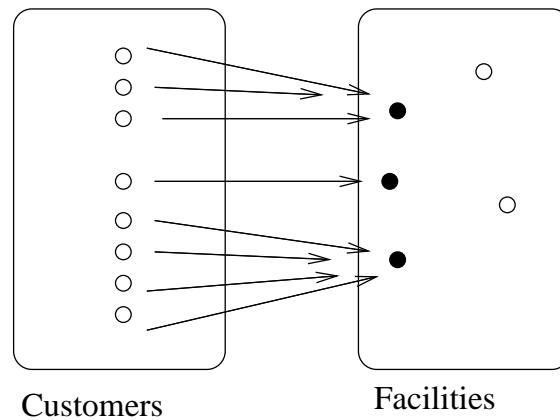


Oblivious medians via online bidding

M. Chrobak, Claire Kenyon, J. Noga and N. Young

k-median

- Choose $\leq k$ facilities: F_k
- $\text{cost}(F_k) = \sum_c \text{customer } d(c, \text{closest facility})$
- Competitive ratio $\text{cost}(F_k)/\text{OPT}_k$.
- Best algorithm: $3 + \epsilon$ [Arya Garg Khandekar Meyerson Munagala Pandit 01 04]

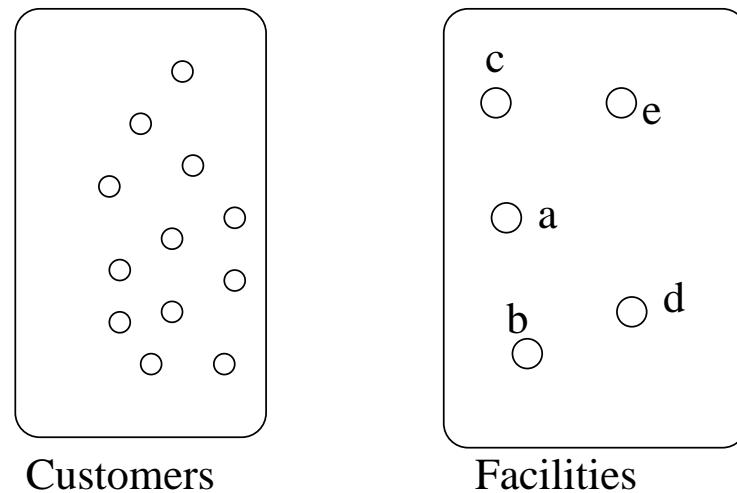


Oblivious median [Mettu Plaxton 99]

- **Algorithm:** Choose $F_1 \subseteq F_2 \subseteq \dots \subseteq F_n$
- **Adversary:** Chooses k
- **Goal:** Minimize

$$\max_k \text{cost}(F_k)/\text{OPT}_k.$$

$$F_1 = \{a\}, F_2 = \{a, b\}, F_3 = \{a, b, c\}, F_4 = \{a, b, c, d\}, F_5 = \{a, b, c, d, e\}$$



Greedy algorithm

$F(0) \leftarrow \text{emptyset}$

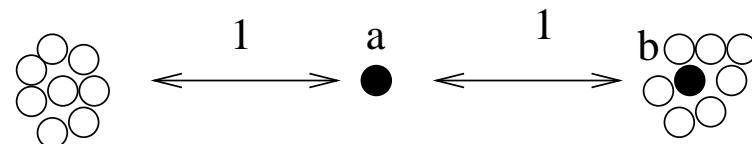
For $i \leftarrow 1$ to n

$F(i) \leftarrow F(i-1) \cup \{f\}$, where

f = greedy choice of next facility to add

Greedy fails

- $F_1 = \{a\}$, $f_2 = \{a, b\}$
- $\text{cost}(F_2) = M$
- $\text{OPT}_2 = 1$
- Ratio: M , unbounded



Mettu-Plaxton's algorithm

```
F(0) <-- emptyset
```

```
For i <-- 1 to n
```

```
    F(i) <-- F(i-1) union {f}, where
```

```
        f = careful choice of next facility to add
```

- Technical algorithm, plus
- Messy analysis, equals
- Competitive ratio around 30

Reverse Greedy [Fiat]

$F(n) \leftarrow F$

For $i \leftarrow n-1$ down to 1

$F(i) \leftarrow F(i+1)$ minus $\{f\}$, where

f = greedy choice of facility to remove

Reverse greedy fails [Chrobak K. Young]

Competitive ratio: $\Omega(\log n / \log \log n)$

Batch reverse greedy

[Chrobak K. Noga Young, Lin Nagarajan Rajaraman Williamson]

Choose a subset K of $\{1, 2, \dots, n\}$

$F(n) \leftarrow F$

For k in K , in decreasing order

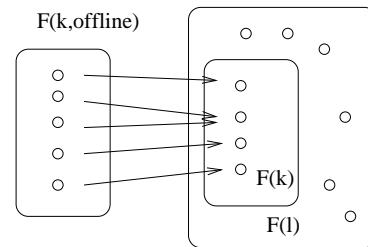
 Assume $F(l)$ is already known, $l > k$, l in K

$F(k) \leftarrow$ a subset of $F(l)$ of size at most k

$F(k+1), F(k+2), \dots, F(l-1) \leftarrow F(k)$

How to choose the subset:

- Let F_k^{off} be a solution to (offline) k -median.
- For $x \in F_k^{\text{off}}$, let $f(x)$ be the point of F_ℓ closest to x .
- Then $F_k = \cup_x \{f(x)\}$



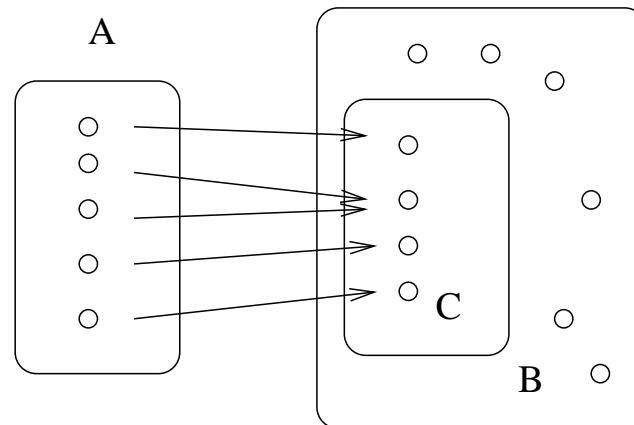
Results

- $(F_k^{\text{off}} = \text{OPT}_k)$ Competitive ratio of 8 (deterministic) or $2e$ (randomized)
- $(F_k^{\text{off}} = (3 + \epsilon)\text{-approximation})$ Competitive ratio of $24 + 8\epsilon$ (deterministic) or $6e + 2e\epsilon$ (randomized)
- **Further improvements [LNRW]:** Competitive ratio of $16 + O(\epsilon)$ (deterministic) or $4e + O(\epsilon)$ (randomized)

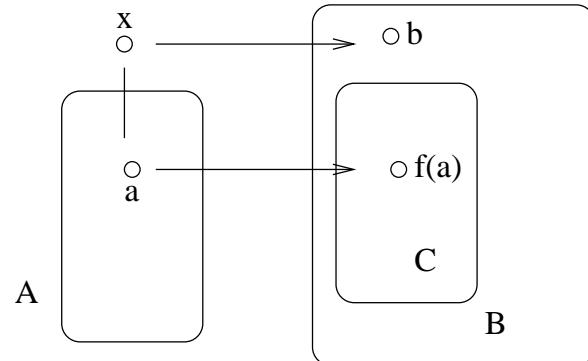
Analyzing one step

Lemma [implicit in Jain-Vazirani]: Let A, B be two sets. Let C be a subset of elements of B containing, for each $a \in A$, an element of B closest to a . Then, for every x ,

$$\text{dist}(x, C) \leq 2\text{dist}(x, A) + \text{dist}(x, B).$$



Proof



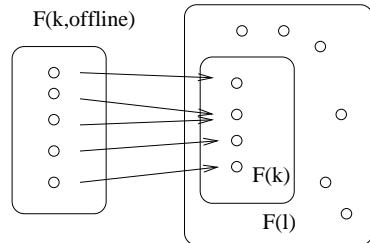
- **a:** point in A closest to x .
- **f(a):** point in B closest to a .
- **b:** point in B closest to x .

$$\text{dist}(x, C) \leq \text{dist}(x, f(a)) \leq \text{dist}(x, a) + \text{dist}(a, f(a))$$

$$\text{dist}(a, f(a)) \leq \text{dist}(a, b) \leq \text{dist}(a, x) + \text{dist}(x, b)$$

$$\text{dist}(x, C) \leq 2\text{dist}(x, A) + \text{dist}(x, B)$$

Analysis (cont'd)



Summing over x :

$$\text{cost}(F_k) \leq 2\text{cost}(F_k^{\text{off}}) + \text{cost}(F_\ell)$$

Iterating:

$$\text{For } k \in K : \quad \text{cost}(F_k) \leq 2 \sum_{\ell \geq k, \ell \in K} \text{cost}(F_\ell^{\text{off}})$$

How to choose K

k	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
$\text{cost}(F_k^{\text{off}})$	50	45	34	31	28	27	20	17	15	11	9	7	5	3	2	1

$$k = 6 \quad \text{OPT}_6 = 27 \quad 16 < 27 \leq 32$$

$$\begin{aligned}
 \text{cost}(F_k) &\leq 2(31 + 15 + 7 + 3 + 2 + 1) \\
 &\leq 2(32 + 16 + 8 + 4 + 2 + 1) \\
 &\leq 2(2F_k^{\text{off}} + F_k^{\text{off}} + F_k^{\text{off}}/2 + \dots) \\
 &\leq 8F_k^{\text{off}}
 \end{aligned}$$

$\leq (24 + 8\epsilon)\text{OPT}_k$ if F_k^{off} is a $(3 + \epsilon)$ -approximation

How to choose K (randomized)

Choose x uniformly at random in $[0, 1]$.

k	\dots	\times	\dots	\times	\dots	\times	\dots	\times
$\text{cost}(F_k^{\text{off}})$	\dots	e^{i+x}	$\dots \text{ cost}(F_k^{\text{off}}) \dots$	e^{2+x}	\dots	e^{1+x}	\dots	e^x

Analysis

$$\text{cost}(F_k) \leq 2e^{i+x} \left(1 + \frac{1}{e} + \frac{1}{e^2} + \dots\right) = \frac{2e^{i+x}}{1 - 1/e}$$

$$E\left(\frac{e^{i+x}}{F_k^{\text{off}}}\right) = \int_0^1 e^z dz = e - 1$$

$$E(\text{cost}(F_k)) \leq \frac{2(e-1)}{1-1/e} F_k^{\text{off}} = 2e F_k^{\text{off}}.$$

Extensions

- λ -relaxed metrics such as squared Euclidian distances
- Oblivious size-competitiveness: $\text{cost}(F_k) \leq \text{OPT}_k$, $|F_k| \leq sk$
- Hierarchical clustering [LNRW]
- Oblivious k -MST, k -vertex cover,... [LNRW]

Relation to previous work

How to choose K : algorithms **look familiar**

Online bidding

- **Universe:** U
- **Adversary:** $T \in U$
- **Algorithm:** sequence of bids $b \in U$, stops as soon as some bid is $\geq T$
- **Cost:** sum of algorithm's bids

Online bidding theorem

The optimal competitive ratio for online bidding equals 4 in the deterministic setting and e in the randomized setting.

Randomized lower bound is new

Recognizing online bidding

- [Charikar Chekuri Feder Motwani 97], [Dasgupta Long 01]
Hierarchical clustering to minimize maximum cluster radius/diameter.
End of analysis: reduction to online bidding.
- [Goemans Kleinberg 96] Minimum latency problem
- [Chakrabarti Phillips Schultz Shmoys Stein Wein 96]
Scheduling algorithms for minsum criteria
- [Motwani Phillips Torng 94] Non clairvoyant scheduling
- [Kao Reif Tate 93] cow path problem

Randomized lower bound

No randomized algorithm for online bidding can be better than e -competitive.

$$U = \{1, 2, \dots, n\}$$

Adversary: T

Algorithm successively bids $b_1 \leq b_2 \leq b_3 \leq \dots$

Partition of U :

$$[1, b_1] \cup [b_1 + 1, b_2] \cup [b_2 + 1, b_3] \cup \dots$$

Algorithm pays for bid b iff there is a $t \leq T$ such that $[t, b]$ is in the partition.

LP relaxation

$$\min \beta$$

$$\begin{cases} \sum_{t \leq T} bx(t, b) \leq \beta T & \forall T \\ \sum_{t \leq T, b \geq T} x(t, b) \geq 1 & \forall T \\ x(t, b) \geq 0 & \forall t, b \end{cases}$$

Write dual

Find feasible dual solution with value e

Title