Advanced Topics in Theoretical Computer Science



# Lecture 13

### **Matching in Insertion-deletion Streams**

#### **Edge-arrival insertion-only Model:**

- Stream consists of sequence of edges of a graph

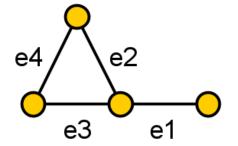
$$S = e_2 e_4 e_3 e_1$$

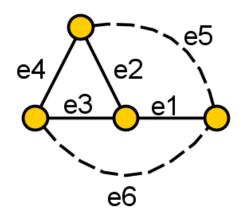


Insertion-deletion (or dynamic) Model: (since 2013)

- Sequence of edge insertions and deletions
- Only inserted edge can be deleted

$$S = e_4 e_3 e_5 \overline{e_5} e_2 e_6 \overline{e_2} e_2 e_1 \overline{e_6}$$





### Strategies:

- Greedy Algorithm for Matchings clearly fails...
- Even worse: Algorithms that *deterministically* store a set of edges may completely fail since stored edges could later be deleted...

### Instead:

- Randomization is crucial
- Linear sketches, in particular,  $l_0$ -sampling!

## Recap on $l_0$ -sampling

#### Turnstile stream:

- Stream describes vector  $f \in \{-m, ..., m\}^n$  by updates to its coordinates  $(m \in \mathbb{N})$
- Initially, f = (0, 0, ..., 0)
- Each item in the stream is an update (j, c), meaning  $f_j \leftarrow f_j + c \ (c \in \{-1, 1\})$

#### lowhari, Sağlam, Tardos, 2011]

There is a turnstile streaming algorithm with space  $O\left(\log^2 n \log \frac{1}{\delta}\right)$  that outputs a uniform random coordinate among the non-zero coordinates of f. It succeeds with proba.  $1 - \delta$ .

Example: f = (2, -4, 0, 0, 1, 0)

Then, the  $l_0$ -sampler outputs 1, 2, or 5 each with probability  $\frac{1}{3}$  (with success prob.  $1 - \delta$ ).

#### **Insertion-deletion Graph Streams are Turnstile Streams:**

- Insertion-deletion graph stream describes vector  $f \in \{0,1\}^{\binom{n}{2}}$  (or  $f \in [m]^{\binom{n}{2}}$  if multi-edges are allowed, for some integer  $m, [m] = \{0, 1, 2, ..., m\}$ )
- $l_0$ -sampling thus corresponds to sampling an edge from the input graph

### Other Applications of $l_0$ -sampling in Graph Streams:

By considering substreams of the input stream we can sample from...

- The set of edges incident to a specific vertex
- A random edge in a specific induced subgraph

## Error Probability in $l_0$ -sampling

lowhari, Sağlam, Tardos, 2011]

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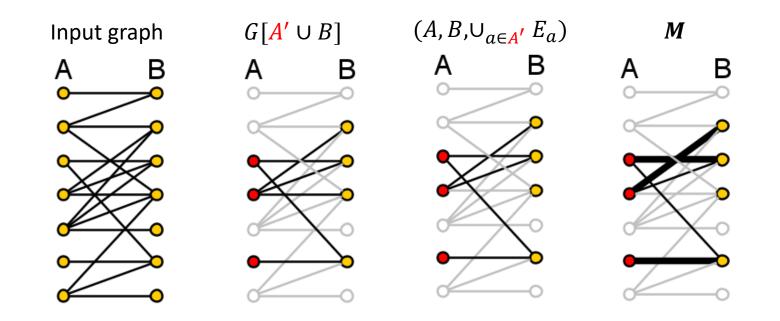
#### Example:

- Suppose we wish to run an  $l_0$ -sampler for each vertex  $v \in V$  in order to sample one incident edge to v. Further, our algorithm should be successful with probability  $\geq \frac{99}{100}$
- Observe that we are running  $n \ l_0$ -samplers
- We will choose  $\delta = \frac{1}{100n}$  in all  $l_0$ -samplers since: (union bound) Pr[at least one sampler errs]  $\leq n \cdot \Pr[\text{one sampler errs}] = n \cdot \delta = n \cdot \frac{1}{100n} = \frac{1}{100}$
- Each sampler thus requires space  $O(\log^2 n \log \frac{1}{\delta}) = O(\log^3 n)$ .
- $O(n \log^3 n)$  total space which is semi-streaming space!

## Offline Matching Algorithm for Bipartite Graphs

**Input:** Bipartite Graph G = (A, B, E) with |A| = |B| = n and integer parameter k

- 1. Sample uniform random subset  $A' \subseteq A$  of size k
- 2. For each  $a \in A'$ , select arbitrary min {deg(a), k} edges incident to a Let  $E_a$  denote this subset
- 3. Compute a maximum matching **M** in the graph  $(A, B, \bigcup_{a \in A'} E_a)$
- 4. Return M



**Lemma.** Suppose that G contains a perfect matching (all vertices are matched). Then the matching algorithm on previous slide has an approximation factor of  $\frac{k}{2n}$ .

#### Proof.

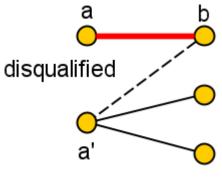
- Let  $M^*$  be a perfect matching in G
- Let  $A'_1 \subseteq A'$  be the subset of vertices a such that  $|E_a| = \deg(a)$ , let  $A'_2 = A' \setminus A'_1$
- First, suppose that  $|A'_1| \ge |A'_2|$  (which implies that  $k \ge |A'_1| \ge \frac{k}{2}$ ). Observe that for every  $a \in A'_1$ , the optimal edge incident to a in  $M^*$  is retained in  $E_a$ . We hence stored at least  $\frac{k}{2}$  edges from  $M^*$  and will thus be able to report a matching of at least that size.

- The approximation factor is thus 
$$\frac{k}{2}/n = \frac{k}{2n}$$

## Analysis (2)

- Next, suppose that  $|A'_2| > |A'_1|$  (which implies  $k \ge |A'_2| \ge \frac{k}{2}$ ). Observe that for each  $a \in A'_2$ , we have  $|E_a| = k$ . Consider the graph  $H = (A'_2, B, \bigcup_{a \in A'_2} E_a)$ . Then the degree of every  $A'_2$  vertex is k.
- We will show that H contains a matching of size  $|A'_2|$ :
- Consider the matching M' produced by running Greedy on an arbitrary sequence of the edges of H. Suppose an edge ab is inserted into M'. Then, for every  $a' \neq a$ , this disqualifies at most one edge incident to a' (i.e., the potential edge a'b) from being added to M'.
- Since the degree of every  $a' \in A'_2$  is k and  $|A'_2| \leq k$ , we are able to insert an edge incident to every vertex in  $A'_2$ .





## Turning the Algorithm into a Streaming Algorithm

**Input:** Bipartite Graph G = (A, B, E) with |A| = |B| = n and integer parameter k

- 1. Sample uniform random subset  $A' \subseteq A$  of size k
- 2. For each  $a \in A'$ , select arbitrary min  $\{ deg(a), k \}$  edges incident to aLet  $E_a$  denote this subset
- 3. Compute a maximum matching **M** in the graph  $(A, B, \bigcup_{a \in A'} E_a)$
- 4. Return *M*
- Step 1 before processing the stream (pre-processing step)
- Step 3 after processing the stream (post-processing step)

#### How can we implement step 2 in the insertion-deletion streaming model?

**Task:** For a given vertex  $a \in A'$ , compute arbitrary min{deg(a), k} edges while processing the stream

 $l_0$ -sampling: (on substream of edges incident to a)

- Returns a uniform random edge incident to *a*
- Strategy: Run enough  $l_0$ -samplers to yield min $\{\deg(a), k\}$  different edges with very high probability
- *Exercise*: If we run 10  $k \log n l_0$ -samplers, then the probability that we do not obtain min{deg(a), k} different edges is at most  $\frac{1}{n^5}$ .

## Final Insertion-deletion Streaming Algorithm

#### **Insertion-deletion Streaming Algorithm:**

**Input:** Bipartite Graph G = (A, B, E) with |A| = |B| = n and integer parameter k

- 1. Sample uniform random subset  $A' \subseteq A$  of size k
- 2. While processing the stream: For each  $a \in A'$ , maintain 10  $k \log n l_0$ -samplers on substream of edges incident to a
- 3. Compute a maximum matching *M* among all the edges sampled
- 4. Return *M*

**Space Complexity:** # samplers × space complexity of samplers

$$O\left(k \cdot 10 \ k \log n \cdot \log^2 n \log \frac{1}{\delta}\right) = O\left(k^2 \log^3 n \cdot \log \frac{1}{\delta}\right)$$

## Error Probability and how to chose $\delta$ ?

#### **Error probability:**

- Error is introduced by the l<sub>0</sub>-samplers and the possibility that sampling 10 k log n uniform random edges incident to a vertex a does not yield min{deg(a), k} different edges
- By the union bound, we can sum up all these error probabilities to bound the total error probability of the algorithm
- *k* times: The probability that  $10 k \log n$  uniform random edges do not yield enough different edges for a specific vertex is at most  $1/n^5$ .
- 10  $k^2 \log n$  times:  $l_0$ -sampler fails with probability  $\delta$ .

Total error probability: (union bound)

$$k \cdot \frac{1}{n^5} + 10k^2 \log n \cdot \delta \leq \frac{1}{n^4} + 10n^3 \delta$$
  
We select  $\delta = \frac{1}{10n^5}$ . Then total error at most  $\frac{1}{n}$ .

**Theorem.** There is an insertion-deletion streaming algorithm for Maximum Matching with approximation ratio  $\frac{k}{2n}$  and space  $O(k^2 \log^4 n)$  that fails with probability at most  $\frac{1}{n}$ .

**Corollary. (by setting**  $k = \sqrt{n}$ ) There is an insertion-deletion semistreaming algorithm with approximation ratio  $\frac{1}{2\sqrt{n}}$  that fails with probability at most  $\frac{1}{n}$ .

### Summary

- The algorithm presented here is due to [Konrad 2015]
- It is known today that the best insertion-deletion semi-streaming algorithm for maximum matching has approximation ratio  $\frac{1}{\Theta(n^{\frac{1}{3}})}$ . The algorithm is due to [Assadi et al. 2016] and the optimality proof is due to [Dark, Konrad 2020]

[Konrad, 2015] Christian Konrad: Maximum Matching in Turnstile Streams. ESA 2015: 840-852

[Assadi et al. 2016] Sepehr Assadi, Sanjeev Khanna, Yang Li, Grigory Yaroslavtsev: Maximum Matchings in Dynamic Graph Streams and the Simultaneous Communication Model. SODA 2016: 1345-1364

[Dark, Konrad 2020] Jacques Dark, Christian Konrad: Optimal Lower Bounds for Matching and Vertex Cover in Dynamic Graph Streams. Computational Complexity Conference 2020: 30:1-30:14