Faster Fully Dynamic Matching With Small Approximation Ratios

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Static Maximum Matching

- Unweighted n-vertex, m-edge graph G
- Goal: find a maximum cardinality matching (MCM) in G
- Exact algorithms:
 - O(mn^{1/2}) [Micali, Vazirani FOCS 1980]
 - \circ O(MM(n)) \sim O(n^{2.36}) [Sankowski FOCS 2004]
- Exact algorithms for bipartite graphs:
 - O(mn^{1/2}) [Hopcroft-Karp, Sicomp, 1973]
 - \circ $\tilde{O}(m^{10/7})$ [Madry FOCS 2013].
- Approximate algorithms:
 - \circ (1 + ε)-approximation in time $\tilde{O}(m/\varepsilon)$

Dynamic Exact MCM

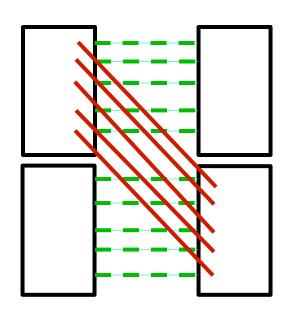
- Dynamic edges are inserted and deleted, we maintain a matching
- □ Trivial:
 - Recompute O(mn^{1/2}) or O(MM(n)) per update
 - Find augmenting path O(m) per update
- Sankowski's algorithm:
 - O(n^{1.495}) randomized [SODA 2007].
- What about dynamic approximate MCM?
- Note: The "adwords" problem is a different dynamic matching problem

Dynamic Approximate MCM

- Variants:
 - Randomized vs. deterministic
 - Worst-case vs. amortized
 - Bipartite vs. non-bipartite
- □ Two very different problems:
 - 2-or-worse approximation.
 - Better-than-2 approximation.

Maximal Matching

- Maximal: can't add an edge to the matching
- Maximal ≠ Maximum
- Maximal matching is 2approximate.
- Maximal matching purely local
 - If x and y are free, can always match x and y, without considering rest of graph.
- 1.99 approximation not purely local.



- Maximum has size 10
- Maximal has size 5

Dynamic Approximate MCM (before us)

- 2-or-worse Approximation:
 - O(log(n)) update, 2-approximation, randomized [Baswana, Gupta, Sen, FOCS 2011]
- Better-than-2 approximation:
 - O(m^{1/2}) update, (1+ε)-approximation. [Gupta, Peng, FOCS 2013]
- Deterministic Algorithms:
 - \circ O(m^{1/2}) update, (1+ ε)-approximation [Gupta, Peng]
 - O(m^{1/3}) update, (3+ ε)-approximation [Bhattacharya, Henzinger, Italiano, SODA 2015]
- Our Goals:
 - faster better-than-2 approximation
 - faster deterministic algorithm.

Our results

Recall State of the art:

- 2-approx.,O(log n) update, randomized
- \circ (1+ ε)-approx., $O(m^{1/2})$ update, deterministic
- \circ (3+ ε)-approx., O(m^{1/3}) update, deterministic

Our results:

- (3/2+ε)-approx., O(m^{1/4}) update, deterministic
- Same bounds (different algorithm and analysis for bipartite (ICALP 2015) and general (SODA 2016) graphs
- Fastest better-than-2 approximation.
- Fastest deterministic algorithm for any constant approx.
- Also: better results for small arboricity graphs (see papers)

Subsequent to our results

Recall our results:

- \circ (3/2+ ε)-approx., O(m^{1/4}) update, deterministic
- Same bounds (different algorithm and analysis for bipartite (ICALP 2015) and general (SODA 2016) graphs
- Fastest better-than-2 approximation.
- Fastest deterministic algorithm for any constant approx.
- Also: better results for small arboricity graphs

Simultaneous:

- Even better results for small arboricity: [Peleg, Solomon SODA 2016]
- Subsequent: [Bhattacharya, Henzinger, Nanongkai, STOC 2016]
 - O(polylog n) update, (2+ ε)-approximation
 - \circ O(n^{ϵ}) update, <2-approximation of value in bipartite graphs

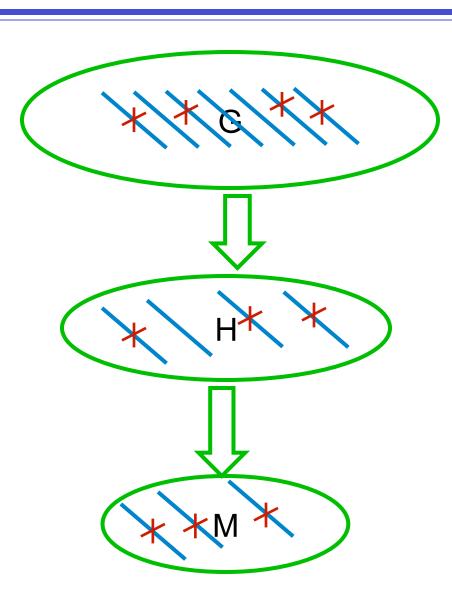
Characterizing Approximate Matchings

- □ Exact matching: matching M contains no augmenting paths → M is maximum
- 2-approximate matching: matching M is maximal -> M is 2-approximate
 - Maximal: for every edge (u,v), u or v (or both) is matched
- Better-than-2 approximation: if all augmenting paths w.r.t M have length > (2k-1) then M is a (1+1/k)-approx. matching.
 - All path lengths > 1→2-approximation
 - All path lengths > 3
 →
 3/2-approximation.
 - O All path lengths > 1/ε \rightarrow (1+O(ε))-approximation.
 - Unlike maximality, cannot be expressed in terms of local constraints on each edge

Characterizing Approximate Matchings

- □ Ideally show: A matching M that satisfies some set of local constraints is a (1+ε)-approximate matching (or at least better-than-2 approximate).
 - We were unable to do this even for bipartite graphs.
- Our results: use local constraints to characterize a small subgraph H of G such that H is guaranteed to contain a large matching. (used in previous work, but with different small subgraph, typically fractional or b-matching.)
- Recall a b-matching is a subgraph where each vertex has degree at most b

Matching-Preserving Subgraph



■ What we want from H:

- 1. M is easy to maintain in H
- 2. Large matching in H is relatively large w.r.t to G
- 3. Easy to maintain H dynamically.

Edge Degree Constrained Subgraph (EDCS)

- \Box **Defn:** $\mu(H)$ = size of maximal matching in H.
- \Box **Defn:** $\delta_H(u)$ = degree of u in H.
- □ A subgraph H of G is an EDCS(B) if it has the following properties (B is parameter of our choosing):
 - 1. For each edge (u,v) in H,

$$\delta_{H}(u) + \delta_{H}(v) \leq B$$

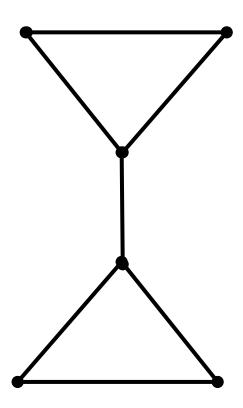
○2. For each edge (u,v) in G\H,

$$\delta_{H}(u) + \delta_{H}(v) \ge B(1-\epsilon^2)$$

Example of EDCS

- □ Example of EDCS(3):
 - 1. For each edge (u,v) in H, δ_H (u) + δ_H (v) ≤ 3
 - 2. For each edge (u,v) in G\H, $\delta_H(u) + \delta_H(v) \ge 2$

$$\mu(G) = 3$$

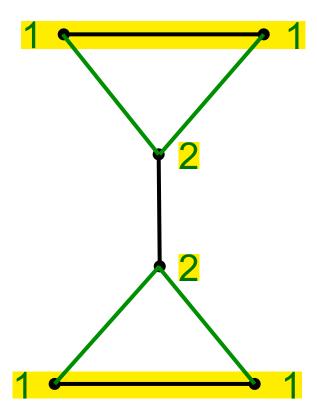


Example of EDCS

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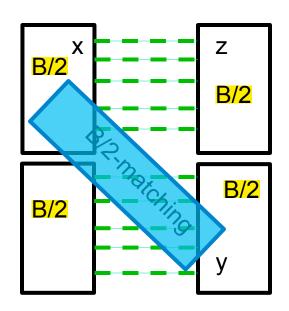
$$\mu(G) = 3$$

$$\mu(H) = 2$$



Example of a non-EDCS

- □ Example of EDCS(B):
 - 1. For each edge (u,v) in H, δ_H (u) + δ_H (v) ≤ B
 - 2. For each edge (u,v) in G\H, $\delta_H(u) + \delta_H(v) \ge B(1-\epsilon^2)$



$$\delta_H(x) + \delta_H(z) = B/2 + 1$$

$$\delta_{H}(x) + \delta_{H}(y) = B$$

- B/2-matching (complete bipartiite graph) only contains a 2approximation to the maximum matching
- B/2-matching is not an EDCS
- Our "local" algorithm will recognize this

What do we need to do?

- Prove the main theorem: Let H be an EDCS(B) of a graph G. Then: μ(H) ≥ (2/3-ε) μ(G)
 - only true if $B > 4/\epsilon^2$
 - H has bounded degree B
- 2. Algorithm problem 1: Show how to maintain an EDCS(B) quickly in a dynamic graph
- 3. Algorithm problem 2: Maintain a (1+ε) approx. matching M in H
 - Easy because H has bounded degree [Gupta, Peng, FOCS 2013]

3/2-approximation follows

□ Size(M) ≥ $(1-\epsilon)\mu(H) \ge (2/3-\epsilon)(1-\epsilon)\mu(G) = (2/3 - O(\epsilon))\mu(G)$

Now, let's briefly talk about maintaining an EDCS, and then the main theorem

■ EDCS(B):

- □ 1. For each edge (u,v) in H, $\delta_H(u) + \delta_H(v) \le B$
- 2. For each edge (u,v) in G\H,

$$\delta_{H}(u) + \delta_{H}(v) \ge B(1-\epsilon^2)$$

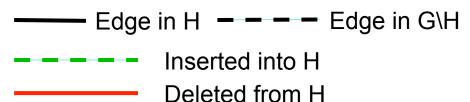
—— Edge in H − − − − Edge in G\H - − − − Inserted into H

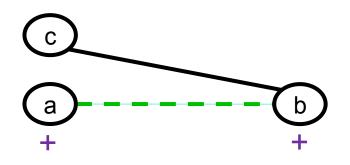
Deleted from H



- □ 1. For each edge (u,v) in H, $\delta_H(u) + \delta_H(v) \le B$
- 2. For each edge (u,v) in G/H,

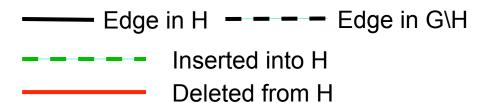
$$\delta_{H}(u) + \delta_{H}(v) \ge B(1-\epsilon^2)$$

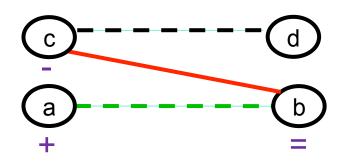




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- 2. For each edge (u,v) in G/H,

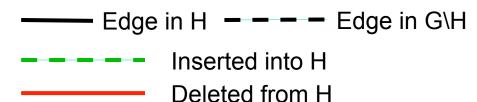
$$\delta_{H}(u) + \delta_{H}(v) \ge B(1-\epsilon^2)$$

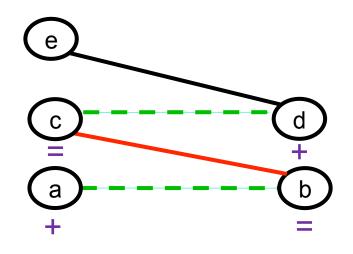




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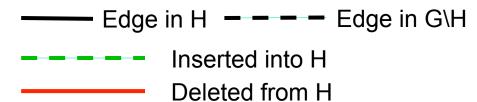
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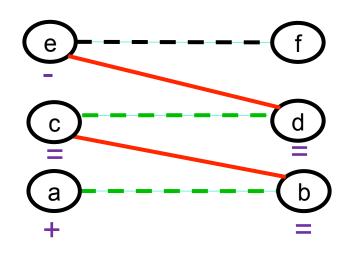




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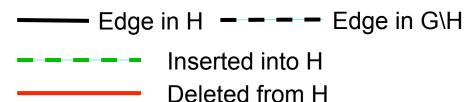
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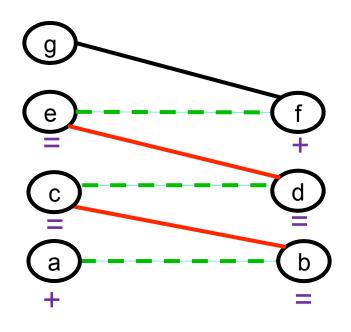




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- 2. For each edge (u,v) in G/H,

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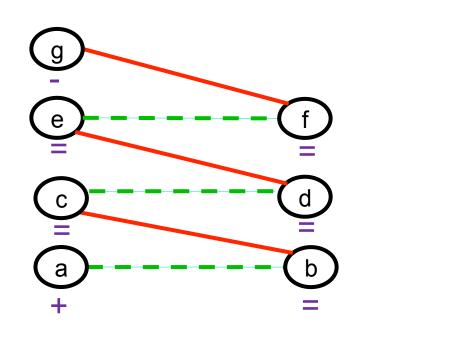




- **□** EDCS(B, λ):
 - □ 1. For each edge (u,v) in H, $\frac{1}{\Delta_H(u) + \delta_H(v)} \leq B$ Inserted into H Deleted from H
 - □ 2. For each edge (u,v) in G/ H, $\delta_H(u) + \delta_H(v) \ge B(1- ε^2)$
- \Box Can show: only need $O(ε^{-2})$ rebalances
- Difference from augmenting path:

Any sequences of rebalances will work and be short.

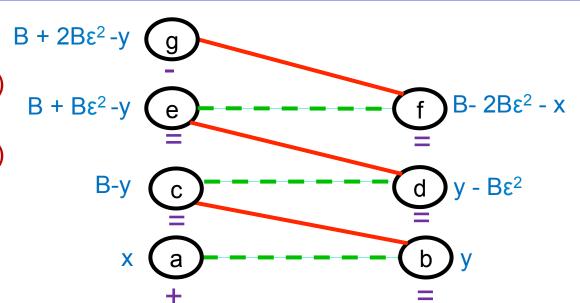
No "backtracking"



Edge in H - - - Edge in G\H

Bipartite running time

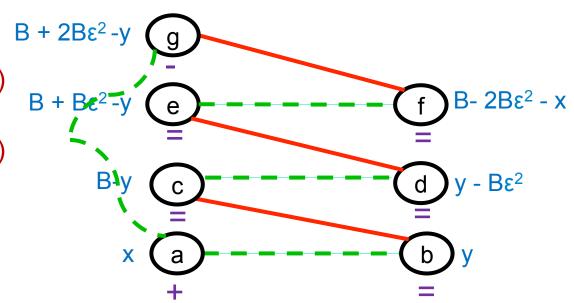
- □ 1. For each edge (u,v) in H, $\delta_H(u) + \delta_H(v) \le B$
- □ 2. For each edge (u,v) in G\H, δ_H (u) + δ_H (v) ≥ B(1- ε²)



- Only O(ε-2) rebalances per edge change in G
- Easy: Rebalance a node in time O(degree) = O(n)
- Can do: Rebalance in time O(arboricity) (dynamic orientation++)
- □ General Graphs: arboricity = m^{1/2}
- Lazy Rebalancing: buffer of Bε-2 between EDCS properties
- Resulting worst case update time: O(m^{1/2} ε⁻²/B)

NonBipartite running time

- **□** EDCS(B, λ):
 - □ 1. For each edge (u,v) in H, δ_H (u) + δ_H (v) ≤ B
 - □ 2. For each edge (u,v) in G\H, δ_H (u) + δ_H (v) ≥ B(1- ε²)

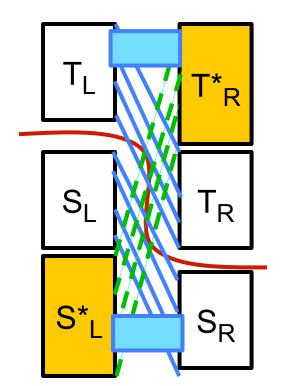


- Can't show O(ε-2) rebalances per edge change in G, but can show a simliar amortized bound, using
 - Potential function
 - Dynamic orientation
 - Lots of data structures
- Implicitly dealing with blossoms

Showing an EDCS contains a large matching

Proof sketch for bipartite graphs

- The residual graph defined by the maximum matching in H (n vertices per side).
- □ Contradiction Assumption: $\mu(H) < (2/3-ε)n$



- H contains no edges crossing cut.
- □ $|S^*_L| = |T^*_R| \ge (n/3)(1+\varepsilon)$
- □ Assume: $|S_1| = |S_R| = |T_1| = |T_R| < n/3$
- \supseteq $(n/3)(1+\epsilon)$ disjoint crossing edges in G\H
- Assume: crossing edges go from S*_L to T*_R
- Then: average degree in H of |S*_L| U |T*_R| ≥ B/2(1-ε²) ~ B/2 (Prop. 2)
- Can't fit degree of S*_L in smaller S_R (Prop. 1)

- □ 1. For edge (u,v) in H, $\delta(u)+\delta(v) \leq B$
- □ 2. For edge (u,v) in G\H, δ (u)+ δ (v) \geq B(1- ϵ ²)

Main Theorem in Nonbipartite Graphs

- Defn: a subgraph H of G is an EDCS(B) if:
 - 1. For each edge (u,v) in H, $\delta_H(u) + \delta_H(v) \le B$
 - 2. For each edge (u,v) in G\H, $\delta_H(u) + \delta_H(v) \ge B(1-\epsilon^2)$
- □ Main Thm: if H is an EDCS(B), μ (H) ≥ (2/3-ε) μ (G)
- Idea: explicitly construct a large fractional matching in H
 - Let val(u,v) be the fractional value on edge (u,v) in a fractional matching
 - \circ val(u,v) will depend on $\delta_H(u)$ and $\delta_H(v)$
- Simplifies several proofs for bipartite case
- Problem: integrality gap in nonbipartite graphs

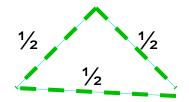
Fractional Matching

- Dfn: let μ_f(G) be the size of maximum fractional matching in G
- □ Bipartite Graphs: $\mu_f(G) = \mu(G)$
- □ Nonbipartite Graphs: $\mu(G) \ge (2/3)\mu_f(G)$
- Can show: if H is an EDCS(B), then

$$\mu_f(H) \ge (2/3 - \varepsilon)\mu(G)$$

■ Not good enough:

$$\mu(H) \ge (2/3)\mu_f(H) \ge (4/9 - \varepsilon)\mu(G)$$



α-Restricted Fractional Matching

- Dfn: Let val(u,v) be the fractional value on edge (u,v) in a fractional matching
- □ A fractional matching is α -restricted (α <1) if for every edge (u,v) either val(u,v) = 1 or val(u,v) ≤ α
- Dfn: let μα_f(G) be the maximum value of an α-restricted fractional matching in G
- Classic Thm:

$$\mu(G) \ge (2/3)\mu_f(G)$$

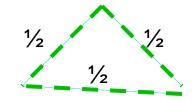
Our Thm:

$$\mu(G) \ge (\alpha/(\alpha+1)) \mu_{f}^{\alpha}(G)$$

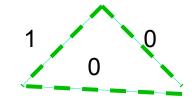
Corollary:

$$\mu(G) \ge (1-\epsilon) \mu_{f}^{\epsilon}(G)$$





$$\alpha = 1/3$$



EDCS contains a large ε-restricted matching

Main Thm: If a subgraph H is an EDCS(B):

$$\mu(H) \ge (2/3-\epsilon) \mu(G)$$

Main Lemma: If a subgraph H is an EDCS(B):

$$\mu^{\epsilon}_{f}(H) \ge (2/3-\epsilon) \mu(G)$$

■ Main Lemma → Main Theorem:

$$\mu(H) \ge (1-\epsilon) \mu^{\epsilon}_{f}(H) \ge (2/3-\epsilon)(1-\epsilon)\mu(G) = (2/3-O(\epsilon))\mu(G)$$

Explicit Construction

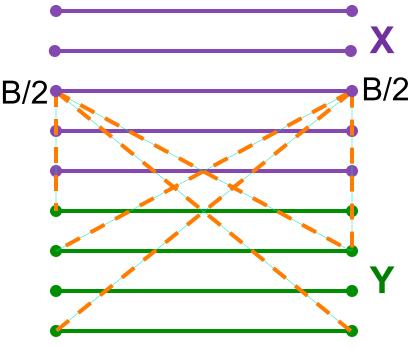
- Given some subgraph H that is an EDCS(B)
 - 1. For each edge (u,v) in H, $\delta_H(u) + \delta_H(v) \le B$
 - 2. For each edge (u,v) in G\H, $\delta_H(u) + \delta_H(v) \ge B(1-\epsilon^2)$
- Want to show: can construct an ε-restricted fractional matching M_H of H such that:

$$val(M_H) \ge (2/3-\epsilon)\mu(G)$$

Very involved proof. For this presentation will make many simplifying assumptions.

Proof Sketch of main theorem

- □ Simplified EDCS constraints (u,v) in H: $\delta_H(u) + \delta_H(v) \le B$ (u,v) in G\H: $\delta_H(u) + \delta_H(v) \ge B$
 - Maximum matching M_G Note: $|X| + |Y| = 2\mu(G)$

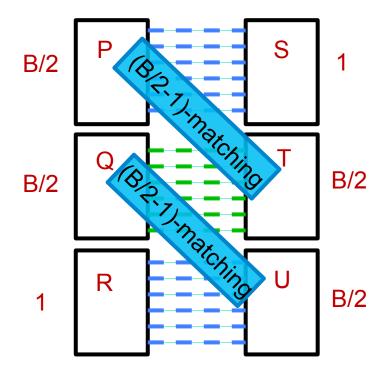


- **Big Assumption:** if edge (u,v) in X-X: $\delta_H(u) = \delta_H(v) = B/2$
- Medium Assumption: All edges in H incident to Y
- □ Fractional Matching: for each edge (u,v) in X-Y, val(u,v) = 2/B
 - Legality: all relevant vertices have degree at most B/2
- Total value: |X|(B/2)(2/B) = |X|
- □ Case 1: $|X| \ge 2\mu(G)/3$
- □ Case 2: $|Y| \ge 4\mu(G)/3$. At least $2\mu(G)/3$ edges in Y-Y
- Actual proof: mix two cases. Probabilistic method.

EDCS is only a 3/2-approximation.

Unweighted EDCS(B):

- 1. For each edge (u,v) in H, $\delta_H(u) + \delta_H(v) \le B$
- 2. For each edge (u,v) in G\H, $\delta_H(u) + \delta_H(v) \ge B(1-\epsilon^2)$

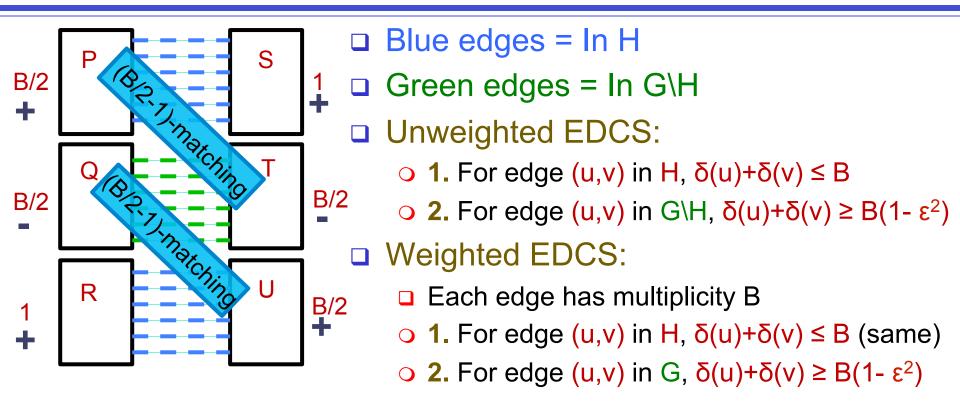


- □ Blue edges = In H
- □ Green edges = In G\H
- $\mu(H) = 2n/3 = 2/3 \mu(G)$
- Note: edges between P-S and R-U have edge degree
 << B(1- ε²)
- □ H is a valid EDCS only because constraint 2 only applies to edges in G\H

Weighted EDCS

- Unweighted EDCS(B):
 - 1. For each edge (u,v) in H, $\delta_H(u) + \delta_H(v) \le B$
 - 2. For each edge (u,v) in G\H, $\delta_H(u) + \delta_H(v) \ge B(1-\epsilon^2)$
- □ **Definition**: **G**^B = G, but with B copies of each edge
- Weighted EDCS(B):
 - Equivalent to Unweighted EDCS(B) on G^B
 - \circ Degree $\delta_H(u)$ now takes into account multiplicity.
 - Multiplicities can be ignored when computing the matching M in H because in M max degree = 2.
- Constraint 2 in a weighted EDCS:
 - Some copy of edge (u,v) is always in G\H
 - So: for each edge (u,v) in G, $\delta_H(u) + \delta_H(v) \ge B(1-\epsilon^2)$

Unweighted VS. Weighted



- □ Unweighted EDCS: $\mu(H) = 2n/3$.
- Weighted EDCS: P-S and R-U edges violate Prop. 2.
 - Must add weight to P-S and R-U edges.
 - Must then remove weight from P-T and Q-U edges (Prop. 1)
 - Must now add weight to Q-T edges (Prop. 2).

Weighted EDCS

- Can show that in a weighted EDCS satisfies
 μ(H) ≥ (1- ε) μ(G)
- Can only maintain a weighted EDCS efficiently in low arboricity bipartite graphs
- Intuitively, with a weighted EDCS, deleting 1 edge in G can deleted a large amount of weight in H.

Review

- Main Result: O(m^{1/4}) update time, (3/2-ε)-approx.
 - Fastest better than 2-approximation
 - Fastest deterministic algorithm for any constant approx.
- □ EDCS(B): subgraph H such that
 - 1. For each edge (u,v) in H, $\delta_H(u) + \delta_H(v) \leq B$
 - 2. For each edge (u,v) in G\H, $\delta_H(u) + \delta_H(v) \ge B(1-\epsilon^2)$
- □ Main Theorem: $\mu(H) \ge (3/2-ε) \mu(G)$
 - Techniques very different for bipartite and nonbipartite case
- Lots of dynamic graph and graph orientation details to achieve running time

Open Problems

- □ Can we maintain a (1+ε) approximation in update time $O(m^{1/2-c})$ update time for any fixed c > 0?
- How well can we approximate matching in polylog update time
- Very interesting even on bipartite graphs, with randomization and amortization.
- Weighted EDCS may be helpful