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

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Definition of the problem:

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

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Performance evaluation:

$$R(\sigma) = \frac{ALG_k(\sigma)}{OPT_k(\sigma)}; \qquad \text{Competitive ratio} = \max_{\sigma}(R(\sigma))$$

 \blacktriangleright ALG_k($\sigma)$: cost of the online algorithm with k servers

▶ $OPT_k(\sigma)$: cost of the optimal offline solution for k servers

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 Competitive ratio $= \max_{\sigma}(R(\sigma))$

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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: 2k 1 for the Work Function Algorithm by (Koutsoupias, Papadimitriou '95)

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

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$$\frac{\mathsf{ALG}_{\mathsf{h}}(\sigma)}{\mathsf{OPT}_{\mathsf{h}}(\sigma)} \approx \mathsf{h} \qquad \qquad \frac{\mathsf{ALG}_{\mathsf{k}}(\sigma)}{\mathsf{OPT}_{\mathsf{h}}(\sigma)} \approx \text{ constant}?$$

Uniform metrics:

- Corresponds to the paging problem
- Tight bounds of k/(k h + 1) by Sleator, Tarjan '85
 - This equals 2 for k = 2h and approaches 1 for $k \to \infty$
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General metrics:

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

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New (deterministic) algorithm for bounded-depth trees:

Its competitive ratio in a depth-d tree is

$$\begin{split} & O\left(d\cdot 2^{d+1}\right) & \quad \text{for } k \to \infty \\ & O\left(d\cdot (2d/\varepsilon)^{d+1}\right) & \quad \text{for } k = (1+\varepsilon)h \end{split}$$

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

- 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



Elementary subtrees:

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

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- \blacktriangleright If cost of x is incurred in the subtree R, DC moves s by $\approx x/k_R$
- \blacktriangleright Our algorithm is more aggressive: we move s by pprox x

New algorithm: Paradigm



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



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:

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



The new potential:

$$\Phi = \sum_{i=1}^{d} (\alpha_i^E E_i + \alpha_i^D D_i)$$

- E_i and D_i measure the excess/deficiency in subtrees rooted at level i
- ▶ α_i^E and α_i^D are coefficients (dependent only on i and k/h)

Core of the analysis:

▶ Proof that ALG always decreases enough excess or deficiency

Is dependence on d necessary?

A constant independent on d would imply a randomized
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?

Answer: Yes!

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

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

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Competitive ratio of WFA in the line:

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