

A Simple, Space-Efficient, Streaming Algorithm for Matchings in Low Arboricity Graphs

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Joint work with Andrew McGregor

Streaming Model: Insert-Only

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In an **insert-only** stream we only insert edges into the graph.

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• 3

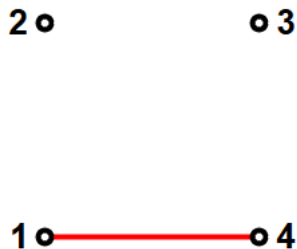
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• 4

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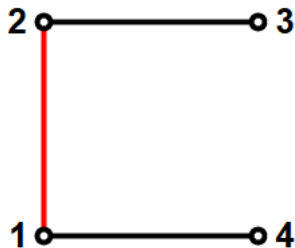


(1, 4), (2, 3)

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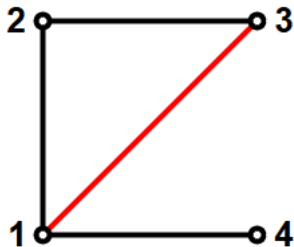


$(1, 4), (2, 3), (1, 2)$

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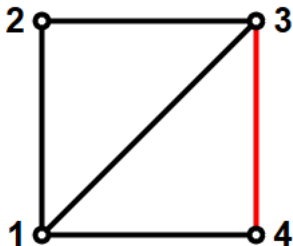


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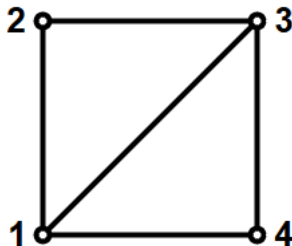


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Streaming Model: Insert-Only

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Streaming Model

Standard model:

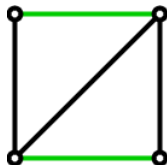
- One pass
- Edges arrive in arbitrary order

Objectives:

- Compute some property of the graph defined by the stream
- Minimize amount of space
- Fast update time is generally encouraged

Approximating Size of Maximum Matching

Matching is a set of edges that don't share endpoints.



In insert-only stream can easily obtain *maximal* matching, which is a 2-approximation of *maximum* matching.

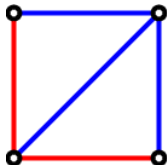
Maximum matching can be as large as $n/2$.

By approximating the **size** of the matching without finding the matching itself, we can use smaller space.

Low Arboricity Graphs

We concentrate on the class of graphs of arboricity α .

Arboricity is the minimum number of forests into which the edges of the graph can be partitioned.



Property: In a graph with arboricity α , every induced subgraph on r vertices has at most αr edges.

Low arboricity graphs are sparse.

Planar graphs have arboricity at most 3.

Results

Space	Approx Factor	Work
$O(\alpha\epsilon^{-2}n^{2/3} \text{polylog } n)$	$(5\alpha + 9)(1 + \epsilon)$	Esfandiari et al. SODA '15
$O(\alpha\epsilon^{-2}n^{2/3} \text{polylog } n)$	$(\alpha + 2)(1 + \epsilon)$	McGregor et al. APPROX '16
$O(\alpha\epsilon^{-3} \log^2 n)$	$(22.5\alpha + 6)(1 + \epsilon)$	Cormode et al. ESA '17
$O(\epsilon^{-2} \log n)$	$(\alpha + 2)(1 + \epsilon)$	here

Space is specified in words.

An edge or a counter can be stored in one word of space.

Approach

Our approach has 2 parts:

- Define E^* that is an $\alpha + 2$ approximation of $\text{match}(G)$
- Algorithm: $1 + \epsilon$ approximation of E^* in insert-only stream

Defining E^*

Let E_α be the set of edges uv where the number of edges incident to u or v that appear in the stream after uv are both at most α .

Lemma

$$\text{match}(G) \leq |E_\alpha| \leq (\alpha + 2) \text{match}(G)$$

Defining E^*

Let E_α be the set of edges uv where the number of edges incident to u or v that appear in the stream after uv are both at most α .

Lemma

$$\text{match}(G) \leq |E_\alpha| \leq (\alpha + 2) \text{match}(G)$$

Let G_t be the graph defined by the first t edges in the stream.

Let E_α^t be $E_\alpha(G_t)$. Then

$$\text{match}(G_t) \leq |E_\alpha^t| \leq (\alpha + 2) \text{match}(G_t)$$

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Let $E^* = \max_t |E_\alpha^t|$.

Since $\text{match}(G_t)$ is non-decreasing function of t ,

$$\text{match}(G) \leq E^* \leq (\alpha + 2) \text{match}(G)$$

Defining E^*

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Upper bound

Fun fact (from Edmond's thm): for any fractional matching with $x_e \leq \lambda$ for all e ,

$$\sum_e x_e \leq (1 + \lambda) \text{match}(G)$$

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Let

$$x_e = \begin{cases} 1/(1 + \alpha) & \text{if } e \in E_\alpha \\ 0 & \text{otherwise} \end{cases}$$

$\{x_e\}_{e \in E}$ is a fractional matching with max weight $1/(\alpha + 1)$. Thus,

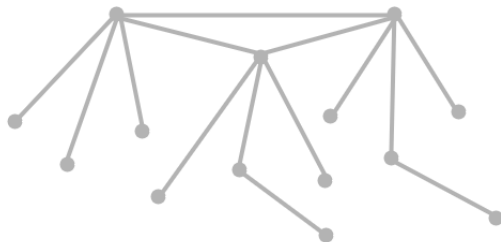
$$\frac{|E_\alpha|}{\alpha + 1} = \sum_e x_e \leq \frac{\alpha + 2}{\alpha + 1} \cdot \text{match}(G)$$

Defining E^*

Lemma

$$\text{match}(G) \leq |E_\alpha| \leq (\alpha + 2) \text{match}(G)$$

Lower bound

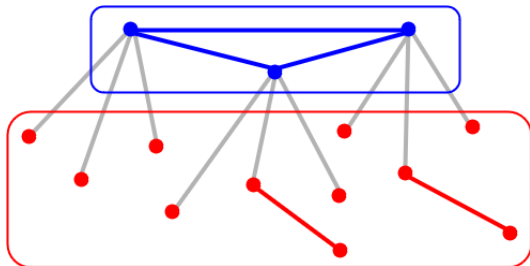


Defining E^*

Lemma

$$\text{match}(G) \leq |E_\alpha| \leq (\alpha + 2) \text{match}(G)$$

Lower bound



V^H = heavy vertices of degree $\geq \alpha + 1$

E^H = heavy edges with 2 heavy endpoints

V^L = light vertices

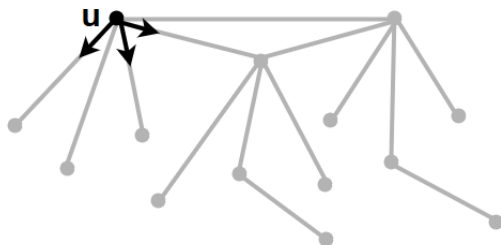
E^L = light edges

Defining E^*

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Lower bound



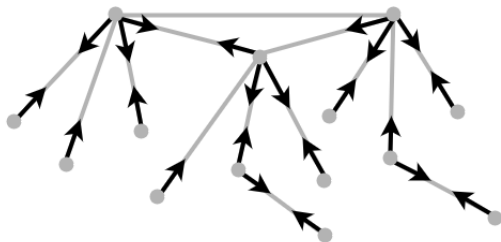
$B_u =$ last $\alpha + 1$ edges on u in the stream

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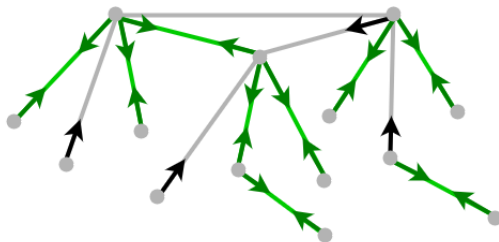
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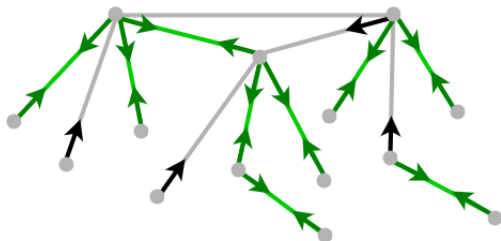
Edge uv is **good** if $uv \in B_u$ and $uv \in B_v$

Defining E^*

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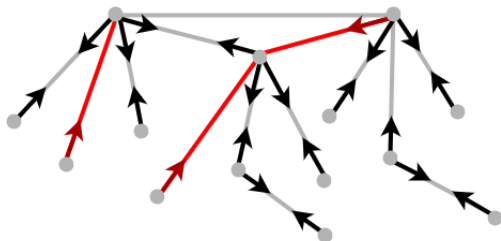
g_i is the number of **good** edges with i heavy endpoints

Defining E^*

Lemma

$$\text{match}(G) \leq |E_\alpha| \leq (\alpha + 2) \text{match}(G)$$

Lower bound



Edge uv is **good** if $uv \in B_u$ and $uv \in B_v$

g_i is the number of **good** edges with i heavy endpoints

Edge uv is **wasted** if $uv \in B_u$ or $uv \in B_v$, but not both

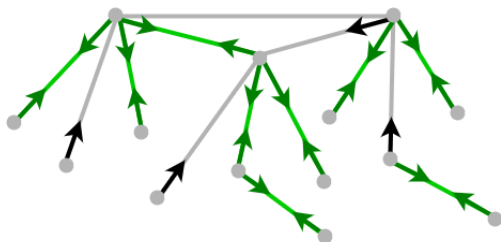
w_2 is the number of **wasted** edges with 2 heavy endpoints

Defining E^*

Lemma

$$\text{match}(G) \leq |E_\alpha| \leq (\alpha + 2) \text{match}(G)$$

Lower bound



(1)

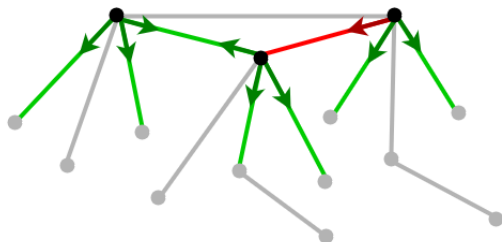
$$|E_\alpha| = g_0 + g_1 + g_2$$

Defining E^*

Lemma

$$\text{match}(G) \leq |E_\alpha| \leq (\alpha + 2) \text{match}(G)$$

Lower bound



(2)

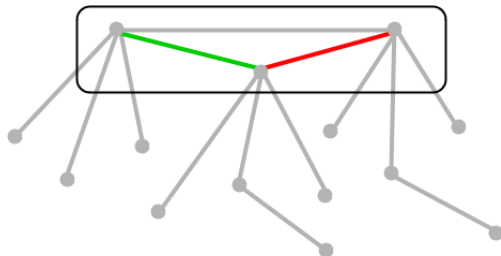
$$(\alpha + 1)|V^H| = \sum_{h \in V^H} |B_h| = g_1 + 2g_2 + w_2$$

Defining E^*

Lemma

$$\text{match}(G) \leq |E_\alpha| \leq (\alpha + 2) \text{match}(G)$$

Lower bound



(3)

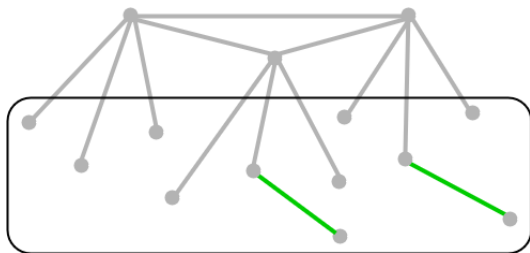
$$\alpha |V^H| \geq |E^H| \geq g_2 + w_2$$

Defining E^*

Lemma

$$\text{match}(G) \leq |E_\alpha| \leq (\alpha + 2) \text{match}(G)$$

Lower bound



(4)

$$|E^L| = g_0$$

Defining E^*

Lower bound

$$(1) |E_\alpha| = g_0 + g_1 + g_2$$

$$(2) (\alpha + 1)|V^H| = g_1 + 2g_2 + w_2$$

$$(3) \alpha|V^H| \geq g_2 + w_2$$

$$(4) |E^L| = g_0$$

$$|E_\alpha| = g_0 + g_1 + g_2 \tag{1}$$

$$= |E^L| + g_1 + g_2 \tag{4}$$

$$= |E^L| + (g_1 + 2g_2 + w_2) - (g_2 + w_2)$$

$$= |E^L| + (\alpha + 1)|V^H| - (g_2 + w_2) \tag{2}$$

$$\geq |E^L| + (\alpha + 1)|V^H| - \alpha|V^H| \tag{3}$$

$$= |E^L| + |V^H|$$

$$\geq \text{match}(G)$$

Every matched edge is either light or incident to a heavy vertex.

Defining E^*

Cormode et al. show

$$\text{match}(G) \leq 3|E_{6\alpha}| \leq (22.5\alpha + 6) \text{match}(G)$$

and provide algorithm approximating $|E_{6\alpha}|$.

Approximating E^* instead allows us to

- improve approximation factor
- shave off α factor in space used
- simplify the analysis

$(1 + \epsilon)$ -approximation of E^*

Idea: keep a sample of edges in E_α^t by sampling with probability that allows us to

- keep an accurate approximation of $|E_\alpha^t|$
- use small amount of space

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Cormode et al. run $\log_{1+\epsilon} n^2$ copies of the algorithm in parallel for $p = 1, (1 + \epsilon)^{-1}, (1 + \epsilon)^{-2}, \dots, 1/n^2$ and discard copies with large sample size.

We run one copy with p starting at 1 and lower sampling probability as needed.

Allows us to shave off $\log n$ factor in space.

$(1 + \epsilon)$ -approximation of E^*

1. Set $p \leftarrow 1$
2. Start sampling each edge with probability p
3. If e is sampled:
 - store e
 - store counters for degrees of endpoints in the rest of the stream
 - if later we detect $e \notin E_\alpha^t$, it is deleted
4. If the number of stored edges $> 40\epsilon^{-2} \log n$
 - $p \leftarrow p/2$
 - delete every edge currently stored with probability $1/2$
5. Return $\max_t \frac{\# \text{ samples at time } t}{p \text{ at time } t}$

$(1 + \epsilon)$ -approximation of E^*

Let k be s.t. $(20\epsilon^{-2} \log n)2^{k-1} \leq E^* < (20\epsilon^{-2} \log n)2^k$.

We show that whp:

1. If sampling probability is high enough ($\geq 1/2^k$), it provides an accurate approximation of $|E_\alpha^t|$ for all t .
From Chernoff and union bounds.
2. We do not switch to probability that is too low ($< 1/2^k$), since the $\#$ edges sampled wp $1/2^k$ never exceeds $(1 + \epsilon)E^*/2^k < (1 + \epsilon)(20\epsilon^{-2} \log n) \leq 40\epsilon^{-2} \log n$.

Conclusion

Problem: approximate the size of maximum matching in arboricity α graph defined by an insert-only stream.

We improve previous best algorithm by Cormode et al. providing

- better approximation factor
- better space
- simpler analysis

Thank you for your attention!

Bibliography

- H. Esfandiari, M. TaghiHajiaghayi, V. Liaghat, M. Monemizadeh, and Krzysztof Onak, *Streaming algorithms for estimating the matching size in planar graphs and beyond*, SODA 2015.
- A. McGregor and S. Vorotnikova, *Planar Matching in Streams Revisited*, APPROX 2016.
- G. Cormode, H. Jowhari, M. Monemizadeh, and S. Muthukrishnan. *The Sparse Awakens: Streaming Algorithms for Matching Size Estimation in Sparse Graphs*, ESA 2017.