## Semi-streaming Model: Graph Matchings

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## Motivation

## Definition (Streaming Model)

Streaming is a model of computation on massive data sets that arrive sequentially in an arbitrary order. We have only poly-log(m) amount of storage space, m being the number of elements in the stream.

Most (massive) graph problems hadn't been explored. Reasons:

- very tight space constraint
- data access in sequential (potentially adversarial) order

The paper by Feigenbaum et al. [1] first explores the semi-streaming model with a more lenient space constraint and gives algorithms for a few graph problems under this model.

## Definition (Semi-Streaming Model)

For an input graph G(V, E), we have n poly-log(n) space, with n = |V|. Hence, we can store G(V) but not G(E).

## Semi-Streaming Model

## Definition (Graph Stream)

A graph stream  $\sigma(G)$  is a sequence of edges  $e_{i_1}, e_{i_2}, \ldots, e_{i_m}$  that appear one at a time as input to an algorithm, where  $e_{ij} \in G(E), m = |E|$  and  $i_1, i_2, \ldots, i_m$  is an arbitrary permutation of [m].

## Definition (Semi-Streaming Graph Algorithm)

It is an algorithm that computes a specific property of a graph G given  $\sigma(G)$ . In doing so, it uses S(n,m) space, T(n.m) time to process each edge of the stream and P(n,m) sequential passes of the streams. It is required that S(n,m) be O(n poly-log(n)) and P(n,m) be O(poly-log(n)).

We will cover a  $\left(\frac{2}{3}-\epsilon\right)$ -algorithm for unweighted bipartite matching and a  $\left(\frac{1}{6}\right)$ -algorithm for weighted matching under this model from the paper by Feigenbaum et al.

## **Graph Bipartiteness**

## Algorithm 1

This algorithm verifies graph bipartiteness during edge streaming. As edges stream in, we use a disjoint set data structure to maintain connected components of the graph so far. We also associate a sign with each vertex such that no edge connects 2 vertices of the same sign. If this condition ever fails even on flipping the sign of a vertex and the vertices in its connected component, the graph is non-bipartite.

## Definition (length-3 augmenting path)

Given a matching M in a bipartite graph  $G = (L \cup R, E)$ , a length-3 augmenting path for an edge  $(u, v) \in M$ , where  $u \in L$  and  $v \in R$ , is a quadruple  $(w_l, u, v, w_r)$  such that  $(w_l, u), (v, w_r) \in E$  and  $w_l$  and  $w_r$  are free vertices. We call  $(u, w_l)$  the **left wing**,  $(v, w_r)$  the **right wing**, and  $w_l$ and  $w_r$  the wing-tips.

## Simultaneously augmentable length-3 augmenting paths

## Definition (Simultaneously augmentable length-3 augmenting paths)

A set of simultaneously augmentable length-3 augmenting paths is a set of length-3 augmenting paths that are vertex disjoint.

## Algorithm 2

### Input:

- Bipartite graph  $G = (L \cup R, E)$
- Matching M for G
- ullet Parameter  $0 < \delta < 1$

## **Output:**

 $\bullet$  simultaneously augmentable length-3 augmenting paths for G under M

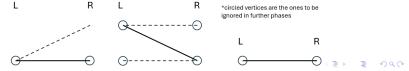
## Algorithm 2

#### Procedure

- In one pass, find a maximal set of disjoint left wings LW. Terminate if the number of left wings  $\leq \delta |M|$ .
- ② In 2nd pass, for edges in M with left wings, find a maximal set of disjoint right wings RW.
- In 3rd pass, identify vertices that are:
  - Endpoints of matched edges with a left wing in LW.
  - Wing-tips of matched edges with both wings (1 each in LW and RW).
  - 3 Endpoints of matched edges that are no longer 3-augmentable.

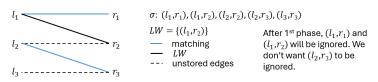
Ignore edges incident on any 1 of these vertices in further passes.

4 Store length-3 augmenting paths found. Repeat steps 1-4.



## Algorithm 2: Execution

- In step 3 of the algorithm, we're basically identifying the vertices that can no longer become a part of length-3 augmenting paths in further passes.
- In order to correctly identify vertices that fall into the 3rd category, we associate an indicator variable with each edge of the matching which indicates if a left wing was detected for that edge in the 1st pass. If the variable stores false, it belongs to the 3rd category.



## Bipartite Matching Algorithm

## Algorithm 3

## Input:

- Bipartite graph  $G = (L \cup R, E)$
- Parameter  $0 < \epsilon < \frac{1}{3}$

### Output:

• a  $(\frac{2}{3} - \epsilon)$ -factor matching on G

#### **Procedure:**

- ullet Find a maximal matching M and bipartition of G in a single pass.
- For every  $k=1,2,\ldots,\left\lceil\frac{\log(6\epsilon)}{\log\left(\frac{8}{9}\right)}\right\rceil$ :
  - Execute Algorithm 2 with G, M, and  $\delta = \frac{\epsilon}{2-3\epsilon}$ .
  - For each edge  $e = (u, v) \in M$  where an augmenting path  $(w_l, u, v, w_r)$  is found:
    - Remove *e* from *M*.
    - Append  $(u, w_l)$  and  $(w_r, v)$  to M.

## Lemma 1

#### Lemma 1

The size of a maximal set of simultaneously augmentable length-3 augmenting paths is at least 1/3 of the size of a maximum set of simultaneously augmentable length-3 augmenting paths.

#### Proof.

- $AP_{max} \triangleq A$  maximum set of simultaneously augmentable length-3 augmenting paths.
- $AP \triangleq A$  maximal set of simultaneously augmentable length-3 augmenting paths.
- $M \triangleq Matching considered$

Each path in the maximal set destroys at most 3 paths that  $AP_{max}$  might have used.

## Proof of Lemma 1

## Proof.

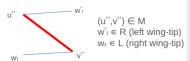
Consider  $(w_i, u, v, w_r) \notin AP_{max}$  but  $\in AP$ 



 $(u,v) \in M$   $w_i \in R$  (left wing-tip)  $w_r \in L$  (right wing-tip)

- **1.**  $(w_i, u', v', w'_r) \in AP_{max} \text{ but } \notin AP$
- $(u',v') \in M$   $w_i \in R$  (left wing-tip)  $w'_r \in L$  (right wing-tip)

**2.**  $(W'_1, U'', V'', W_r) \in AP_{max} \text{ but } \notin AP$ 



3.  $(w''_i, u, v, w''_r) \in AP_{max} \text{ but } \notin AP$ 



#### Proof.

From the above figure, we can see that a path in the maximal set destroyed 3 paths that  $AP_{max}$  has.

These are:

- Path involving left wing-tip
- Path involving right wing-tip
- Path involving matched-edge used

Therefore, a maximal set has a size of at least 1/3 of the maximum set.

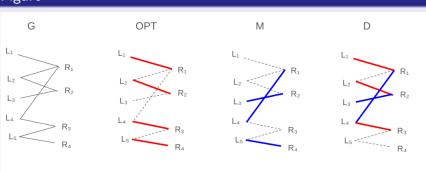


#### Lemma 2

#### Lemma 2

Let X be the maximum-sized set of simultaneously augmentable length-3 augmenting paths for a maximal matching M. Let  $\alpha = \frac{|X|}{|M|}$  and  $\overrightarrow{OPT}$  a maximum matching. Then  $|M|(1+\alpha) \ge 2/3|OPT|$ 

## Figure



Component 1: {(L<sub>1</sub>,R<sub>1</sub>), (L<sub>4</sub>,R<sub>1</sub>), (L<sub>4</sub>,R<sub>3</sub>)} Component 2: {(L2,R2), (L3,R2)}

## Proof of Lemma 2

### Proof.

- $D \triangleq \text{symmetric difference } OPT \nabla M$
- $y \triangleq$  no. of edges which common to both M and OPT
- $c_1 \triangleq$  no. of components in D with one edge from M and two edges from OPT
- $c_2 \triangleq (\text{total no. of components in } D)$   $c_1$
- $|e_{M_i}| \triangleq$  no. of edges that come from M in the  $i^{th}$  component of D
- $|e_{OPT_i}| \triangleq$  no. of edges that come from OPT in the  $i^{th}$  component of D

In each connected component, the following hold true:

 $\rightarrow |e_{OPT_i}| \ge |e_{M_i}| \ \forall i$  (if not, defn. of OPT being maximum would not hold true, since we could replace edges in  $e_{OPT_i}$  with those in  $e_{M_i}$ )

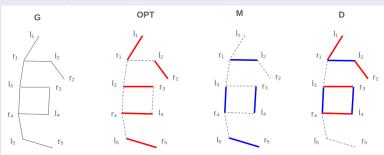


#### Proof.

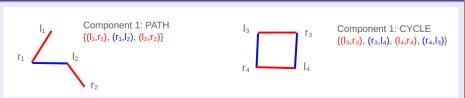
 $\rightarrow$  There is at most one more edge from *OPT* than there is from *M*.

$$|e_{OPT_i}| \le |e_{M_i}| + 1 \forall i \tag{1}$$

(Since OPT is a matching, the edges would be disjoint and hence for the component to be connected, we need edges from matching M to connect them.)



#### Proof.



(Therefore each vertex in the component will have degree either 2 or 1. So, the component is either a path or a cycle.)

- $\rightarrow$  No connected component consists of only a single edge that came from OPT (because M is maximal)
- $\rightarrow$  (Using defn. of X)

$$c_1 \leq |X| \tag{2}$$



## Proof.

#### Claim

In all components other than those in  $c_1$ , the ratio of  $e_{M_i}$  to  $e_{OPT_i}$  is at least 2 : 3

#### Proof of Claim

Case I: Components with  $|e_{M_i}| = 1$  and  $|e_{OPT_i}| = 1$ . Ratio is 1:1 in this case and hence claim holds true.

Case II:  $|e_{M_i}| \geq 1$ 

Using equation (1), we get

$$\frac{|e_{OPT_i}|}{|e_{M_i}|} \le 1 + \frac{1}{|e_{M_i}|} \le \frac{3}{2} \tag{3}$$

$$\implies \frac{|e_{M_i}|}{|e_{OPT_i}|} \ge \frac{2}{3} \tag{4}$$

#### Proof.

We can write the following equations

$$|M| = c_1 + y + \sum_{i=1}^{c_2} |e_{M_i}| \tag{5}$$

$$|OPT| = 2c_1 + y + \sum_{i=1}^{c_2} |e_{OPT_i}|$$
 (6)

Using equation (2) and equation (4),

$$|M| + |X| \ge 2c_1 + y + \sum_{i=1}^{c_2} |e_{M_i}| \tag{7}$$

$$\geq 2c_1 + y + \sum_{i=1}^{c_2} \frac{2}{3} |e_{OPT_i}| \tag{8}$$

#### Proof.

$$|M| + |X| \ge \frac{4}{3}c_1 + \frac{2}{3}y + \sum_{i=1}^{c_2} \frac{2}{3}|e_{OPT_i}|$$
 (9)

$$=\frac{2}{3}(2c_1+y+\sum_{i=1}^{c_2}|e_{OPT_i}|) \tag{10}$$

$$=\frac{2}{3}|OPT|\tag{11}$$

Substituting  $|X| = \alpha |M|$ , we get

$$|M| + \alpha |M| \ge \frac{2}{3}|OPT| \tag{12}$$

$$|M|(1+\alpha) \ge \frac{2}{3}|OPT| \tag{13}$$

Semi-streaming Model

## Lemma 3

#### Lemma 3

Algorithm 2 finds  $\frac{\alpha|M|-2\delta|M|}{3}$  simultaneously augmentable length-3 augmenting paths in  $3/\delta$  passes.

#### Proof.

- $L(M) \triangleq$  set of the end vertices in M that are in L
- $V_L(M) \triangleq \{v \in R | v \text{ is free w.r.t. } M \text{ and } \exists u \in L(M)s.t.(u,v) \in E \}$
- phase  $\triangleq$  one repetition of step 1-4 in Algorithm 2

Algo 2 terminates only when no. of left wings found in the  $1^{st}$  pass of a phase is  $\leq \delta M$  as defined in Algo 2.

The no. of phases is at most  $1/\delta$  because at least  $\delta |M|$  edges in M are removed at each phase.

Since each phase has 3 passes, we can say that no. of passes is at most  $3/\delta$ 

#### Proof.

- $G' \triangleq$  graph restricted to the remaining vertices in L(M) and  $V_I(M)$ when the Algo. 2 terminates.
- $G'' \triangleq G \setminus G'$
- $X \triangleq A$  maximum sized set of simultaneously augmentable length-3 augmenting paths for the given matching M in G

#### Claim

A maximum set of simultaneously augmentable length-3 augmenting paths in G'' would have a size at least  $\alpha |M| - 2\delta |M|$ .

#### Proof of claim

- $\rightarrow$  When Algo. 2 terminates, the no. of left wings found is at least  $\delta |M|$ .
- $\rightarrow$  This set of left-wings form a maximal matching in G'.

#### Proof.

#### Proof of claim contd.

- ightarrow Hence there are fewer than  $2\delta |M|$  disjoint left wings that could have been found at this phase.
- $\rightarrow$  Consequently, there are fewer than  $2\delta |M|$  simultaneously augmentable length-3 augmenting paths in G'.
- $\rightarrow$  A maximum set of simultaneously augmentable length-3 augmenting paths in G'' would have a size at least  $\alpha |M| 2\delta |M|$ .

Note that the set of length-3 augmenting paths found by Algo. 2 form a maximal set w.r.t G''. (Since Algo 2. uses disjoint wing sets) Using the claim proved above and Lemma 1, we get that the size of a maximal set is at least  $\frac{\alpha|M|-2\delta|M|}{2}$ .

## Bipartite Matching: Main Theorem

#### Theorem 1

For any  $0<\epsilon<\frac{1}{3}$  and a bipartite graph, Algorithm 3 finds a  $(\frac{2}{3}-\epsilon)$ -approximation maximum matching in  $O((\log 1/\epsilon)/\epsilon)$  passes and  $O(n\log n)$  space. Each edge is processed in O(1) time except the first pass, where we spend O(n) time per edge in find and union operations of sets.

#### Proof.

- ullet OPT  $\triangleq$  size of maximum matching
- $M_i \triangleq \text{matching } M \text{ of the algorithm after } i \text{th iteration}$
- $X_i \triangleq \text{maximum-sized}$  set of simultaneously augmentable length-3 augmenting paths for  $M_i$
- $Y_i \triangleq$  set of simultaneously augmentable length-3 augmenting paths found by *Algorithm 2* for  $M_i$



## Proof of Theorem 1

### Proof.

- $\alpha_i \triangleq |X_i|/|M_i|$
- $s_i \triangleq |M_i|/\mathsf{OPT}$

Case I:  $\exists i : \alpha_i \leq \frac{3\epsilon}{2-3\epsilon}$ . Using Lemma 2,

$$|M_i| \ge \frac{2}{3(1+\alpha_i)} \mathsf{OPT} \ge \frac{2}{3\frac{2}{2-3\epsilon}} \mathsf{OPT} \ge \left(\frac{2}{3}-\epsilon\right) \mathsf{OPT}$$
 (14)

Since 
$$\forall j \mid M_{j+1} \mid \geq \mid M_j \mid$$
,  $\Longrightarrow \mid M_k \mid \geq \mid M_i \mid \geq \left(\frac{2}{3} - \epsilon\right) \text{ OPT}$  (15)

Case II:  $\alpha_i > \frac{3\epsilon}{2-3\epsilon} \ \forall i \in [k]$ . Using Lemma 2,

$$\alpha_i |M_i| \ge \frac{2}{3} \mathsf{OPT} - |M_i| \implies \alpha_i s_i \ge \frac{2}{3} - s_i$$
 (16)



## Proof of Theorem 1 contd.

#### Proof.

Also,  $\delta = \frac{\epsilon}{2-3\epsilon} \leq \frac{\alpha_i}{3} \ \forall i$ . By Lemma 3,

$$Y_i \ge \frac{\alpha_i - 2\delta}{3} |M_i| \ge \frac{\alpha_i}{9} |M_i| \tag{17}$$

$$|M_{i+1}| = |M_i| + |Y_i| \ge \left(1 + \frac{\alpha_i}{9}\right) |M_i|$$
 (18)

$$\implies s_{i+1} \ge s_i + \frac{\alpha_i s_i}{9} \ge s_i + \frac{1}{9} \left( \frac{2}{3} - s_i \right) = \frac{8}{9} s_i + \frac{2}{7}$$
 (19)

As  $M_0$  is a maximal matching,  $s_0 \ge 1/2$ . Solving the above recurrence gives  $s_k \ge \frac{2}{3} - \frac{1}{6}(\frac{8}{9})^k \ge \frac{2}{3} - \epsilon$ 

Number of passes 
$$\leq k \cdot \frac{3}{\delta} = \left\lceil \log_{\frac{9}{8}} \frac{1}{6\epsilon} \right\rceil \frac{6 - 9\epsilon}{\epsilon} = O\left(\frac{\log(1/\epsilon)}{\epsilon}\right)$$
 (20)



## Weighted Matching Algorithm

## Algorithm 4

We maintain a matching M at all times. When we see a new edge e, we compare w(e) with w(C), the sum of the weights of the edges of  $C = \{e' \in M \mid e' \text{ and } e \text{ share an end point}\}.$ 

- If w(e) > 2w(C), we update  $M \leftarrow M \cup \{e\} \setminus C$ .
- If  $w(e) \leq 2w(C)$ , we ignore e.

#### Theorem 2

Algorithm 4 gives  $\frac{1}{6}$ -factor weighted matching in 1 pass,  $\mathcal{O}(n \log n)$  space.

#### **Notations**

- $w(S) \triangleq \sum_{e \in S} w(e)$ , where  $S \subseteq E$
- born edge  $\triangleq$  an edge that is ever part of M



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## Proof of Theorem 2

#### **Notations**

- killed edge  $\triangleq$  a born edge murdered (i.e., removed from M) by a newer, heavier edge.
- survivor edge  $\triangleq$  a born edge that is never killed
- $S \triangleq$  set of survivor edges
- $T(e) \triangleq C_1 \cup C_2 \cup ... \ \forall e \in S$ , where
  - $C_0 = \{e\}$
  - $C_1 = \{ \text{the edges murdered by e} \}$
  - $C_i = \bigcup_{e' \in C_{i-1}} \{ \text{the edges murdered by e'} \}$
- OPT  $\triangleq$  a maximum weighted matching on G
- $M_{all} \triangleq \bigcup_{e \in S} (T(e) \cup e)$

#### Claim

$$w(T(e)) \le w(e) \tag{21}$$

## Proof of Theorem 2 contd.

#### Proof.

For each murdering edge e, w(e) is at least twice the cost of murdered edges, and an edge has at most one murderer. Hence

$$w(C_i) \ge 2w(C_{i+1}) \ \forall i \tag{22}$$

$$\implies 2w(T(e)) = \sum_{i \ge 1} 2w(C_i) \le \sum_{i \ge 0} w(C_i) = w(T(e)) + w(e)$$
 (23)

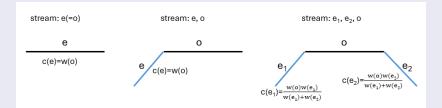
### Proof of Theorem 2.

We charge costs of edges in OPT to edges in  $M_{all}$ . An edge  $e \in M_{all}$  is **accountable** to  $o \in \mathsf{OPT}$  if e = o or if o wasn't born because e was in M when o arrived.

## Proof of Theorem 2 contd.

#### Proof.

- Case I: If only one edge e is accountable to o (implying  $w(o) \le 2w(e)$ ), we assign charge  $c_o(e) \triangleq w(o) \le 2w(e)$ .
- Case II: If two edges  $e_1$  and  $e_2$  are accountable to o (implying  $w(o) \le 2(w(e_1) + w(e_2))$ ), we assign charges  $c_o(e_1) \triangleq \frac{w(o)w(e_1)}{w(e_1) + w(e_2)} \le 2w(e_1)$  and  $c_o(e_2) \triangleq \frac{w(o)w(e_2)}{w(e_1) + w(e_2)} \le 2w(e_2)$



## Proof of Theorem 2 contd.

#### Proof.

We redistribute charges as follows: for distinct  $u_1, u_2, u_3$ , if an edge  $e = (u_1, v)$  is charged by  $o = (u_2, v)$ , and then killed by another edge  $e' = (u_3, v)$ , the charge is transferred from e to e'. Since  $w(e') \geq 2w(e) \geq c_o(e)$ , hence  $c_o(e') = c_o(e) \leq 2w(e')$ . This ensures that each killed edge is charged by at most one edge in OPT.

stream: 
$$(u_1,v), (u_2,v), o' = (u_4,u_1), (u_3,v)$$

$$\underbrace{u_1} v \underbrace{u_2,v} v \underbrace{u_1} \underbrace{u_2,v} \underbrace{u_1} \underbrace{v} \underbrace{u_1} \underbrace{u_2,v} \underbrace{u_2,v} \underbrace{u_1} \underbrace{v} \underbrace{u_2,v} \underbrace{u_1} \underbrace{v} \underbrace{u_2,v} \underbrace{u_1} \underbrace{v} \underbrace{u_2,v} \underbrace{u_1} \underbrace{v} \underbrace{u_2,v} \underbrace{u_2,v} \underbrace{u_1} \underbrace{v} \underbrace{u_2,v} \underbrace{u_2,v} \underbrace{u_2,v} \underbrace{u_1} \underbrace{u_2,v} \underbrace{u_2,v} \underbrace{u_2,v} \underbrace{u_1} \underbrace{u_2,v} \underbrace{u_2,v} \underbrace{u_2,v} \underbrace{u_2,v} \underbrace{u_1} \underbrace{u_2,v} \underbrace{u_2,v} \underbrace{u_2,v} \underbrace{u_2,v} \underbrace{u_1,v} \underbrace{u_2,v} \underbrace{$$

$$w(\mathsf{OPT}) \le \sum_{e \in S} (2w(T(e)) + 4w(e)) \le 6w(S) \text{ (Using claim)} \tag{24}$$

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## Lower Bound on s - t Connectivity

#### Lemma 4

Testing for s-t connectivity in a directed graph G=(V,E) requires  $\Omega(m)$  bits of space, where |E|=m.

#### Proof.

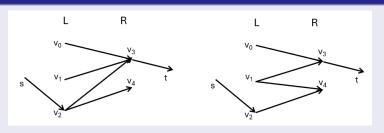
- $\mathcal{G} \triangleq$  set of all directed bipartite graphs
- $\mathcal{F} \triangleq \{G = (L \cup R \cup \{s,t\}, E) \in \mathcal{G} : (u,v) \in E \implies (u \in L, v \in R) \lor ((u,v) = (s,l)) \lor ((u,v) = (r,t))\}$ , where l and r are fixed vertices in L and R respectively.

Suppose the stream provides all L-R edges first, then (s,l) and (r,t). At the point when all edges from L to R have appeared, any correct algorithm will have different memory configuration for each graph in  $\mathcal{F}$  since there are continuations that could lead to different answers for any two graphs in  $\mathcal{F}$ .

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## Proof of Lemma 4

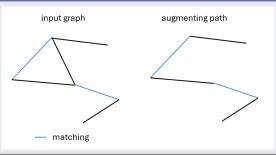
#### Proof.



To have a unique memory configuration for each graph, we need  $\Omega(\log_2 |\mathcal{F}|)$  bits. Even for the restricted case when |L| = |R| = (n/2) - 1, the number of such graphs is  $((2^{(n/2)-1})^{(n/2)-1} = 2^{\Omega(n^2)})$ , which gives  $\Omega(\log_2 |\mathcal{F}|)$  to be  $\Omega(n^2) = \Omega(m)$ .

## Verifying Maximum Matching: Main Theorem

## Definition (Augmenting Path)



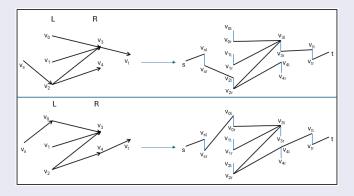
#### Theorem 3

Given a bipartite graph  $G = (L \cup R, E)$ , determining existence of an augmenting path from  $s \in R$  to  $t \in L$  requires  $\Omega(m)$  storage.

## Proof of Theorem 3

#### Proof.

Given  $G = (L \cup R \cup \{v_s, v_t\}) \in \mathcal{F}$ , we construct an undirected bipartite graph G' with nodes s, t such that  $\exists$  an augmenting path from s to t in G' iff  $\exists$  a directed path from  $v_s$  to  $v_t$  in G.



### References



J. FEIGENBAUM, S. KANNAN, A. McGregor, S. Suri and J. Zhang, "On graph problems in a semi-streaming model," *Theoretical Computer Science, Volume 348, Issues 2–3, Pages 207-216*, 8 December 2005.

# Thank You