$

$$\sum_{j \in I} \alpha_j \quad \square$$

Prop. 3.2.14:  $(\alpha, \beta)$  is feasible for (B).

Cor. 3.2.15: The modified greedy is a 3-approx. alg. for discrete weighted util.

Proof of Prop. 3.2.14:

(1)  $\alpha_j \leq d_{ij} + 2\alpha_{j'}$   $\forall i \in I, j, j' \in J: \alpha_j \geq d_{ij}, \alpha_{j'} \geq d_{ij'}$ .

If  $\alpha_{j'} \geq \alpha_j$  ✓. So let  $\alpha_{j'} < \alpha_j$ .

$\alpha_j \geq d_{ij}$  since at which  $j$  is assigned to some facility (ii).

$\alpha_{j'} < \alpha_j \Rightarrow j' \rightarrow i$  (iii)  $\Rightarrow$  later  $\Rightarrow \alpha_j \leq d_{ij}$

$$d_{ij} \leq d_{ij'} + \alpha_{j'} + d_{ij} \leq 2\alpha_{j'} + d_{ij}$$

$$\Rightarrow \alpha_j \leq d_{ij} \leq 2\alpha_{j'} + d_{ij}$$

(2)  $f_i \geq \sum_{j \in J} \max\{0, \frac{\alpha_j}{3} - d_{ij}\} \quad \forall i \in I$

$C_i = \{j \in J : \frac{\alpha_j}{3} \geq d_{ij}\} = \{1, \dots, k\} \subseteq I, \alpha_1 \leq \dots \leq \alpha_k$

$\alpha_j \geq 3d_{ij}$ , clients don't pay a lot for  $f_i$

$$\Rightarrow \sum_{j \in J} \max\{0, \frac{\alpha_j}{3} - d_{ij}\} = \sum_{j \in C_i} (\frac{\alpha_j}{3} - d_{ij}) = \sum_{j=1}^k (\frac{\alpha_j}{3} - d_{ij})$$

Claim:  $\sum_{i=1}^p \max\{0, \alpha_i - d_{ij}\} \leq f_i$

Subst.: At time  $\alpha_1, f_j(\alpha_1) = \max\{0, \alpha_1 - d_{ij}\}$

$\alpha_j/3 \geq d_{ij} \Rightarrow \alpha_j \geq d_{ij} \quad \forall j \in C_i$

$$\sum_{j \in J} \alpha_j \leq \sum_{j=1}^k (d_{ij} + 2\alpha_1) = 2\sum_{j=1}^k (\alpha_1 - d_{ij}) + 3\sum_{j=1}^k d_{ij}$$

$$\Rightarrow f_i \geq \sum_{j=1}^k f_i \geq \sum_{j=1}^k (\frac{\alpha_j}{3} - d_{ij}) = \sum_{j \in J} \max\{0, \frac{\alpha_j}{3} - d_{ij}\}$$

(3)  $(\alpha, \beta) = (\sum_{i=1}^p \max\{\frac{\alpha_i}{3} - d_{ij}, 0\})$  is feasible for (A)

$\sum_{j \in J} \beta_j = \sum_{j \in J} \max\{\frac{\alpha_j}{3} - d_{ij}, 0\} \stackrel{(2)}{=} f_i$

$$x_j - f_{ij} = \sum_{i \in I} \max\left\{\frac{f_i}{|S|} - d_{ij}, 0\right\} \leq d_{ij}$$

$\forall j, \delta \geq 0, \square$

3.2.16 Prop. Alg. 3.2.9 and 3.2.12 produce the same solution.

Proof: Consider some point in time  $t$  in Alg. 3.2.12. Then

$$x_j = t \quad \forall j \in J'$$

If 2a) doesn't hold  $x_j = t \leq d_{ij} \quad \forall i \in I^*, j \in J'$ .

If 2b) "  $\sum_{j \in J'} \max\{0, t - d_{ij}\} \leq f_i \quad \forall i \in I \setminus I^*$

$$\sum_{j \in S} (t - d_{ij}) = f_i \Leftrightarrow t = \frac{f_i + \sum_{j \in S} d_{ij}}{|S|}$$

The next decision in Alg. 3.2.2 is at time

$$t' = \min \left\{ \min_{i \in I^*, j \in J'} d_{ij}, \min_{i \in I \setminus I^*, S \subseteq J'} \frac{f_i + \sum_{j \in S} d_{ij}}{|S|} \right\}$$

$$= \min_{i \in I^*, S \subseteq J'} \frac{0 + \sum_{j \in S} d_{ij}}{|S|}$$

This is exactly the criterion used by the greedy algorithm 3.2.9.  $\square$

Thm. 3.2.17 The modified greedy algorithm 3.2.9 for the metric

UFL is 3-approximative.

05.07.12

### 3.3. Local Search

3.3.1 Def. (Neighborhood,  $\Gamma$ ) for the  $k$ -median problem: given

$V = \{1, \dots, n\}^k$ , consider the  $k$ -median problem  $p/\mathbb{Z}/\|\cdot\|/2$ :

- a)  $\mathcal{S} = \{S \in \mathbb{Z} : |S| = k\}$  set of  $k$ -subsets (feasible solutions)
- b)  $c(S) := \sum_{i \in \mathbb{Z}} \|s_i\| = \sum_{i \in \mathbb{Z}} \min_{j \in S} \|i, j\|$  cost of a solution
- c)  $\Gamma(S) := \{S' \in \mathbb{Z} : |S' \Delta S| = 2\}$  neighborhood of a solution
- d)  $i \rightarrow i'$ :  $\{S \in \mathbb{Z} : i \in S, i' \notin S, S \setminus \{i\} \cup \{i'\} \}$  swap
- e)  $S \in \mathbb{Z}$  local optimum:  $\forall S' \in \Gamma(S) : c(S) \leq c(S')$

3.3.2 Alg. (Local Search for the  $k$ -median problem):

- Input:  $p/\mathbb{Z}/\|\cdot\|/2, k, \mathbb{Z}, S \in \mathbb{Z}$
- Output:  $\bar{S} \in \mathbb{Z}$  local optimum
- 1. while  $\exists S' \in \Gamma(S) : c(S') < c(S)$   
 $S \leftarrow S'$
- 2. output  $S$

3.3.2 Prop.  $p/\mathbb{Z}/\|\cdot\|/2 \in (UMC) \mid \sum_{i \in \mathbb{Z}} \|i\| = F, f \in \text{conv.}$

3.3.3 Lem. Let  $\bar{S}$  be a local,  $S^*$  a global optimum. We want to know:

- a) How large can  $c(\bar{S})/c(S^*)$  be?
- b) How fast does Alg. 3.3.2 converge?

Any guy, Munksgaard, Munksgaard & Freund [FOCS] showed

- a)  $c(\bar{S}) \leq 2c(S^*)$
- b)  $\forall S \in \mathbb{Z} : \exists S' \in \Gamma(S) : c(S) - c(S') \geq \frac{c(S) - c(S^*)}{4}$   
decrease per swap

and this yields an  $1 + O(1/n)$ -approximate polynomial time local search algorithm.

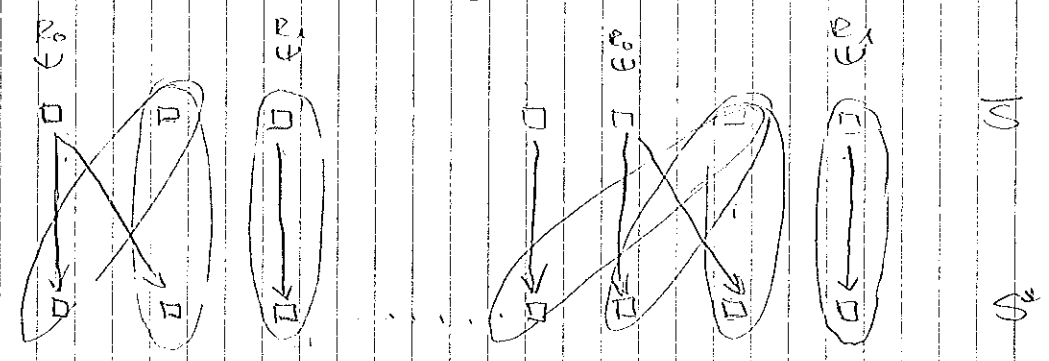
3.3.5 Thm (5-approximation ratio for local search)  $\forall$  Alg. 3.3.3

yields a local optimum  $\bar{S}$  such that  $c(\bar{S}) \leq 5c(S^*)$ .

Proof: Let  $\bar{S}$  be the output of Alg. 3.3.3,  $S^*$  an optimal solution. Define a map

$$\eta: S^* \rightarrow \bar{S}, i^* \mapsto \arg \min_{i \in \bar{S}} \|i^*, i\|$$

which maps each facility  $i^* \in S^*$  to the closest facility in  $\bar{S}$ , b.l.a. let  $\textcircled{2}$



$R_k := \{i \in \bar{S} : \eta(i^*) = i \text{ for exactly } k \text{ facilities } i^* \in S^k\}, k=0,1.$

Construct a set  $P$  of  $k$  swaps, one for each  $i^* \in S^k$ , as follows:

a)  $i \in R_1 \Rightarrow i \rightarrow \eta^{-1}(i) \in P$

b)  $|R_0| + |\bar{S} \setminus (R_0 \cup R_1)| = |S^k \setminus \eta^{-1}(R_1)| \geq \frac{1}{2} |S^k \setminus \eta^{-1}(R_1)|$   
 $\Rightarrow 2|R_0| \geq |S^k \setminus \eta^{-1}(R_1)|$

$i \in R_1 \Rightarrow i \rightarrow i^* \in P$  for at most 2 arbitrarily chosen  $i \in S^k \setminus \eta^{-1}(R_1)$

Rem: a)  $i \in R_1$  is close to  $\eta^{-1}(i)$ , all other  $i^* \in S^k$  are far away

$\Rightarrow i \rightarrow \eta^{-1}(i)$  can be handled by assigning all of its clients to  $\eta^{-1}(i)$

b)  $i \in \bar{S} \setminus (R_0 \cup R_1)$  is close to several facilities  $\eta^{-1}(i)$ .

$\Rightarrow i \rightarrow i^* \in \eta^{-1}(i)$  and assigning all of its clients to  $i^*$  can

cost much, so we consider only swaps  $i \rightarrow i^* \in P$ .

Claim 1: For  $i \in \bar{S}, i^*, i^* \in S^k$ , and  $i \rightarrow \begin{matrix} R_0 \\ R_1 \end{matrix} \in P$  it holds  $\eta(i^*) \neq i$ .

Let

$\bar{\varphi}: J \rightarrow \bar{S}, j \mapsto \arg \min_{i \in \bar{S}} \|S, j\|$

$\varphi^*: J \rightarrow S^k, j \mapsto \arg \min_{i^* \in S^k} \|S^*, j\|$

function mapping  $j$  to the closest facility in  $\bar{S}$  and  $S^k$ , respectively.

$N(i) = \bar{\varphi}^{-1}(i), i \in \bar{S}$

$N^*(i) = \varphi^{*-1}(i), i \in S^k$

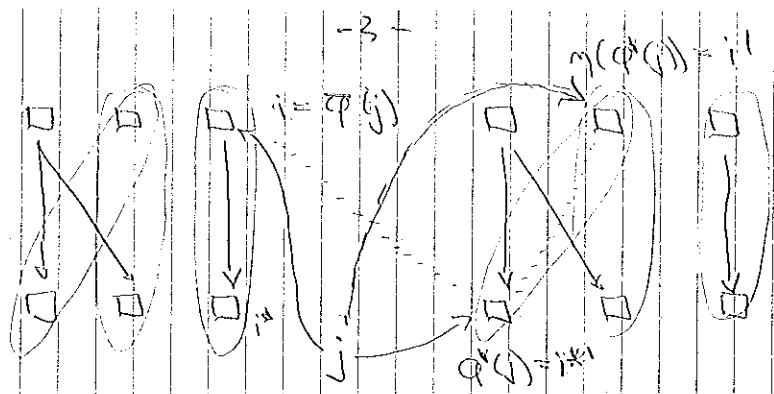
be the sets of clients assigned to  $i \in \bar{S}$  and  $i \in S^k$ , respectively.

Claim 2: For each swap  $i \rightarrow i^* \in P$  it holds

0  $\leq c(\bar{S} \setminus \{i\} \cup \{i^*\}) - c(\bar{S})$

$\leq \sum_{j \in N^*(i^*)} (\|S^*, j\| - \|\bar{S}, j\|) + \sum_{j \in N(i)} 2 \|S^*, j\|$

Subst: (1) is because  $\bar{S}$  is a local optimum.



Consider the following assignment  $\phi^1: J \rightarrow S \setminus \{i\} \cup \{i^*\}$

$$\phi^1(j) = \begin{cases} i^* & , j = N^*(i^*) \\ i' = \gamma(\phi^*(j)) & , j \in N(i) \setminus N^*(i^*) \\ \phi(j) & , j \notin N^*(i^*) \cup N(i) \end{cases}$$

Because of claim 1,  $i \neq i'$ , so  $\phi^1$  is well defined. Therefore

$$c(S \setminus \{i\} \cup \{i^*\}) - c(S) = \sum_{j \in J} (\|S \setminus \{i\} \cup \{i^*\}, j\| - \|S, j\|)$$

$$= \sum_{j \in N^*(i^*)} (\|i^*, j\| - \|S, j\|) + \sum_{j \in N(i) \setminus N^*(i^*)} (\|i', j\| - \|i, j\|)$$

$$\leq \|i^*, j\| + \|i^*, i'\| - \|i, j\|$$

$$\leq \|i^*, i\|$$

$$\leq \|i^*, j\|$$

$$\leq 2 \|i^*, j\|$$

$$= 2 \|S^*, j\|$$

$$= \sum_{j \in N^*(i^*)} (\|S^*, j\| - \|S, j\|) + \sum_{j \in N(i) \setminus N^*(i^*)} 2 \|S^*, j\|$$

$$\leq \|S^*\| + \sum_{j \in N(i)} 2 \|S^*, j\|$$

Summing over all  $i \rightarrow i^* \in P$  gives

$$0 \leq \sum_{i^* \in S^*} \sum_{j \in N^*(i^*)} (\|S^*, j\| - \|S, j\|) + \sum_{i \in S} \sum_{j \in N(i)} 2 \|S^*, j\|$$

↑  
appears at most 2 times

$$= 5 c(S^*) - c(S)$$

$$\rightarrow c(S) \leq 5 c(S^*)$$