## The (h, k)-Server Problem on Bounded-Depth Trees

# Nikhil Bansal ${ }^{1}$, Marek Eliás ${ }^{1}$, Łukasz Jeż², and Grigorios Koumoutsos ${ }^{1}$ 

1 TU Eindhoven
2 University of Wrocław

## The k-server problem

- One of the central problems in Online Algorithms
- Introduced by Manasse, McGeoch, Sleator '90


## The $k$-server problem

- One of the central problems in Online Algorithms
- Introduced by Manasse, McGeoch, Sleator '90


## Definition of the problem:

- We have $k$ servers in a metric space $S$
- At time t :
- Request arrives at some point $\sigma_{t} \in S$
- We have to decide what server to move to $\sigma_{t}$
- Target: minimize the total distance traveled by the servers


## The k-server problem

- $k$ servers in a metric space $S$
- At time t :
- Request arrives at some point $\sigma_{t} \in S$
- We have to decide what server to move to $\sigma_{t}$

$$
\mathrm{t}=0
$$



## The k-server problem

- $k$ servers in a metric space $S$
- At time t :
- Request arrives at some point $\sigma_{t} \in S$
- We have to decide what server to move to $\sigma_{t}$

$$
\mathrm{t}=1:
$$



- $a$


Ob

$$
\sigma=(\mathrm{c}, ?, ?, ?, \ldots)
$$

## The k-server problem

- $k$ servers in a metric space $S$
- At time t :
- Request arrives at some point $\sigma_{t} \in S$
- We have to decide what server to move to $\sigma_{t}$

$$
\mathrm{t}=1:
$$



0

$$
\sigma=(\mathrm{c}, ?, ?, ?, \ldots)
$$

## The k-server problem

- $k$ servers in a metric space $S$
- At time t :
- Request arrives at some point $\sigma_{t} \in S$
- We have to decide what server to move to $\sigma_{t}$

$$
\mathrm{t}=1
$$

cost:


O a
0
$\operatorname{dist}(\mathrm{c}, \mathrm{b})$

- b

0

$$
\sigma=(c, ?, ?, ?, \ldots)
$$

## The k-server problem

- $k$ servers in a metric space $S$
- At time t :
- Request arrives at some point $\sigma_{t} \in S$
- We have to decide what server to move to $\sigma_{t}$

$$
t=2
$$

cost:
$\operatorname{dist}(\mathrm{c}, \mathrm{b})$

$$
\left.\begin{array}{cccc}
c \bigcirc & \bigcirc a & 0 \\
& \bullet b & \bullet \\
\sigma=(\mathrm{c}, \mathrm{~b}, ?, ?, \ldots
\end{array}\right)
$$

## The k-server problem

- $k$ servers in a metric space $S$
- At time t :
- Request arrives at some point $\sigma_{t} \in S$
- We have to decide what server to move to $\sigma_{t}$

$$
t=2
$$

cost:
$\operatorname{dist}(\mathrm{c}, \mathrm{b})$

$$
\left.\begin{array}{cc}
c \bigcirc 0 & \bigcirc 0 \\
\sigma=(\mathrm{c}, \mathrm{~b}, ?, ?, \ldots
\end{array}\right)
$$

## The k-server problem

- $k$ servers in a metric space $S$
- At time t :
- Request arrives at some point $\sigma_{t} \in S$
- We have to decide what server to move to $\sigma_{\mathrm{t}}$

$$
t=2
$$

cost: $\operatorname{dist}(\mathrm{c}, \mathrm{b})$
$+\operatorname{dist}(\mathrm{a}, \mathrm{b})$
c 0

- $a$

$$
\sigma=(\mathrm{c}, \mathrm{~b}, ?, ?, \ldots)
$$

## Competitive ratio

## Performance evaluation:

$$
\mathrm{R}(\sigma)=\frac{\mathrm{ALG}_{\mathrm{k}}(\sigma)}{\mathrm{OPT}_{\mathrm{k}}(\sigma)} ; \quad \text { Competitive ratio }=\max _{\sigma}(\mathrm{R}(\sigma))
$$

- $\mathrm{ALG}_{\mathrm{k}}(\sigma)$ : cost of the online algorithm with k servers
- $\mathrm{OPT}_{k}(\sigma)$ : cost of the optimal offline solution for $k$ servers


## Competitive ratio

## Performance evaluation:

$$
\mathrm{R}(\sigma)=\frac{\mathrm{ALG}_{\mathrm{k}}(\sigma)}{\mathrm{OPT}_{k}(\sigma)} ; \quad \text { Competitive ratio }=\max _{\sigma}(\mathrm{R}(\sigma))
$$

- $\mathrm{ALG}_{\mathrm{k}}(\sigma)$ : cost of the online algorithm with k servers
- $\mathrm{OPT}_{k}(\sigma)$ : cost of the optimal offline solution for $k$ servers

Achievable ratio is of order $\Theta(k)$ (in deterministic case):

- LB: $k$ for any metric space of at least $k+1$ points (Manasse, McGeoch, Sleator '90)
- UB: k for DC in tree metrics (Chrobak et al. '91; Chrobak, Larmore '91)
- UB: $2 \mathrm{k}-1$ for the Work Function Algorithm by (Koutsoupias, Papadimitriou '95)


## The (h, k)-server problem

Servers as a precious resource:

- Does adding new servers help to decrease the cost?
- Standard setting:
- If we add more servers to ALG, we also compare it to OPT with more servers
- Does not tell us whether the cost of ALG decreased or not


## The (h, k)-server problem

Servers as a precious resource:

- Does adding new servers help to decrease the cost?
- Standard setting:
- If we add more servers to ALG, we also compare it to OPT with more servers
- Does not tell us whether the cost of ALG decreased or not

The ( $\mathrm{h}, \mathrm{k}$ )-server problem:

- h: \# of servers of OPT; k: \# of servers of ALG
- Fix $h$ and add new servers only to ALG, i.e. $k>h$ :


## The (h, k)-server problem

Servers as a precious resource:

- Does adding new servers help to decrease the cost?
- Standard setting:
- If we add more servers to ALG, we also compare it to OPT with more servers
- Does not tell us whether the cost of ALG decreased or not

The ( $\mathrm{h}, \mathrm{k}$ )-server problem:

- h: \# of servers of OPT; $k$ : \# of servers of ALG
- Fix $h$ and add new servers only to ALG, i.e. $k>h$ :

$$
\frac{\operatorname{ALG}_{h}(\sigma)}{\mathrm{OPT}_{h}(\sigma)} \approx h \quad \frac{\operatorname{ALG}_{k}(\sigma)}{\mathrm{OPT}_{h}(\sigma)} \approx \text { constant? }
$$

## Known results for the ( $\mathrm{h}, \mathrm{k}$ )-server problem

## Uniform metrics:

- Corresponds to the paging problem
- Tight bounds of $k /(k-h+1)$ by Sleator, Tarjan '85
- This equals 2 for $k=2 h$ and approaches 1 for $k \rightarrow \infty$
- Later generalized to weighted star metrics by Young '94


## Known results for the (h, k)-server problem

## Uniform metrics:

- Corresponds to the paging problem
- Tight bounds of $k /(k-h+1)$ by Sleator, Tarjan ' 85
- This equals 2 for $k=2 h$ and approaches 1 for $k \rightarrow \infty$
- Later generalized to weighted star metrics by Young '94


## General metrics:

- Lower bound of 2 irrespective of $k$ by Bar Noy and Schieber for the line
- Upper bound for WFA by Koutsoupias '99:
- $2 h$ for general metrics and $h+1$ for line irrespective of $k$
- No o(h) result known, even if $k \rightarrow \infty$


## Our results

Lower bound of $\Omega(h)$ for DC and WFA:

- In depth-2 trees for DC and depth-3 trees for WFA


## Our results

Lower bound of $\Omega(h)$ for DC and WFA:

- In depth-2 trees for DC and depth-3 trees for WFA

General lower bound for depth-2 trees:

- No algorithm can achieve a ratio better than 2.4 irrespective of $k$


## Our results

Lower bound of $\Omega(h)$ for DC and WFA:

- In depth-2 trees for DC and depth-3 trees for WFA

General lower bound for depth-2 trees:

- No algorithm can achieve a ratio better than 2.4 irrespective of $k$

New (deterministic) algorithm for bounded-depth trees:

- Its competitive ratio in a depth- $d$ tree is

$$
\begin{array}{lr}
O\left(d \cdot 2^{d+1}\right) & \text { for } k \rightarrow \infty \\
O\left(d \cdot(2 d / \epsilon)^{d+1}\right) & \text { for } k=(1+\epsilon) h
\end{array}
$$

- First bound sublinear in $h$ for a metric which is not uniform


## Roadmap for the rest of this talk

1. Definition of the depth-d trees
2. Lesson to learn from the lower bounds for DC and WFA
3. Description of our algorithm
4. New potential function based on Excess and Deficiency
5. Open problems

## Trees of bounded depth



Depth-d tree:

- Rooted tree
- Each path from root to leaf has d edges
- Requests only in leaves


## Trees of bounded depth

level d


Elementary subtrees:

- Subtrees of depth one
- They form a uniform metric
- In uniform metric: if $k \geqslant 2 h$, we can be 2-competitive (ST'85)


## Drawbacks of DC and WFA



## Drawbacks of DC and WFA



- DC and WFA bring help too slowly
- If cost of $x$ is incurred in the subtree R, DC moves $s$ by $\approx x / k_{R}$


## Drawbacks of DC and WFA



- DC and WFA bring help too slowly
- If cost of $x$ is incurred in the subtree R, DC moves $s$ by $\approx x / k_{R}$
- Our algorithm is more aggressive: we move $s$ by $\approx x$


## New algorithm: Paradigm

level d


Basic paradigm of the algorithm:

- Similar to DC by Chrobak et al.
- Servers are allowed to stay anywhere
- We move servers adjacent to the request


## New algorithm: Paradigm

level d


Basic paradigm of the algorithm:

- Each adjacent server keeps moving towards the request until
- its path to the request is blocked by some other server
- request is served
- Key part of our algorithm: careful choice of the speeds


## Algorithm: Phase 1



- One unit of speed divided between the servers in the elementary subtree containing the requested point
- Second unit of speed is divided to the other incoming servers proportionally to the number of servers they represent


## Algorithm: Phase 2



- The descending server is moving with the speed 1
- Second unit of speed is divided to the other incoming servers proportionally to the number of servers they represent


## Analysis: New potential function

## Excess and deficiency:

- Let T be a subtree rooted at a vertex at level $i$
- $\mathrm{k}_{\mathrm{T}}$ : number of ALG's servers in T
- $h_{\mathrm{T}}$ : number of OPT's servers in T
- If $\mathrm{k}_{\mathrm{T}} \geqslant \beta h_{\mathrm{T}}, \mathrm{T}$ is excessive; otherwise it is deficient
- Excess threshold $\beta$ depends on $i$ and on $k / h$
- $\beta=2$ for elementary subtrees if $k / h$ is large
- For subtrees rooted in higher levels $\beta$ increases geometrically



## Analysis: New potential function

The new potential:

$$
\Phi=\sum_{i=1}^{\mathrm{d}}\left(\alpha_{i}^{\mathrm{E}} \mathrm{E}_{i}+\alpha_{i}^{\mathrm{D}} \mathrm{D}_{i}\right)
$$

- $E_{i}$ and $D_{i}$ measure the excess/deficiency in subtrees rooted at level i
- $\alpha_{i}^{\mathrm{E}}$ and $\alpha_{i}^{\mathrm{D}}$ are coefficients (dependent only on $i$ and $k / h$ )

Core of the analysis:

- Proof that ALG always decreases enough excess or deficiency


## Open problems

Is dependence on d necessary?

- A constant independent on d would imply a randomized $\mathrm{O}(\log n)$-competitive algorithm for any metric on $n$ points

Can we achieve a similar result for other metrics?

- e.g. line would be very interesting
- Current best is $\frac{k}{k+1}(h+1)$ for DC

Is there a metric which allows a lower bound bigger than 2.4?

- Maybe even for trees of higher depth


## Thank You

## Can the ratio become worse with $k$ increasing?

## Can the ratio become worse with $k$ increasing?

## Answer: Yes!

## Can the ratio become worse with $k$ increasing?

Answer: Yes!

Performance of DC algorithm in the line and trees:

- In the standard setting optimal for line and trees
- Competitive ratio for DC is precisely

$$
\frac{k}{k+1}(h+1)
$$

- Equals $h$ for $k=h$, and increases towards $h+1$ for $k \rightarrow \infty$


## Can the ratio become worse with $k$ increasing?

## Answer: Yes!

Performance of DC algorithm in the line and trees:

- In the standard setting optimal for line and trees
- Competitive ratio for DC is precisely

$$
\frac{k}{k+1}(h+1)
$$

- Equals $h$ for $k=h$, and increases towards $h+1$ for $k \rightarrow \infty$

Competitive ratio of WFA in the line:

- Equals $h$ for $k=h$, but increases to $h+1 / 3$ for $k=2 h$.

