

# Subquadratic Weighted Matroid Intersection Under Rank Oracles

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**Results.** The first **subquadratic**, i.e.,  $\tilde{O}(n^{7/4})$ , **rank-query** algorithm for **weighted** matroid intersection.

**Techniques.** Efficient implementation of previous work.

1. Use binary searches in **weight adjustments**.
2. A subquadratic shortest-path algorithm in **weighted** exchange graphs.

## Definition

A matroid is a tuple  $\mathcal{M} = (V, \mathcal{I})$  over a *ground set*  $V$  and a non-empty family of *independent sets*  $\mathcal{I}$  such that

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2. if  $R, S \in \mathcal{I}$  and  $|R| < |S|$ , then there exists  $x \in S \setminus R$  such that  $R \cup \{x\} \in \mathcal{I}$ .

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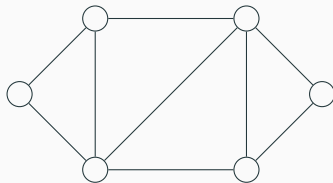
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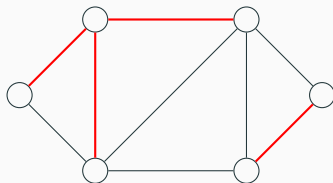
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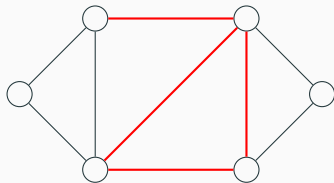
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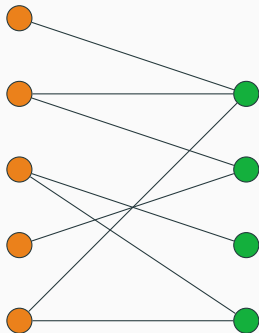
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- Independence query: Is  $S \in \mathcal{I}$ ?
- Rank query: What is  $\text{rank}(S)$ ? ( $\text{rank}(S) = \max_{R \in \mathcal{I}; R \subseteq S} |R|$ )

# Matroid Intersection

Given two matroids  $\mathcal{M}_1 = (V, \mathcal{I}_1)$ ,  $\mathcal{M}_2 = (V, \mathcal{I}_2)$ , find the largest  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$ .

- Bipartite matching
- Colorful spanning tree
- Disjoint spanning trees
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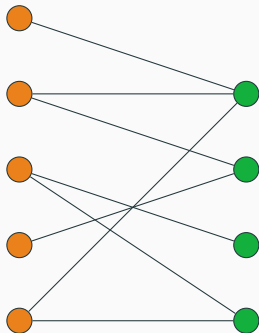
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$\mathcal{M}_2 = r \in R$  has at most one edge

# Weighted Matroid Intersection

Given two matroids  $\mathcal{M}_1 = (V, \mathcal{I}_1)$ ,  $\mathcal{M}_2 = (V, \mathcal{I}_2)$  with weights  $w : V \rightarrow \mathbb{Z}$ , find the  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$  maximizing  $w(S)$ .

- (weighted) Bipartite matching
- (weighted) Colorful spanning tree
- (weighted) Disjoint spanning trees

# Prior Work

'70s	Edmonds, Lawler, Aigner-Dowling	$O(n^3)$ indep
'86	Cunningham	$O(n^{5/2})$ indep
'15	Lee-Sidford-Wong	$\tilde{O}(n^2)$ rank
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'21	Blikstad-v.d.Brand-Mukhopadhyay-Nanongkai	$\tilde{O}(n^{9/5})$ indep

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1. Get a  $(1 - \epsilon)$ -approximate solution in  $\tilde{O}(\frac{n}{\epsilon})$  queries using **weight adjustments**.
2. Find the remaining  $O(\epsilon n)$  **augmenting paths** by solving the **shortest-path problem**.

$\tilde{O}(n^{2-2\delta})$  shortest-path algorithm  $\implies \tilde{O}(n^{2-\delta})$  weighted matroid intersection algorithm

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**Goal.** Find the shortest-path tree rooted at  $s$ .

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**Dijkstra's Algorithm.** Find the **nearest unvisited** vertex in each iteration by **relaxing** edges from **visited** vertices.



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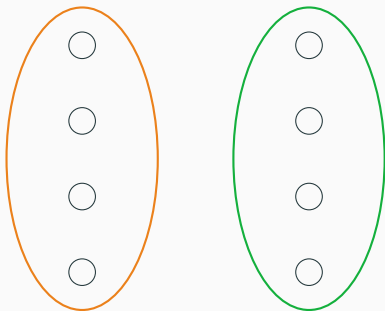
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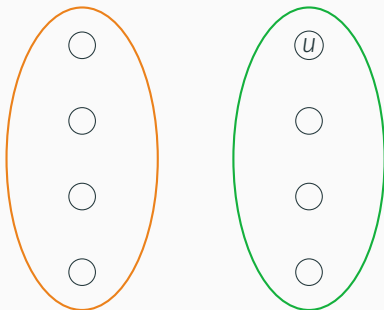
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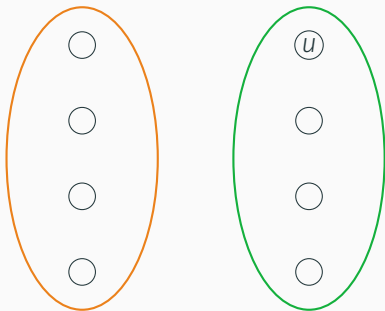
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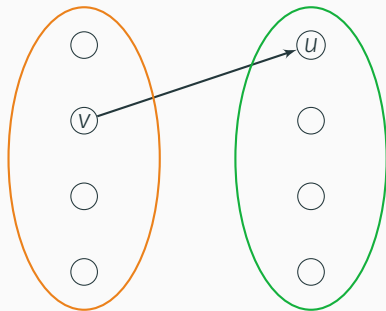


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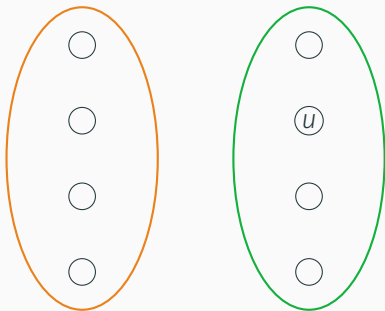


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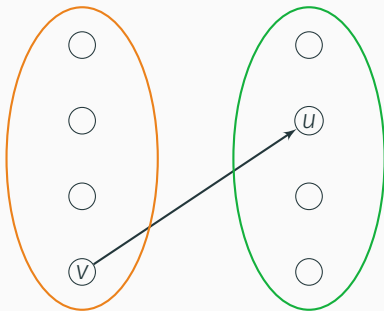


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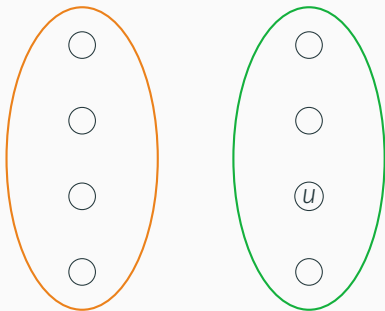


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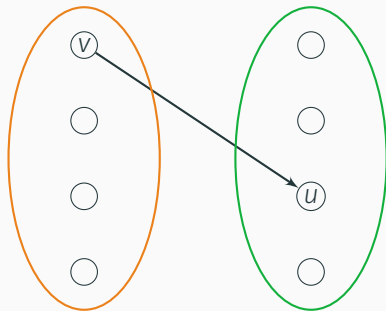
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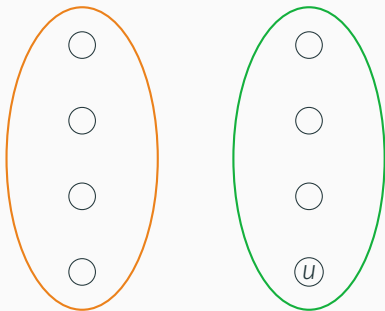


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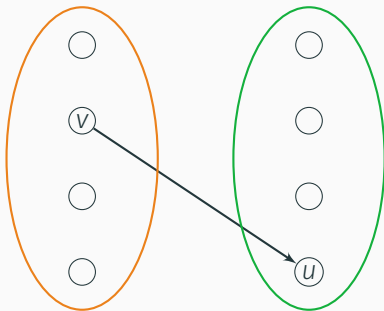


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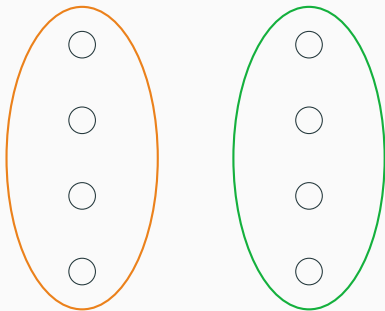
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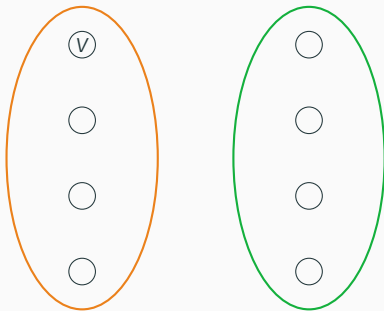
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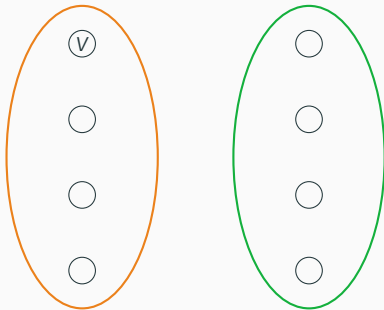
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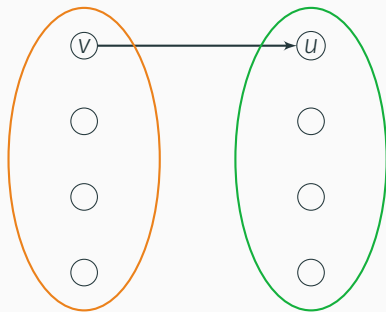


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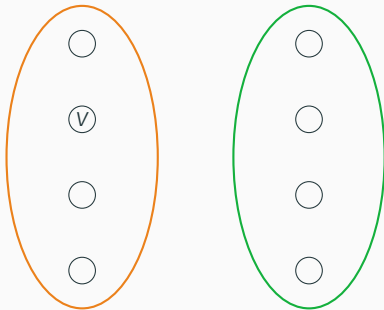
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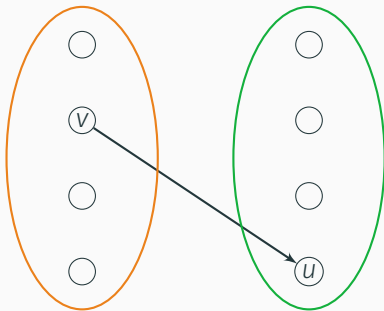


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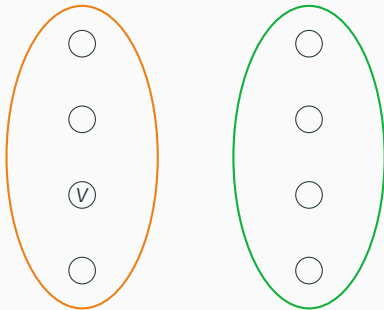


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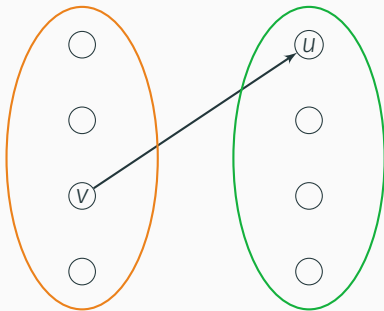


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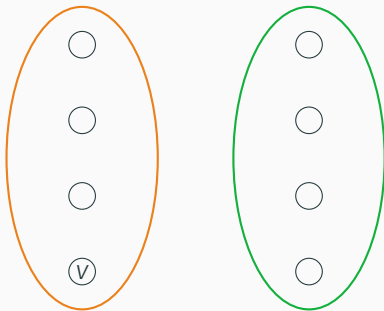


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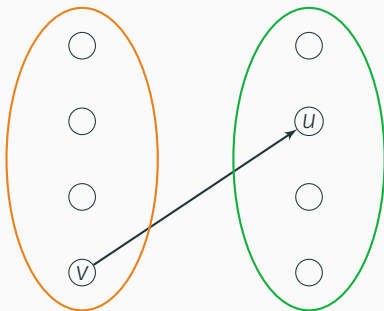


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# A Hybrid Approach

$F$  = visited vertices,  $B \subseteq F$  = **recently** visited vertices

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Total number of queries:  $\tilde{O}(\sqrt{n} \cdot n) + \tilde{O}\left(n \cdot \frac{n}{\sqrt{n}}\right) = \tilde{O}(n\sqrt{n})$

1. Get a  $(1 - \epsilon)$ -approximate solution in  $\tilde{O}(\frac{n}{\epsilon})$  queries using **weight adjustments**.
2. Find the remaining  $O(\epsilon n)$  **augmenting paths** by solving the **shortest-path problem**.

$\tilde{O}(n^{2-2\delta})$  shortest-path algorithm  $\implies \tilde{O}(n^{2-\delta})$  weighted matroid intersection algorithm

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$\tilde{O}(n^{3/2})$  shortest-path algorithm  $\implies \tilde{O}(n^{7/4})$  weighted matroid intersection algorithm

# Open Problems

- Match the unweighted algorithm ( $\tilde{O}(n^{7/4})$  versus  $\tilde{O}(n^{3/2})$ )? Perhaps via a  $\tilde{O}(n)$  shortest-path algorithm.
- Improved strongly polynomial time algorithm?
- What about independence-query algorithms? No  $\tilde{O}(n^2)$  “combinatorial” algorithm yet.
- Lower bounds?

Thanks for listening! Any question?