

Matching in randomly ordered Streams

Workshop on Algorithms for Data Streams 2012

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joint work with

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Graph Streams and Bipartite Matching

$G = (A, B, E)$ bipartite, $n = |A| = |B|$, $m = |E|$

Graph stream: sequence of edges, any order
 $\pi = (3, 2), (7, 6), (1, 2), (7, 8), \dots (5, 6)$

Bipartite Matching in Streaming:

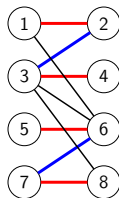
perform one pass, compute large matching using *little* space

Memory considerations: [Feigenbaum, Kannan, McGregor, Suri, Zhang, SODA 2005]
deciding basic graph properties such as bipartiteness and connectivity
requires $\Omega(n)$ space

Semi-Streaming Model: $O(n \text{ polylog } n)$ space

From now on:

M^* : fixed maximum matching (matching of maximal size)
Simplification: graph has perfect matching (all vertices matched)

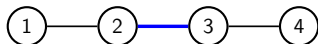


Adversarial Arrival Order

Input sequence: No assumption on the order

Upper Bound: $\frac{1}{2}$ -approximation, Greedy Algorithm

- start with empty matching, insert incoming edge if possible
- **Example:** $\pi = (2, 3), (1, 2), (3, 4)$



$$\text{Greedy}(\pi) = \{(2, 3)\} \quad M^* = \{(1, 2), (3, 4)\}$$

- **Maximal matchings:** cannot be enlarged by simply adding an edge
 - Maximal matchings are of size at least $\frac{1}{2}|M^*|$
 - Greedy computes a maximal matching $\rightarrow \frac{1}{2}$ **approximation**

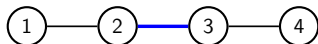
Lower Bound: [Karpalov, 2012] $1 - \frac{1}{e} \approx 0.63$

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Open question: Can we break $\frac{1}{2}$ in one pass?

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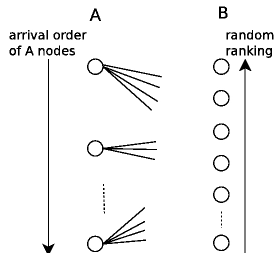
Two passes: [Konrad, Magniez, Mathieu, APPROX 2012] $\frac{1}{2} + 0.019$ approx.

Vertex Arrival Order

Input sequence: Edges sorted with respect to incident A node

Upper Bounds: $1 - \frac{1}{e}$ approximation

- [Karp, Vazirani, Vazirani, STOC 1990]
 - Online Algorithm: upon arrival of a node with its edges, match node irrevocably
 - Rank B nodes randomly, match A node to free B node with highest rank
- [Goel, Kapralov, Khanna, SODA 2012]
deterministic Algorithm achieving same approximation



Lower Bound: [Kapralov, 2012] $1 - \frac{1}{e}$

Random Arrival Order

Input sequence: Edges come in in uniform random order

Upper Bound: [Konrad, Magniez, Mathieu, APPROX 2012]

$\frac{1}{2} + 0.005$ approximation in expectation

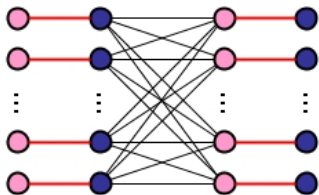
- Random Arrival Order allows to break $\frac{1}{2}$
- *randomized Greedy Algorithm*

Analysis of Greedy Matching Algorithms:

- **Another Greedy Algorithm:** choose randomly some vertex, and then randomly an incident edge
- [Aronson, Dyer, Frieze, Suen, 1995] $\frac{1}{2} + 0.0000025$ approximation
- [Poloczek, Szegedy, FOCS 2012] $\frac{1}{2} + 0.0039$ approximation

Some Intuition: Hard Instance for Greedy

$$G = (A, B, E), |A| = |B| = N$$

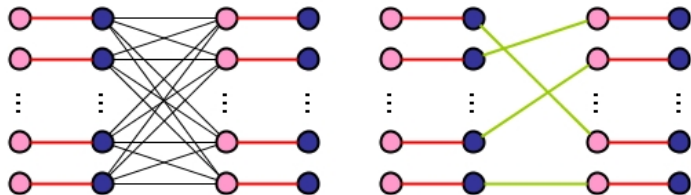


Analysis:

- Perfect matching $|M^*| = N$
- Greedy : $|\mathbb{E}_\pi \text{Greedy}(\pi)| = \frac{1}{2}N$

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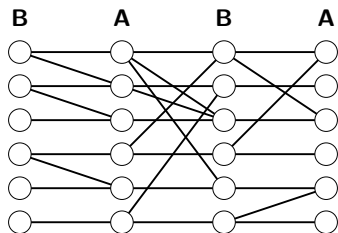
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Structure of small maximal matchings ($\approx \frac{1}{2}$ -approximations):

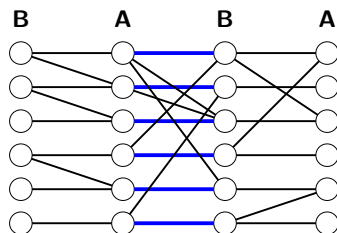


Almost all edges form 3-augmenting paths with optimal edges

Three passes Streaming Algorithm

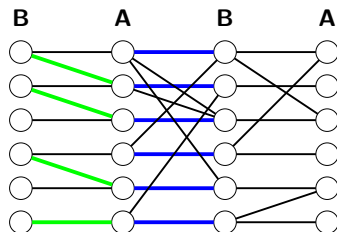


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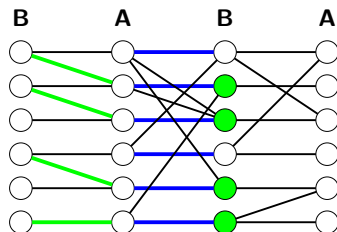
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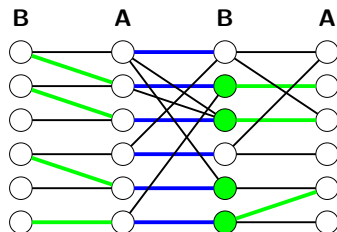
- 1 Maximal matching M_0 : Greedy
- 2 Left wings M_1 : Greedy between $A(M_0)$ and $B \setminus B(M_0)$

Three passes Streaming Algorithm



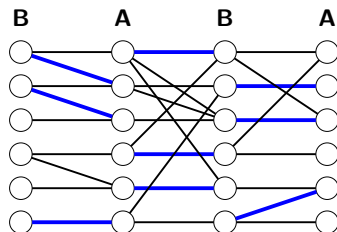
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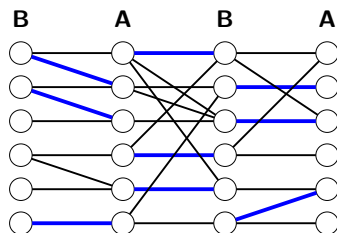
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- 4 Augment M_0 with M_1 and M_2

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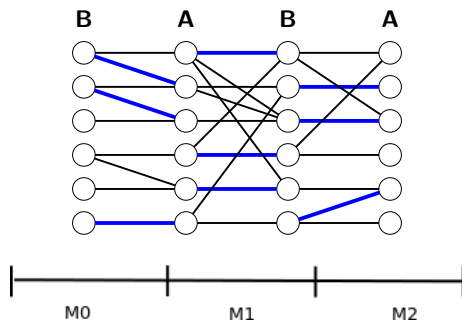
Can we implement this strategy with less passes?

Difficulty: highly linear approach

M_1 depends on M_0 , M_2 depends on M_1 and M_0

One-pass random order

Idea: split stream into 3 parts, run on each part a pass



Crucial properties:

- M_0 : maximal matching
- Sufficiently many edges for augmentation in second part of the stream (guaranteed by random order assumption)

New Property of Greedy

Lemma: If Greedy performs badly then Greedy converges quickly

If $|\mathbb{E}_\pi \text{Greedy}(\pi)| = (\frac{1}{2} + \epsilon)|M^*|$ then

$$|\mathbb{E}_\pi \text{Greedy}(\pi[1, \alpha m])| = (\frac{1}{2} - (\frac{1}{\alpha} - 2)\epsilon)|M^*|$$

Corollary: $(\alpha = \frac{1}{2}) |\mathbb{E}_\pi \text{Greedy}(\pi[1, \frac{1}{2}m])| \geq \frac{1}{2}|M^*|$

Some Intuition:

- Greedy performs badly: it misses almost all optimal edges
- Random order assumption: many optimal edges arrive *early*
- Early optimal edges blocked: many non-optimal edges taken early

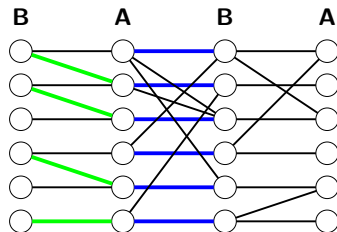
Blocks: $[0, 0.43m], [0.43m, 0.76m], [0.76m, m]$

$\rightarrow \frac{1}{2} + 0.005$ approximation in expectation for random order

Two-passes for adversarial order

Two pass Algorithm for adversarial order:

- **First pass:** M_0 and M_1 (Greedy matching + left wings)
- **Second pass:** M_2 (right wings)



Difficulty: M_1 depends strongly on M_0 :

$$M_1 = \text{Greedy between } A(M_0) \text{ and } B \setminus B(M_0)$$

Another new property of Greedy

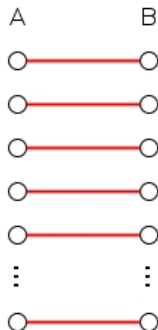
Matching subsets of B :

- **Lemma:** π any input sequence, $B' \subset B$ uniform random sample such that $\forall b \in B : P[b \in B'] = p$. Then:

$$\mathbb{E}_{B'} |\text{Greedy}(\pi, G|_{A \times B'})| \geq \frac{p}{1+p} |M^*|$$

- **Intuition:**

- Graph with perfect matching M^* , $B' \subset B$
Potential ϕ : perfect edges in $G|_{A \times B'}$
 $\mathbb{E} \phi_0 = |M^*| p$



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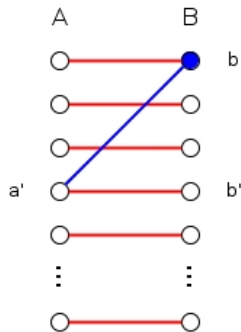
- **Intuition:**

- Graph with perfect matching M^* , $B' \subset B$
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- Consider edge $a'b$ incident to node $b \in B'$

Bad case: $\Delta\phi = 2$ if $b' \in B'$

Good case: $\Delta\phi = 1$ if $b' \notin B'$

$$\mathbb{E} \Delta\phi = p \cdot 2 + (1 - p) \cdot 1 = 1 + p$$



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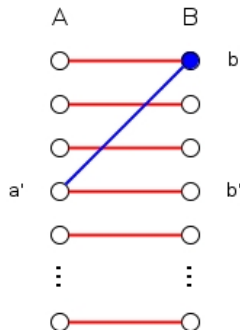
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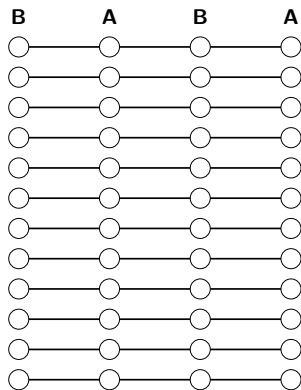
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- Matching size: # rounds until potential = 0

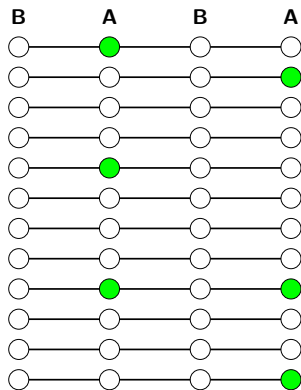
$$\frac{\mathbb{E} \phi_0}{\mathbb{E} \Delta\phi} = \frac{p}{1+p} |M^*|$$



Two passes Algorithm

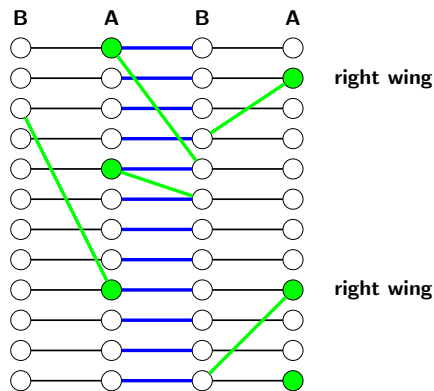


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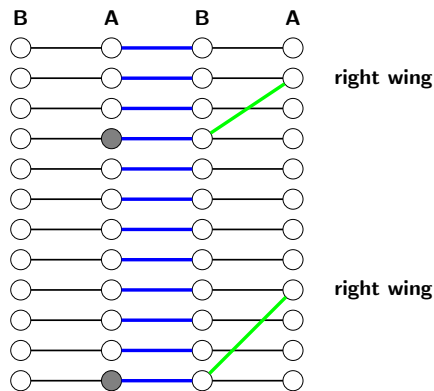
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Two passes Algorithm



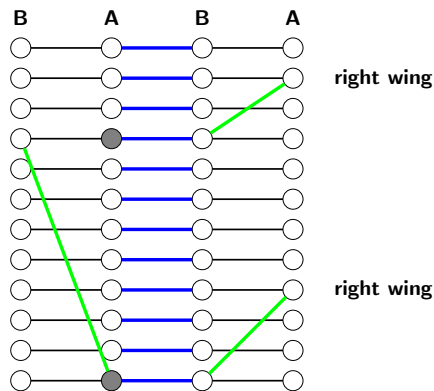
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- 3 in one pass: find left wings M_2 for \bullet nodes (Greedy matching)

Bipartite Matching:

Order	Passes	Upper Bound Approx.	Lower Bound
Adversarial	1 pass	$1/2$	$1 - 1/e$
Adversarial	2 passes	$1/2 + 0.019$	-
Vertex Arrival	1 pass	$1 - 1/e$	$1 - 1/e$
Random	1 pass	$1/2 + 0.005$	-

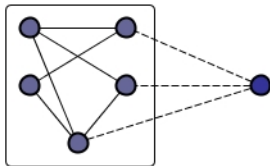
Remarks:

- Deterministic 2-passes version for adversarial order
- No upper bounds require randomization
- Presented algorithms extends to general graphs

Open Problem: Extended vertex arrival order

Extended vertex arrival order

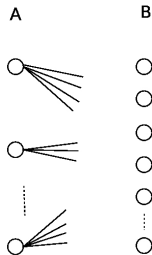
Input sequence: Vertices come in together with incident edges to already present vertices



Model captures vertex arrival model:

B vertices arrive first, then A vertices

Can $\frac{1}{2}$ be broken in this model?



Thank you for your attention.

Questions?