Improved On-line Broadcast Scheduling with Deadlines*

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Abstract. We study an on-line broadcast scheduling problem in which requests have deadlines, and the objective is to maximize the weighted throughput, i.e., the weighted total length of the satisfied requests. For the case where all requested pages have the same length, we present an online deterministic algorithm named BAR and prove that it is 4.56-competitive. This improves the previous algorithm of Kim and Chwa [11] which is shown to be 5-competitive by Chan et al. [4]. In the case that pages may have different lengths, we prove a lower bound of $\Omega(\Delta/\log\Delta)$ on the competitive ratio where Δ is the ratio of maximum to minimum page lengths. This improves upon the previous $\sqrt{\Delta}$ lower bound in [11, 4] and is much closer to the current upper bound of $(\Delta + 2\sqrt{\Delta} + 2)$ in [7]. Furthermore, for small values of Δ we give better lower bounds.

1 Introduction

Data broadcast scheduling is a core problem in many applications that involve distribution of information from a server to a large group of receivers. In contrast to the traditional point-to-point mode of communication, broadcasting technologies are employed so that different clients requesting the same data can be satisfied simultaneously by only one broadcast. For example, in Hughes' DirecPC system [12], clients make requests over phone lines and the server satisfies the requests through broadcasts via a satellite. Typical information that will be broadcasted include movies (video on-demand), stock market quotation, traffic

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and landmark information, etc. Very often, the information to be disseminated is time critical and thus it is important to meet the deadlines of the requests.

Motivated by these applications, we study the following *On-line Scheduling of Broadcasts* (Broadcasting): We are given a set of pages to be broadcasted to clients upon request. Each request r has four attributes, namely, $p(r)$: the requested page, $a(r)$: the arrival time, $d(r)$: the deadline by which the requested page has to be received in its entirety, and $w(r)$: the weight of the request. A request is not known until it arrives, i.e., at time $a(r)$. When it arrives, all $p(r)$, $d(r)$ and $w(r)$ become known. When the server broadcasts a page, all requests to the same page that have arrived will receive the content of the page simultaneously. Upon completion, each of these requests will be satisfied, provided that the completion time is before its respective deadline. The server is allowed to abort the current page it is broadcasting before its completion and start a new one. To satisfy an aborted request, the requested page has to be broadcasted again from the beginning. Thus it is an on-line scheduling problem with preemptions and restarts. Our goal is to maximize the total weighted throughput, i.e., the total weighted lengths of all satisfied requests.

Related Work. Most of the previous works on the problem of on-line broadcast scheduling concentrate on minimizing the flow time where the flow time of a request is the time elapsed between its arrival and its completion. For example, [10, 5, 6, 8] studied the problem of minimizing the total flow time while Bartal and Muthukrishan [2] studied the minimization of the maximum flow time. Aksoy and Franklin [1] presented a practical parameterized algorithm and evaluated it with extensive experiments. While the flow time is important and related to how the clients perceive the responsiveness of the system, the objective of maximizing the throughput is crucial for applications in which requests are associated with deadlines. Jiang and Vaidya [9] considered the problem of maximizing the percentage of satisfied requests assuming knowledge of the requests distribution.

Kim and Chwa [11] were the first to design algorithms with provable worst case performance bounds for this problem. In particular, one of their results is a 5.828-competitive algorithm for our problem. Using a tighter analysis, Chan et al [4] showed that Kim and Chwa's algorithm actually has competitive ratio at most 5, through a reduction to a job scheduling problem with cancellations. It was shown [14] that we need new techniques to further improve the bound. For the case of different page lengths, let Δ be the ratio between the length of the longest and shortest page. There is a $(\Delta + 2\sqrt{\Delta} + 2)$-competitive algorithm [7] and a $\sqrt{\Delta}$ lower bound [11, 4].

A related on-line interval scheduling problem is studied by Woeginger [13]. Translated into our terminologies, his problem is to schedule requests with tight deadlines (i.e., the length of time interval between its arrival time and deadline is exactly the length of the requested page) and pages have different lengths. He proved that when the page length and the weight of requests can be arbitrarily related, no deterministic algorithm can have constant competitive ratio. He went on to give a 4-competitive heuristic for several special cases in which the page length and the weight of requests satisfy certain relationship. In particular, the

heuristic works for the case of unit page length and arbitrary weights. He then complemented his upper bound with several lower bounds, including a tight lower bound for this special case of unit page length. In some sense, our problem is a generalization of Woeginger's problem allowing non-tight intervals.

Our Results. In this paper we give three different results for Broadcasting. Our first contribution is to give an improved algorithm for the case of unit page length. We consider the deadlines of the requests, a parameter ignored by previous algorithms, in our scheduling decision. By considering the fact that some of the currently-serving requests might have distant deadlines and can be served later after the completion of new requests, we improve the competitive ratio from 5 [4] to 4.56. In Section 3 we describe this algorithm and its analysis.

Our second contribution is to give an improved lower bound for the case of different page lengths. We give a lower bound of $\Omega(\Delta/\log\Delta)$ on the competitive ratio, an improvement over the previous $\Omega(\sqrt{\Delta})$ bound and which almost matches the linear upper bound. This is discussed in Section 4.

All existing lower bound proofs for the case of different page lengths do not work very well when $\Delta > 1$ is small, and thus the lower bound for those cases is still 4, that being the lower bound for the unit page length case. In Section 5 we describe our third contribution of proving better lower bounds for these cases. The lower bound for the competitive ratio is improved to e.g., 4.245, 4.481, 4.707 and 5.873 for $\Delta = 2$, 3, 4 and 10 respectively. The result is obtained by extending the lower bound proof for the unit page length case [13] to the case of different page lengths.

Due to space constraints, most proofs are omitted from this version. They can be found in the full version of the paper.

2 Notations

We first state the problem formally. Assume there are n pages, $P_1, \ldots, P_n$, in the system. A request for some page may arrive in arbitrary time with arbitrary deadline. If a page is fully broadcasted, then a request for that page is *satisfied* if the requests arrive before the broadcast of the page, and the broadcast finishes before the deadline of the request. A broadcast can be aborted at any time, and can later be restarted from the beginning.

A schedule S is a sequence of broadcasts $J_1, J_2, \ldots$ where each broadcast J_i is a set of requests to the same page started being served at time $s(J_i)$. The broadcasts are indexed such that $s(J_i) < s(J_{i'})$ for $i < i'$. For convenience, we will write J_i to represent both the set of requests and the requested page. Let $l(J_i)$ be the length of the page broadcasted by J_i. If $s(J_i) + l(J_i) > s(J_{i+1})$, then the broadcast J_i is *aborted* by J_{i+1}; otherwise J_i is said to be *completed*. The *profit* of a completed request is its weight times the length of the page it requests. The profit of a broadcast J_i, denoted by $|J_i|$, is the sum of $w(r) \times l(J_i)$ over all r in J_i. We denote by $|S|$ the total profit of completed broadcasts in the schedule S, i.e., we only count those satisfied requests. The objective is to maximize the total profit of satisfied requests $|S|$ during a time period.

Given an input I (a set of requests) and an algorithm A, we denote by $S_A(I)$ and $S^*(I)$ the schedules produced by A and by an optimal offline algorithm on I respectively. When A and I are understood from the context, we will simply denote the schedules by S and S^* respectively. To gauge the quality of the schedules produced by an on-line algorithm, the competitive ratio analysis [3] is often used. The competitive ratio of algorithm A is defined as $r_A = \sup_I \frac{|S^*(I)|}{|S_A(I)|}$.

3 Unit Page Length: the BAR Algorithm

In this section, we consider the case where each page is of the same length. Thus we assume without loss of generality that broadcasting each page requires one unit of time. We present our BAR algorithm (for Bi-level Abortion Ratio) and prove that it is 4.56-competitive.

At any moment, there is a (possibly empty) pool of requests that have arrived and are waiting to be served. Requests in the pool will be discarded when their deadlines cannot be met even if the server starts to serve them immediately. The algorithm is triggered either when a broadcast is completed or when a new request arrives. In the former case, the server will pick the page with the largest total profit of requests in the pool to be broadcasted next. When a new request arrives while the server is broadcasting a page, the server will either continue the current broadcast J (and add the new requests to the pool), or abort J in favour of a new broadcast R with larger total profit of requests in the pool (including newly arrived requests and possibly part of J). The decision is made according to the relative profits of J and R. It will also consider the previous broadcast and the deadlines of requests currently in the system.

More precisely, let J_0 be the broadcast aborted by J. If J does not abort any broadcast, define $J_0 = \phi$ (the empty set). Let J' be the largest-profit set of requests that can be satisfied in a single broadcast after completing J, assuming no more requests arrive. Similarly R' denotes the largest-profit set of requests that can be satisfied in one broadcast after completing R, if we abort J and start R now. See Figure 1. Let α, β be some constants such that $1.5 \leq \alpha < 2 \leq \beta < 2.5$. The exact values will be determined later. If either one of the following conditions is satisfied, we abort J and start R:

C1: $\beta|J| \leq |R|$ and $\beta^2|J_0| \leq |R|$, or
C2: $\alpha|J| \leq |R| < \beta|J|$, $\beta|J| + |J'| \leq |R| + |R'|$, and $\beta|J_0| \leq |J|$.

Otherwise, we continue with the broadcasting of J and add the new requests into the pool.

We give some intuitive rationale behind these conditions. In previous algorithms [11, 4], the abortion is simply determined by considering whether $\beta|J| \leq |R|$. This completely ignores the deadlines of the requests. The improvement of BAR comes from the introduction of [C2], which gives an alternative abortion condition with a lower threshold α and, as we show below, considers the deadlines of requests.

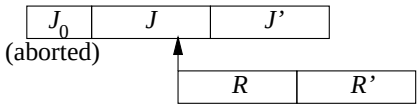

Fig. 1. BAR is broadcasting J and determining whether to abort J and start R.

The first part of condition [C1] is the usual abortion condition by considering the profit of requests. The second part of [C1] enforces some technical properties that are required in bounding the profit of the requests, despite some abortions being caused by condition [C2].

The second part of condition [C2] utilizes deadlines of requests. Rather than directly comparing the deadlines of requests, we consider the total profit of requests that can later be satisfied (before their deadlines) in deciding whether to abort the page currently being broadcast. Suppose an abortion happens due to condition [C2]. Then some part of R' must come from J (which is aborted). Otherwise, the requests in R' come from the pool only, and since the broadcast J' will finish earlier than R', R' is a possible choice of J'. Hence $|R'| \leq |J'|$ and the condition $\beta|J| + |J'| \leq |R| + |R'|$ cannot be true. So $R' \cap J$ is not empty, and they must be requesting the same page. That means once condition [C2] happens, there must be some requests in J that have long deadlines, long enough to be completed after R is completed. Therefore, even though $|R|$ is only α times larger than $|J|$, we may still satisfy enough requests to achieve a good competitive ratio.

The third part of condition [C2] ensures there will not be two consecutive abortions caused by [C2], so that this weaker-threshold condition will not be used too often.

The remaining of this section is devoted to the proof of the following theorem.

Theorem 1. *BAR is* 4.56*-competitive for* Broadcasting *in the unit page length case.*

3.1 Basic Subschedules

We now elicit certain useful structures in a schedule produced by BAR. A sequence of broadcasts $(J_1, \ldots, J_k)$ is called a *chain* if J_i is aborted by J_{i+1} for all $i = 1, \ldots, k-1$ and J_1 is preceded by either an idle interval or a completed broadcast. A chain $C = (J_1, \ldots, J_k)$ in which J_k is completed is called a *basic subschedule*. Thus a basic subschedule consists of a sequence of zero or more aborted broadcasts followed by a completed broadcast; and the sequence cannot be extended at the front. Furthermore, the broadcast before an idle interval must be completed. Therefore, the whole schedule can be decomposed into a sequence of basic subschedules.

Consider an arbitrary basic subschedule, $B = (J_1, J_2, \ldots, J_k)$ where J_k is a completed broadcast and all the others are aborted ones. The total profit of

requests satisfied by BAR in this basic subschedule is $|B| = |J_k|$. To analyze the competitive ratio, we will associate with B a carefully chosen set of requests satisfied by the offline optimal algorithm $\mathcal{O}$.

Consider a broadcast J by BAR. We can make use of condition [C1] and/or [C2] to argue that requests started by $\mathcal{O}$ while BAR is broadcasting J cannot be too large, if these requests are available in the pool maintained by BAR. Note, however, that $\mathcal{O}$ can also serve requests with arbitrarily large profits that have been satisfied by BAR before without violating [C1, C2]. Thus, we classify the requests satisfied by $\mathcal{O}$ into two types according to whether the request has been satisfied by BAR at the time $\mathcal{O}$ starts them. (Since $\mathcal{O}$ is an offline algorithm, we assume that it will never abort a broadcast. Thus, saying that a request is started by $\mathcal{O}$ is equivalent to saying that it will be satisfied by $\mathcal{O}$.) More precisely, we define J_i^*, for $i = 1, \ldots, k$, as the set of requests started by $\mathcal{O}$ within the interval $[s(J_i), s(J_{i+1}))$ but have not been satisfied by BAR before, where $s(J_i)$ is the start time of J_i and we take $s(J_{k+1}) = s(J_k) + 1$. Also, we define B^* as the set of requests in J_k that are started by $\mathcal{O}$ after the basic subschedule B. We will try to obtain an upper bound on $\sum_{i=1}^{k} |J_i^*| + |B^*|$.

Note that if a broadcast J_i is aborted by a broadcast J_{i+1} due to condition [C1], the ratio $|J_{i+1}|/|J_i|$ is at least β. However if the abortion is due to condition [C2], $|J_{i+1}|/|J_i|$ may be smaller than β. Nevertheless, we can still bound the profits of J_i's and J_i^*'s by geometric series in the following lemmas. Consider a chain $C = (J_1, \ldots, J_k)$. It is said to have *big endian* if $|J_k| \geq \beta |J_{k-1}|$; and *small endian* otherwise. If C has only one broadcast, we take $J_{k-1} = \phi$. Thus C will be considered to have big endian.

The following two lemmas bound the profits of J_i and J_i^*.

Lemma 1. *Consider a chain $C = (J_1, \ldots, J_k)$ with big endian. Then $|J_i| \leq |J_k|/\beta^{k-i}$ for all $i = 1, \ldots, k$.*

Lemma 2. *Consider a chain $C = (J_1, \ldots, J_k)$ with big endian. Then $|J_i^*| < |J_k|/\beta^{k-i-1}$ for all $i = 1, \ldots, k$. Hence $\sum_{i=1}^{k} |J_i^*| < \frac{\beta^2}{\beta-1}\left(1 - \frac{1}{\beta^k}\right)|J_k|$.*

The following lemma bounds the profit of requests served by $\mathcal{O}$ in a basic subschedule.

Lemma 3. *Consider a basic subschedule $B = (J_1, \ldots, J_k)$. If B has small endian,*

$$\sum_{j=1}^{k} |J_j^*| + |B^*| \leq \left(\beta + \frac{1}{\beta - 1}\right)\beta |J_{k-1}| + |J_k| + |J_k'| - \frac{\beta}{\beta - 1}|J_1|.$$

If B has big endian,

$$\sum_{j=1}^{k} |J_j^*| + |B^*| \leq \left(\alpha + \frac{\beta}{\beta - 1} + 1\right)|J_k| + |J_k'| - \frac{\beta}{\beta - 1}|J_1|.$$

Proof. Suppose B has small endian. We observe that the chain $(J_1, \ldots, J_{k-1})$ must have big endian by construction of BAR. (Note that $k \geq 2$ if B has small endian since $k = 1$ implies B has big endian.) By Lemma 2, we have $\sum_{j=1}^{k-1} |J_j^*| < \frac{\beta^2}{\beta-1}\left(1 - \frac{1}{\beta^{k-1}}\right)|J_{k-1}|$. Also, we have $|J_k^*| < \beta^2|J_{k-1}|$ for otherwise, J_k^* would have aborted J_k due to condition [C1]. Thus $\sum_{j=1}^{k} |J_j^*| < \frac{\beta^3}{\beta-1}\left(1 - \frac{1}{\beta^k}\right)|J_{k-1}|$. As for B^*, we note that $\beta|J_{k-1}| + |J'_{k-1}| \leq |J_k| + |J'_k|$ since J_k aborts J_{k-1} by condition [C2]. Moreover, $|B^*| \leq |J'_{k-1}|$ because requests in B^* have deadlines no earlier than that of J'_{k-1} and they have not been satisfied by BAR at time $s(J'_{k-1})$. Hence $|B^*| \leq |J_k| - \beta|J_{k-1}| + |J'_k|$. Combining these two bounds, we have $\sum_{j=1}^{k} |J_j^*| + |B^*| \leq \frac{\beta^3}{\beta-1}\left(1 - \frac{1}{\beta^k}\right)|J_{k-1}| + (|J_k| - \beta|J_{k-1}| + |J'_k|) = \left(\beta + \frac{1}{\beta-1}\right)\beta|J_{k-1}| + |J_k| + |J'_k| - \frac{1}{(\beta-1)\beta^{k-3}}|J_{k-1}| \leq \left(\beta + \frac{1}{\beta-1}\right)\beta|J_{k-1}| + |J_k| + |J'_k| - \frac{\beta}{\beta-1}|J_1|$.

We omit the proof for the big endian case. □

In Lemma 3, no matter B has big or small endian, we can bound $|J'_k|$ from above by the profit of the first broadcast in the basic subschedule after B. If B is followed by an idle interval, then we can actually argue that $|J'_k| = 0$. That is, we associate $|J'_k|$ with the basic subschedule following B. By the same token, B will also be associated with such value from the preceding basic subschedule. In the next subsection we will see how this association is used in the analysis.

3.2 Aggregated Subschedules

Let $B_i = (J_{i,1}, \ldots, J_{i,k_i})$ denote the i-th basic subschedule in a sequence of basic subschedules. For notational convenience define $J_{i,0} = \phi$, and $prev(B_i) = J_{i,k_i-1}$, i.e., the second last broadcast in a basic subschedule.

Lemma 4. *The last basic subschedule before an idle interval must have big endian.*

Based on the above lemma, we can partition the original schedule into a number of *aggregated subschedules*, each of which containing zero or more basic subschedules with small endians followed by one basic subschedule with big endian.

Consider an arbitrary aggregated subschedule, $A = (B_1, \ldots, B_m)$ where for $i = 1, \ldots, m$, $B_i = (J_{i,1}, \ldots, J_{i,k_i})$ is a basic subschedule with k_i broadcasts.

Obviously, the total profit of requests satisfied by BAR in A is

$$|A| = \sum_{i=1}^{m} |J_{i,k_i}|. \tag{1}$$

Also, since $|J_{i,k_i}| \geq \alpha|prev(B_i)|$ for $i = 1, \ldots, m-1$, we have

$$|A| \geq \alpha\left(\sum_{i=1}^{m-1} |prev(B_i)|\right) + |J_{m,k_m}|. \tag{2}$$

Recall that B_i's have small endians for $i = 1, \ldots, m-1$. We have $k_i \geq 2$ since basic subschedules with only one broadcast must have big endians by definition. By condition [C2], $\beta|prev(B_i)|+|J'| \leq |J_{i,k_i}|+|R'|$ where $|R'| \leq |J_{i+1,1}|$ because R' is a candidate set of requests to be served after J_{i,k_i} is completed. Hence we have $\beta|prev(B_i)| \leq |J_{i,k_i}| + |J_{i+1,1}|$, and together with (1),

$$|A| \geq \sum_{i=1}^{m-1} \beta|prev(B_i)| + |J_{m,k_m}| - \sum_{i=2}^{m} |J_{i,1}|. \tag{3}$$

On the other hand, the total profit of requests satisfied by $\mathcal{O}$ and associated with aggregated subschedule A is:

$$\begin{aligned}
|A^*| &= \left(\sum_{j=1}^{k_1} |J^*_{1,j}| + \cdots + \sum_{j=1}^{k_m} |J^*_{m,j}|\right) + (|B^*_1| + \cdots + |B^*_m|) \\
&\leq \sum_{i=1}^{m-1} \left(\beta + \frac{1}{\beta - 1}\right) \beta|prev(B_i)| + \left(\alpha + \frac{\beta}{\beta - 1}\right) |J_{m,k_m}| \\
&\qquad + \sum_{i=1}^{m} |J_{i,k_i}| + \sum_{i=2}^{m} |J_{i,1}| + |J'_{m,k_m}| - \frac{\beta}{\beta - 1} \sum_{i=1}^{m} |J_{i,1}|
\end{aligned}$$

where the inequality follows from Lemma 3. Consider (1) + (2) $\times(\frac{\beta^2}{\alpha})$+ (3) $\times\frac{1}{\beta-1}$. After some algebraic manipulations (which we omit) we have

$$\left(1 + \frac{\beta^2}{\alpha} + \frac{1}{\beta - 1}\right) |A| \geq |A^*| + |J_{1,1}| - |J'_{m,k_m}| \tag{4}$$

as long as $\frac{\beta^2}{\alpha} + \frac{1}{\beta-1} \geq \frac{\beta}{\beta-1} + \alpha$. We can bound $|J'_{m,k_m}|$ from above by the profit of the first broadcast (i.e., $|J_{1,1}|$) in the next aggregated subschedule. Thus, if we have a sequence of aggregated subschedules $A_1, \ldots, A_l$, then from (4) we have

$$\left(\frac{\beta^2}{\alpha} + \frac{\beta}{\beta - 1}\right) (|A_1| + \cdots + |A_l|) \geq |A^*_1| + \cdots + |A^*_l| - |J'| \tag{5}$$

where J is the last broadcast in A_l. Since there is no more broadcast after A_l, $|J'| = 0$.

If the whole aggregated subschedule consists of only one basic subschedule with big endian, i.e., $A = (B_1)$, then $|A| = |J_{1,k_1}|$ and we can verify that inequality (5) still holds.

The condition $\frac{\beta^2}{\alpha} + \frac{1}{\beta-1} \geq \frac{\beta}{\beta-1} + \alpha$ can be satisfied by having $\alpha^2 + \alpha \leq \beta^2$, i.e., $\alpha \leq \sqrt{\beta^2 + \frac{1}{4}} - \frac{1}{2}$. Setting $\alpha = \sqrt{\beta^2 + \frac{1}{4}} - \frac{1}{2}$, the competitive ratio of BAR is

$$\frac{\sum_{i=1}^{l} |A^*_i|}{\sum_{i=1}^{l} |A_i|} \leq \frac{\beta}{\beta - 1} + \frac{\beta^2}{\alpha} = \frac{\beta}{\beta - 1} + \frac{\beta^2}{\sqrt{\beta^2 + \frac{1}{4}} - \frac{1}{2}} = \frac{3}{2} + \frac{1}{\beta - 1} + \sqrt{\beta^2 + \frac{1}{4}}.$$

This has a minimum value of approximately 4.561 attained when $\beta \approx 2.015$, and $\alpha \approx 1.576$.

4 Variable Page Length: An Improved Lower Bound

In this section we consider the case where the pages can have different lengths. We give a lower bound on the competitive ratio of any deterministic online algorithm for Broadcasting. Let Δ be the ratio between the length of the longest and shortest page.

Theorem 2. *The competitive ratio of any deterministic online algorithm for* Broadcasting *cannot be smaller than* $\Omega(\Delta/\log\Delta)$.

Proof. Assume that there are two pages, P and Q whose lengths are Δ and 1, respectively. Given any online algorithm $\mathcal{A}$, we construct a sequence of requests as follows. At time 0, a request for P arrives with deadline at time Δ, i.e., it has a tight deadline. The weight of the request is 1. There are at most $\lceil\Delta\rceil$ requests for Q, denoted by r_i for $0 \le i \le \lceil\Delta\rceil - 1$. r_i arrives consecutively, i.e., $a(r_i) = i$, and they all have tight deadlines, i.e., $d(r_i) = i+1$. The weight of r_i, i.e., $w(r_i)$, is $\Delta^r(i+1)^k$ where r and k are some constants which will be defined later. If $\mathcal{A}$ broadcasts Q at any time t, no more request of r_i arrives for $i > t$.

Now we analyze the performance of $\mathcal{A}$ against the optimal offline algorithm $\mathcal{O}$. There are two cases. (1) If $\mathcal{A}$ satisfies the request for P by broadcasting P at time 0, $\mathcal{O}$ will satisfy all requests r_i by broadcasting Q at time i for $0 \le i \le \lceil\Delta\rceil - 1$. Hence, we have $|\mathcal{A}| = \Delta$ and $|\mathcal{O}| = \sum_{i=0}^{\lceil\Delta\rceil-1} \Delta^r(i+1)^k$. Since $\sum_{i=1}^{x} i^y = \Theta(x^{y+1}/(y+1))$, $|\mathcal{O}|/|\mathcal{A}| = \Theta(\Delta^{r-1}\lceil\Delta\rceil^{k+1}/(k+1)) \ge \Theta(\Delta^{r+k}/(k+1))$.

(2) If $\mathcal{A}$ broadcasts Q at time t, only r_t can be satisfied. However, $\mathcal{O}$ can either satisfy the request for P by broadcasting P at time 0 or satisfy all r_i by broadcasting Q at time i for $0 \le i \le t$. We have $|\mathcal{A}| = \Delta^r(t+1)^k$ and $|\mathcal{O}| = \max\{\Delta, \sum_{i=0}^{t} \Delta^r(i+1)^k\}$. Hence, $|\mathcal{O}|/|\mathcal{A}| = \max\{\Delta, \Theta(\Delta^r(t+1)^{k+1}/(k+1))\}/\Delta^r(t+1)^k = \max\{\Delta^{1-r}/(t+1)^k, \Theta((t+1)/(k+1))\}$. In order to minimize the ratio, $\mathcal{A}$ should choose $t = (\Delta^{1-r}(k+1))^{1/(k+1)} - 1$. In that case, the ratio is $\Theta(\Delta^{(1-r)/(k+1)}/(k+1)^{k/(k+1)}) \ge \Theta(\Delta^{(1-r)/(k+1)}/(k+1))$.

In order to maximize the minimum ratio among the two cases, we let $r = (1-k-k^2)/(k+2)$. Hence, the competitive ratio is $\Theta(\Delta^{1-1/(k+2)}/(k+1)) \ge \Theta(\Delta^{1-1/(k+2)}/(k+2))$. We further let $k+2 = \ln\Delta$ where the function $\Delta^{1-1/(k+2)}/(k+2)$ achieves the maximum, i.e., $\Delta^{1-1/\ln\Delta}/\ln\Delta$. Since $\Delta^{1/\ln\Delta}$ is the constant e, i.e., the base of natural logarithm, we have proved that the competitive ratio is $\Omega(\Delta/\log\Delta)$. □

5 Lower Bound for Small Δ

The current lower bound for the broadcasting problem is 4 when $\Delta = 1$ [13]. The $\Omega(\Delta/\log\Delta)$ lower bound we just proved as well as the previous $\sqrt{\Delta}$ one [4] gives very small lower bounds (much smaller than 4) for small values of Δ. In this section we give better lower bounds for this case.

Let α be the unique positive real root of the equation

$$2\alpha^2 - 4\alpha^{3/2} + 2\alpha + 1 = \sqrt{(\lfloor\Delta\rfloor - 1)^2 + 4\alpha^2} + \lfloor\Delta\rfloor. \tag{6}$$

The following table shows some values of α.

Δ	1	2	3	4	10	very large
α	4	4.245	4.481	4.707	5.873	$\sqrt{\Delta}$

Theorem 3. *For $\Delta \geq 2$, no deterministic algorithm for* Broadcasting *can be better than α-competitive, where α is the unique positive root of* (6).

The proof of this theorem uses a modified construction from the lower bound proof in [13], by adding requests with different lengths and carefully setting the arrival time of requests in a different way.

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