

Finding a Maximum Weight Triangle in $O(n^{3-\delta})$ Time, With Applications

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The Problem

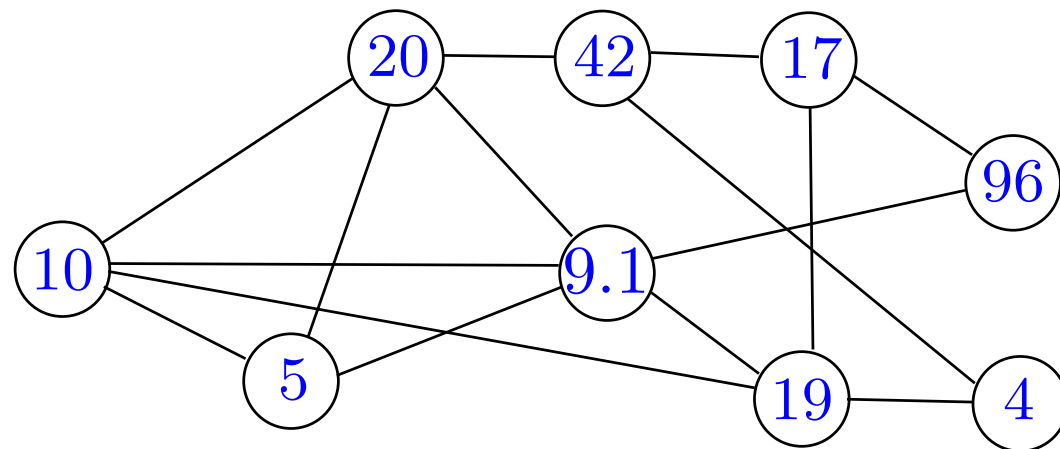
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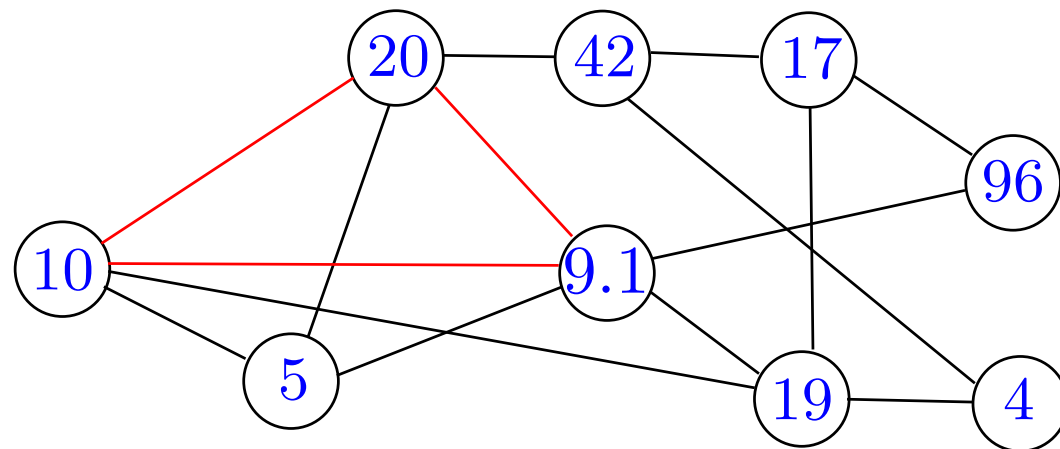
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Check if $(A \wedge (A \times A)) \neq 0$

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Their paper ends with:

“A related problem is finding a minimum weighted circuit in a weighted graph. It is unclear to us whether our methods can be modified to answer this problem too.”

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1. Push weights from nodes to edges: $w(u, v) = (w(u) + w(v))/2$

(Reduce Node-Weighted Triangle to Edge-Weighted Triangle)

2. Compute $MAX_{i,j} \{((-A) \star (-A))[i, j] - A[i, j]\}$

(Min Weight Triangle: $MIN_{i,j} \{(A \star A)[i, j] + A[i, j]\}$)

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Truly Sub-Cubic Algorithm?

Talk Outline

- Deterministic Algorithm

$$O(B \cdot n^{(3+\omega)/2}) \leq O(B \cdot n^{2.688}), \text{ where } B \text{ is the bit precision}$$

- Randomized (Strongly Polynomial) Algorithm

$$O(n^{(3+\omega)/2} \log n) \leq O(n^{2.688})$$

- Some Applications

Deterministic Algorithm

Key Steps:

- Suffices to check if there's a triangle of weight $\geq K$
- Compute a matrix A' s.t.

$$A'[i, j] = |\{k : i \rightarrow k \rightarrow j, w(i) + w(k) + w(j) \geq K\}|.$$

- Check if $\exists i, j$ where $A[i, j]$ and $A'[i, j]$ are non-zero.

Computing C

Def. The **dominance product** of A and B is the matrix C s.t.

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Thm. If Dominance Product is in $O(f(n))$ time, then can check if there's a triangle of weight $\geq K$, in $O(f(n) + n^2)$ time.

(Virginia will prove this)

Dominance Product in $n^{(3+\omega)/2}$

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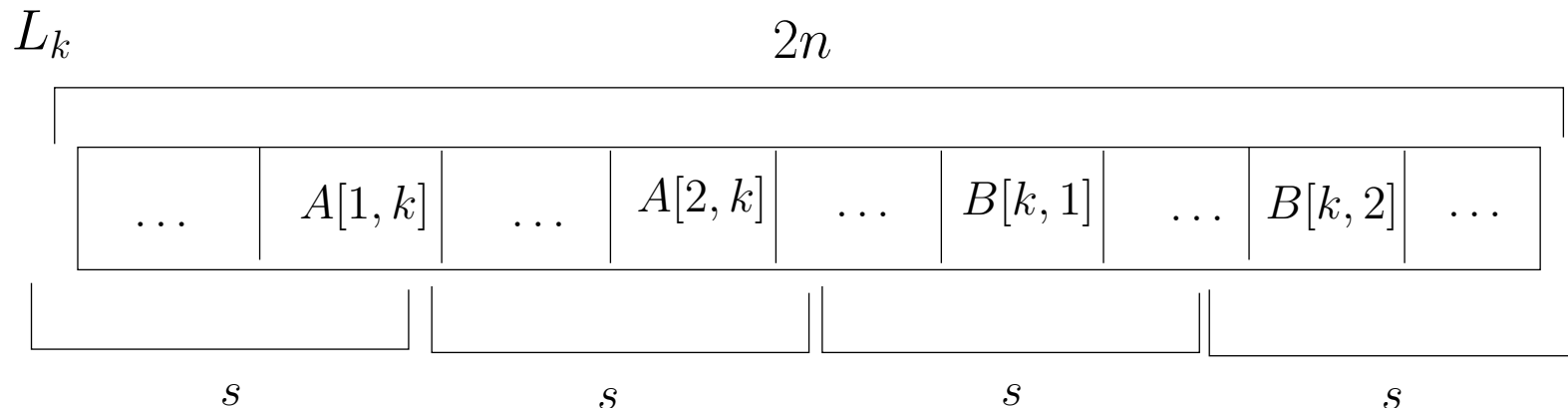
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Partition each L_i into “buckets” with s elements in each bucket
(roughly $2n/s$ buckets in total)



Dominance Product in $n^{(3+\omega)/2}$, Cont.

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1. Pairs $(A[i, k], B[k, j])$ such that $A[i, k] \leq B[k, j]$, but $A[i, k]$ and $B[k, j]$ fall in **the same** bucket of L_k
 - Only $O(n^2s)$ possible pairs of this form
 - Can compute these in $O(1)$ amortized time

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2. Pairs $(A[i, k], B[k, j])$ such that $A[i, k] \leq B[k, j]$, but $A[i, k]$ and $B[k, j]$ fall in **different** buckets of L_k
 - Can count these using $2n/s$ matrix multiplications
(One matrix multiply for each bucket)

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2. Do binary search on K to find the maximum weight W of a triangle.
3. Find a triangle of weight W .

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- row vector $A[i, ;] = (A[i, 1], \dots, A[i, n])$ s.t.

$$A[i, j] = \begin{cases} K - w(i) & \text{if there is an edge from } i \text{ to } j \\ \infty & \text{otherwise.} \end{cases}$$

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$$A[i, j] \leq B[j, k] \iff K \leq w(i) + w(k) + w(j) \text{ and } (i, j), (j, k) \in E$$

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But this algorithm is NOT **strongly polynomial** because of the binary search.

A Strongly Polynomial Randomized Algorithm: Outline

1. show how to **sample** a triangle of weight in any interval $[W_1, W_2]$ efficiently **uniformly** at random
2. search using the weights of triangles chosen at random

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- the number E_{ij}^1 of k such that

$(\dots, w(i) + w(k), \dots)$ dominated in coord. k by $(\dots, W_2 - w(j), \dots)$

$i \rightarrow k \rightarrow j$ and $w(i) + w(j) + w(k) \leq W_2$

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- the number E_{ij}^2 of k such that

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- the number E_{ij}^2 of k such that

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- the number $E_{ij} = (E_{ij}^1 - E_{ij}^2)$ of k such that

$i \rightarrow k \rightarrow j$ and $W_1 \leq w(i) + w(j) + w(k) \leq W_2$

Getting a uniform random triangle in $[W_1, W_2]$

Recall: E_{ij} is the number of k such that $i \rightarrow k \rightarrow j$ and

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$f = \sum_{(i,j) \in E} E_{ij}$ is $3 \times$ [number of triangles in $[W_1, W_2]$].

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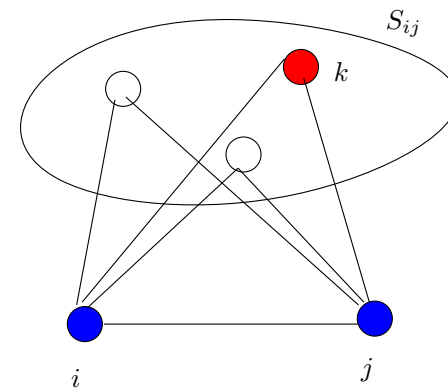
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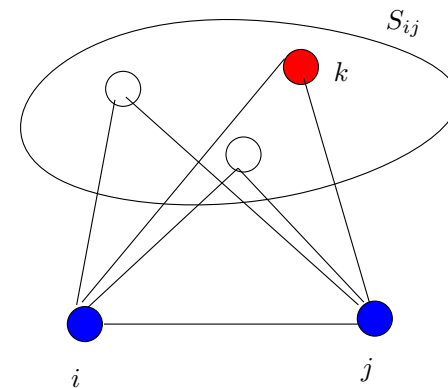
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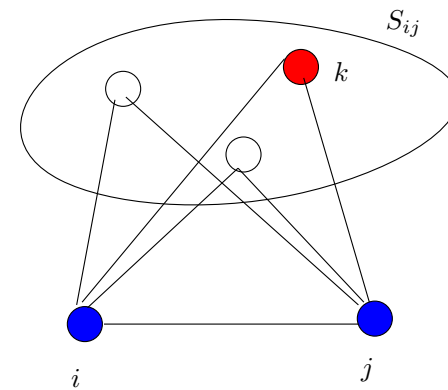
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$\{i, j, k\}$ is a random triangle with weight in $[W_1, W_2]$.

Strongly Polynomial Algorithm

1. Let $M = 3 \cdot \max_{i \in V} w(i)$ and $K = 0$.
2. Pick a random triangle T in $[K, M]$. Set K to its weight.
3. Check if there exists a triangle of weight $> K$. If not, return T .
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The algorithm will terminate in $O(n^{\frac{3+\omega}{2}} \log n)$ expected worst case time.

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- generalized K -SUM
- computing K most significant bits of distance product in $O(2^K \cdot n^{\frac{3+\omega}{2}} \log W \log n) \dots$

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- **Conjecture:**
Dominance product can be computed in $O(n^{\omega+o(1)})$ time.

Thank You!