

## Lecture 16: Dynamic Algorithms for Maximum Matching Size

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## 1 Approximate Maximum Matching Size

Given a graph  $G = (V, E)$  on  $n$  vertices, the goal is to design efficient algorithms that can compute good approximations to (the size of) maximum matchings. Relevant works are in two different settings: (1) the sublinear setting where the graph is accessed via queries to its adjacency matrix/list; (2) the dynamic setting where the graph is updated by a sequence of edge insertions and deletions. Previous works and new results are summarized in Table 1.

Table 1: Previous works and new results

Reference	Approximation ratio	(Update) time	Setting	Output
[BGS18]	1/2	$O(\text{polylog}(n))$	Dynamic	Edges
[BLM20]	$1/2 + \Omega_\epsilon(1)$	$O(n^\epsilon)$	Dynamic	Edges
[Beh21]	$1/2 - \epsilon$	$\tilde{O}(n/\epsilon^2)$	Sublinear	Size
[BRRS23]	$1/2 + \Omega_\epsilon(1)$	$O(n^{1+\epsilon})$	Sublinear	Size
[Beh23]	$1/2 + \Omega(1)$	$O(\text{polylog}(n))$	Dynamic	Size
[Beh23]	$2/3 + \Omega(1)$	$O(\sqrt{n})$	Dynamic, Bipartite	Size

## 2 Warmup: Lazy Approach

Throughout, we assume  $\mu(G) = \Theta(n)$ . This is without loss of generality using known techniques. The basic idea is to transform sublinear algorithms into dynamic algorithms. Suppose there is a sublinear algorithm  $A$  for  $\alpha$ -approximate maximum matching size in  $T$  time. It implies a dynamic algorithm  $A'$  for  $(\alpha - \epsilon)$ -approximate maximum matching size in  $O(T/(\epsilon n))$  update time (amortized, but can be de-amortized to be worst-case) as follows.

1. Run  $A$  to approximate  $\mu(G)$ .
2. Do nothing for  $\epsilon n$  steps, and then repeat the first step.

Observe that we do need sublinear algorithms because  $T = \Theta(n^2)$  only yields  $\Theta(n/\epsilon)$  update time. Using this lazy approach, [Beh21] implies an (almost) 1/2-approximation in  $O(\text{polylog}(n))$  update time, and [BRRS23] recovers the result of [BLM20] (but only for the size not the edges). It also means that for our purpose, it is sufficient to design a sublinear algorithm for  $(1/2 + \Omega(1))$ -approximate maximum matching size in  $\tilde{O}(n)$  time.

**Lemma 1** (Lemma 4.1 of [Beh23]). *If there is a (randomized) semi-dynamic algorithm  $A$  with update time  $U(n)$  and query time  $Q(n, \epsilon)$  that outputs  $\tilde{\mu}$  satisfying  $\alpha\mu(G) - \epsilon n \leq$*

$\mathbb{E}[\tilde{\mu}] \leq \mu(G)$ , then there is a (randomized) fully-dynamic algorithm  $B$  with update time  $O((U(n) + Q(n, \epsilon^2)/n) \cdot \text{poly}(\log n, 1/\epsilon))$  that w.h.p. outputs  $\tilde{\mu}'$  satisfying  $(\alpha - \epsilon)\mu(G) \leq \tilde{\mu}' \leq \mu(G)$ .

### 3 Overview of [Beh21]

Given a permutation  $\pi$  over the edge set  $E$  of a graph  $G = (V, E)$ , let  $\mathbf{GMM}(G, \pi)$  denote the greedy maximal matching obtained by greedily adding edges of  $G$  in the order of  $E$  whenever possible. A crucial component used in the proof is a sublinear algorithm of [Beh21] for approximating  $\mathbf{GMM}(G, \pi)$  using query access to the adjacency matrix of  $G$ .

**Proposition 2** (Proposition 4.4 of [Beh23]). *There exists a randomized algorithm with query access to adjacency matrix that w.h.p. outputs  $\tilde{g}$  satisfying*

$$\mathbb{E}_\pi[|\mathbf{GMM}(G, \pi)|] - \epsilon n \leq \tilde{g} \leq \mathbb{E}_\pi[|\mathbf{GMM}(G, \pi)|]$$

in  $\tilde{O}(n/\epsilon^3)$  time.

At a very high level, the algorithm of [Beh21] is roughly as follows. Let  $M$  be a matching. Suppose we have oracle access that given  $v \in V$ , returns whether  $v$  is matched by  $M$  using  $Q$  queries. Then we can sample  $T$  vertices and output  $\tilde{g} = (\#\text{matched vertices}/T) \cdot (n/2)$ . Setting  $T = \Theta(1/\epsilon^2)$  approximates the size of  $M$  with  $\epsilon n$  additive error. The running time is  $QT$ . So the task is basically minimizing  $Q$  for a random vertex, which depends on  $M$ . It turns out that it is easy to check whether  $v$  is matched for a random greedy maximal matching.

### 4 Bipartite Graphs & Oblivious Adversaries

We first present a semi-dynamic algorithm with update time  $U(n) = O(\text{polylog}(n))$ , query time  $Q(n, \epsilon) = \tilde{O}(n/\epsilon^3)$ , and approximation parameter  $\alpha = 2 - \sqrt{2}$ , in the case of bipartite graphs and oblivious adversaries. Plugging into Lemma 1, we can get a  $(2 - \sqrt{2}) \approx 0.585$ -approximate dynamic algorithm. The starting point is the following query algorithm inspired by a random-order streaming algorithm of [KMM12].

1. Let  $M$  be a maximal matching that we maintain.
2. Let  $M' \subseteq M$  include each matching edge independently with probability  $p$ .
3. Let  $V' = V(M')$  and  $U = V \setminus V(M)$ .
4. Let  $H = G[V', U]$ .
5. Let  $g = \mathbb{E}_\pi[|\mathbf{GMM}(H, \pi)|]$ .
6. Return  $\tilde{\mu} = |M| + \max(0, g - |M'|)$ .

The update time is  $O(\text{polylog}(n))$  for maintaining  $M$  using [BGS18]. To see the query time, the computation of  $M', V', U$  can all be done in  $O(n)$  time given  $M$ . Although explicitly maintaining  $H$  is too time-consuming,  $g$  can be approximated up to  $\epsilon n$  additive error in  $\tilde{O}(n/\epsilon^3)$  time by Proposition 2, which is tolerable by Lemma 1. So it remains to show  $\tilde{\mu}$  is a  $(2 - \sqrt{2})$ -approximation of  $\mu(G)$  in expectation. To this end, we need the following result from [KMM12].

**Proposition 3.** *Let  $0 < p \leq 1$ . Let  $G = (A \sqcup B, E)$  be a bipartite graph. Let  $A' \subseteq A$  include each vertex independently with probability  $p$ . Let  $H$  be the induced bipartite subgraph on  $A', B$ . Then, for any permutation  $\pi$  over  $E$ , it holds that*

$$\mathbb{E}_{A'}[|\mathbf{GMM}(H, \pi)|] \geq p/(1+p) \cdot \mu(G).$$

*Proof of approximation ratio.* We first show the lower bound. Define  $F_L = G[V(M) \cap L, U \cap R]$ ,  $F_R = G[V(M) \cap R, U \cap L]$ ,  $H_L = G[V(M') \cap L, U \cap R]$ , and  $H_R = G[V(M') \cap R, U \cap L]$ . Fix a maximum matching  $M^*$  of  $G$ . Consider the symmetric difference  $M^* \oplus M$ . There are exactly  $|M^*| - |M| = \mu(G) - |M|$  disjoint augmenting paths with respect to  $M$  by the optimality of  $M^*$ . Furthermore, every augmenting path has length at least 3 since  $M$  is a maximal matching (length-1 augmenting path is simply an isolated edge). Any such augmenting path must start with an edge in  $F_L \cap M^*$  and end with an edge in  $F_R \cap M^*$  (or conversely). As all edges of  $M^*$  are disjoint, we can get  $\mu(F_L) \geq \mu(G) - |M|$  and  $\mu(F_R) \geq \mu(G) - |M|$ . Applying Proposition 3, we have that for any  $\pi$ ,

$$\mathbb{E}_{M'}[|\mathbf{GMM}(H_L, \pi)|] \geq p/(1+p) \cdot \mu(F_L) \geq p/(1+p) \cdot (\mu(G) - |M|),$$

and

$$\mathbb{E}_{M'}[|\mathbf{GMM}(H_R, \pi)|] \geq p/(1+p) \cdot \mu(F_R) \geq p/(1+p) \cdot (\mu(G) - |M|).$$

Since  $H_L$  and  $H_R$  are disjoint, for  $H = H_L \sqcup H_R$ , we can get

$$\mathbb{E}_{M'}[|\mathbf{GMM}(H, \pi)|] \geq 2p/(1+p) \cdot (\mu(G) - |M|).$$

Finally, taking expectation over  $\tilde{\mu}$ , we have

$$\begin{aligned} \mathbb{E}_{M'}[\tilde{\mu}] &= \mathbb{E}_{M'}[|M| + \max(0, g - |M'|)] \\ &\geq |M| + \max(0, \mathbb{E}_{M'}[g - |M'|]) \\ &\geq |M| + \max(0, 2p/(1+p) \cdot (\mu(G) - |M|) - p|M|) \\ &\geq (1 - 2p/(1+p) - p)|M| + 2p/(1+p) \cdot \mu(G). \end{aligned}$$

The lower bound follows by setting  $p = \sqrt{2} - 1$ .

Upper bound: to be continued. □

## References

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