

# Matching on the Line Admits No $o(\sqrt{\log n})$ -Competitive Algorithm

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## Abstract

We present a simple proof that the competitive ratio of any randomized online matching algorithm for the line exceeds  $\sqrt{\log_2(n+1)}/15$  for all  $n = 2^i - 1 : i \in \mathbb{N}$ , settling a 25-year-old open question.

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## 1 Online matching, on the line

In *online metric matching* [7, 9]  $n$  points of a metric space are designated as *servers*. One by one  $n$  requests arrive at arbitrary points of the space; upon arrival each must be matched to a yet unmatched server, at a cost equal to their distance. Matchings should minimize the ratio between the total cost and the *offline* cost attainable if all requests were known beforehand. A matching algorithm is  $c(n)$ -competitive if it keeps this ratio no higher than  $c(n)$  for all possible placements of servers and requests.

It is widely acknowledged [1, 10, 14] that the line is the most interesting metric space for the problem. Matching on the line models many scenarios, like a shop that must rent to customers skis of approximately their height, where a stream of requests must be serviced with minimally mismatched items from a known store. Despite matching being specifically studied on the line since at least 1996 [8], no tight competitiveness bounds are known.

As for upper bounds, the line is a doubling space and thus admits an  $O(\log n)$ -competitive randomized algorithm [5]; a sequence of recent developments [1, 12, 13] yielded the same ratio without randomization. Better bounds have been obtained only by algorithms with additional power, such as that to re-assign past requests [6, 11] or predict future ones [2].

As for lower bounds, the competitive ratio is at least 4.591 for randomized algorithms and 9 for deterministic ones since the *cow-path* problem is a special case of matching on the line [8]. These bounds were conjectured tight [8] until a complex adversarial strategy yielded a lower bound of 9.001 for deterministic algorithms [4]. Beyond some  $\Omega(\log n)$  bounds for restricted classes of algorithms [3, 10, 12], there has been no further progress on the lower-bound side before this work.



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## 2 An $\Omega(\sqrt{\log n})$ -competitiveness bound

We prove a simple  $\Omega(\sqrt{\log n})$  lower bound on the competitive ratio of randomized online matching algorithms for the line.

For any  $n = 2^i - 1$  with  $i \in \mathbb{N}$  consider the  $[0, n+1]$  interval; for each positive integer  $j \leq n$  place a server at point  $j$ , and place  $n$  requests over  $\log_2(n+1)$  rounds as follows. On the  $r^{\text{th}}$  round (for  $1 \leq r \leq \log_2(n+1)$ ) partition the interval into  $(n+1)/2^r$  subintervals of length  $2^r$ , choose within each uniformly and independently at random an *origin* point, and place a request on the closest integer multiple of  $2^{-n}$  breaking ties arbitrarily. “Discretizing” requests instead of directly using the corresponding origins prevents some technical difficulties – see our remark at the end.

We prove in Lemma 1 that the expected distance between the  $\ell^{\text{th}}$  leftmost server and the  $\ell^{\text{th}}$  leftmost origin is  $O(\sqrt{\log n})$ , so servers and requests can be matched with an expected offline cost  $O(n\sqrt{\log n})$ . Conversely, we prove in Lemma 2 that *any* online matching algorithm ALG incurs an expected  $\Omega(n)$  cost in any given round, for a total cost  $\Omega(n \log n)$ . The two results can be combined to prove that on *some* request sequence ALG incurs  $\Omega(\sqrt{\log n})$  times the offline cost.

► **Lemma 1.** *The expected distance between the  $\ell^{\text{th}}$  leftmost origin and the  $\ell^{\text{th}}$  leftmost server is at most  $\sqrt{\log_2(n+1)} + 3$ .*

**Proof.** Let  $S_\ell$  be  $\ell^{\text{th}}$  leftmost server and  $g_\ell$  be the number of origins to its left. Note that if  $g_\ell$  equals respectively  $\ell$  or  $\ell - 1$ , the  $\ell^{\text{th}}$  origin is the first immediately to the left, or to the right of  $S_\ell$ ; and since the first round placed one origin in every subinterval of size 2, such an origin is within distance 3 of  $S_\ell$ . By the same token, denoting by  $\delta_\ell$  the quantity  $|g_\ell - (\ell - \frac{\ell}{n+1})|$ , the  $\ell^{\text{th}}$  leftmost origin is within distance  $2\delta_\ell + 3$  of  $S_\ell$ . Note that  $\delta_\ell$  is the absolute deviation from the mean of  $r_\ell$ , since  $r_\ell$  is the sum of  $n$  independent indicator random variables each denoting whether a given origin was placed to the left of  $S_\ell$ , with total expectation  $\frac{n}{n+1}\ell = \ell - \frac{\ell}{n+1}$  (by construction, the expected density of origins is constant throughout the main interval). At most one such variable in a given round has variance greater than 0, albeit obviously at most  $1/4$ : that corresponding to the origin placed in a subinterval holding  $S_\ell$  strictly in its interior. Adding the individual variances we obtain that the variance of  $r_\ell$ , i.e. the expectation of  $\delta_\ell^2$ , is at most  $\log_2(n+1)/4$ ; and since by Jensen’s inequality  $E[\delta_\ell] \leq E[\delta_\ell^2]^{\frac{1}{2}}$ , the expected distance between  $S_\ell$  and the  $\ell^{\text{th}}$  leftmost origin is at most  $\sqrt{\log_2(n+1)} + 3$ . ◀

► **Lemma 2.** *Any randomized online matching algorithm incurs an expected cost greater than  $(n+1)/12$  in each round.*

**Proof.** Consider an origin placed uniformly at random in a subinterval of size  $2^r$  during the  $r^{\text{th}}$  round. Assume  $m$  unmatched servers in the interior points of that subinterval divide it into  $m + 1$  segments of (integer) length  $d_0, \dots, d_m$ . Then the probability the corresponding request falls within a segment of length  $d$  is  $d/2^r$ , in which case the expected distance of the request from the segment’s closer endpoint is  $d/4$ . Adding over all the  $s_r$  segments in all the round’s subintervals, applying Jensen’s inequality, and noting that  $s_r$  does not exceed the number of subintervals (i.e.  $(n+1)/2^r$ ) plus the total number of unmatched servers (i.e.  $(n+1)/2^{r-1} - 1$ ), the expected cost to service all requests in the round is at least:

$$\sum_{h=1}^{s_r} \frac{d_h}{4} \cdot \frac{d_h}{2^r} \geq \frac{1}{4 \cdot 2^r} s_r \left( \frac{n+1}{s_r} \right)^2 > \frac{(n+1)^2}{4 \cdot 2^r} \cdot \frac{2^r}{3(n+1)} = \frac{n+1}{12}. \quad \blacktriangleleft$$

We can then easily prove the following:

► **Theorem.** *The competitive ratio of any randomized online matching algorithm for the line exceeds  $\sqrt{\log_2(n+1)}/15$  for all  $n = 2^i - 1 : i \in \mathbb{N}$ .*

**Proof.** Let  $C_A(\sigma)$  be the expected cost incurred by a randomized online matching algorithm ALG on a request sequence  $\sigma$ , and  $C_O(\sigma)$  the offline cost; and let  $p_\sigma$  be the probability of generating  $\sigma$  through the origin-request process described earlier. Since  $\forall a_i, b_i > 0$  we have that  $(\sum_i a_i)/(\sum_i b_i)$  is a convex linear combination of the individual ratios  $a_i/b_i$ , focusing on the case  $\sqrt{\log_2(n+1)}/15 \geq 1$  for which  $\sqrt{\log_2(n+1)} + 3 + 2^{-n} < (5/4)\sqrt{\log_2(n+1)}$ :

$$\max_{\sigma:p_\sigma \neq 0} \frac{C_A(\sigma)}{C_O(\sigma)} \geq \frac{\sum_{\sigma:p_\sigma \neq 0} C_A(\sigma)p_\sigma}{\sum_{\sigma:p_\sigma \neq 0} C_O(\sigma)p_\sigma} > \frac{(n+1)\log_2(n+1)/12}{n(\sqrt{\log_2(n+1)} + 3 + 2^{-n})} > \frac{\sqrt{\log_2(n+1)}}{15}. \quad \blacktriangleleft$$

► **Remark.** Without discretized requests the term  $\sum_{\sigma:p_\sigma \neq 0} C_A(\sigma)p_\sigma$  in the theorem's proof would have been an integral, potentially ill-defined (for example, if ALG serviced requests for rational points in an interval with one server and for irrational points with another).

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