All Pairs Bottleneck Paths in Truly Subcubic Time

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STOC

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joint work with Ryan Williams and Raphael Yuster

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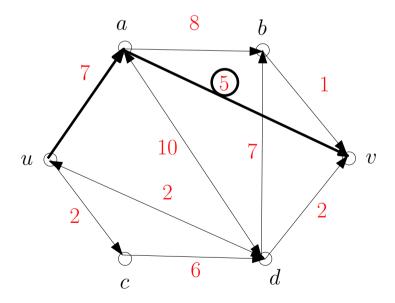
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This talk: truly subcubic algorithm for APBP – studied alongside APSP.

Bottleneck paths - definitions

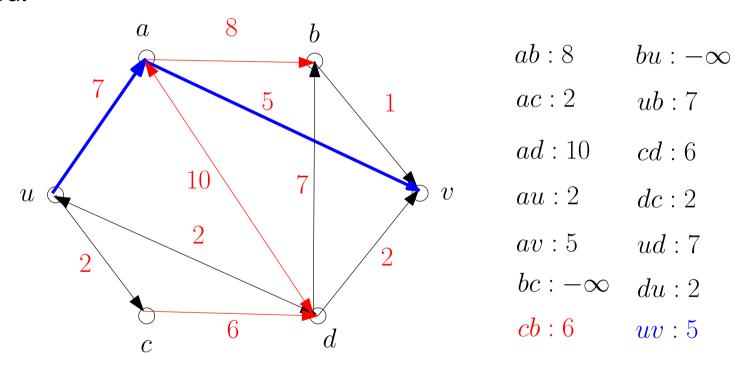
Given: graph G=(V,E) with arbitrary edge weights $w:E\to\mathbb{R}$.

The bottleneck edge of a path in G from vertex u to vertex v is the edge of smallest weight on the path.



Maximum bottleneck paths

In many applications (e.g. max flow), the path of maximum bottleneck is needed.



In this talk we will consider the all pairs max bottlenecks problem: for all pairs of vertices s and t in the graph, find the weight of the maximum bottleneck edge on a path from s to t.

All pairs bottleneck paths – related work

- Pollack 1960: introduced APBP and showed a cubic algorithm.
- Hu 1961: undirected, edge weighted max spanning tree. $\leftarrow O(n^2)$
- Shapira, Yuster, Zwick 2007: directed, node weighted in $O(n^{2.58})$.
- this work: directed, edge weighted in $O(n^{2.79})$.

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Adjacency matrix for weighted graph G = (V, E, w):

$$A[i,j]=w_{ij}$$
, $w_{ii}=\infty$, $w(i,j)=-\infty$ if $(i,j)\notin E$.

 $(A \bullet A)[i,j]$ is the maximum bottleneck edge weight over all paths of length ≤ 2 from i to j.

 $\underbrace{A \bullet A \bullet \ldots \bullet A}_{\text{times}}$: the maximum bottleneck weights for all vertex pairs.

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This work: first truly subcubic algorithm for the MaxMin product.

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- 3. set for all i, j, $C[i, j] = \max\{a_{ij}, b_{ij}\}$.

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We want
$$a_{ij} = \max_{k} \{A[i, k] \mid A[i, k] \leq B[k, j]\}.$$

- 1. Take the rows of A and sort the entries of each row.
- 2. Bucket the entries of each row of A, in their sorted order into s roughly equal buckets.

3. For each bucket b create a matrix A(b) containing only the elements in bucket b and ∞ in all other entries.

$$A(1) = \left(egin{array}{cccccc} \infty & -1.1 & \infty & 3.2 \\ 2 & \infty & \infty & 1 \\ \infty & \infty & -2 & -3 \\ \infty & 2.1 & \infty & 2.1 \end{array}
ight) \quad A(2) = \left(egin{array}{ccccc} 10 & \infty & 5.1 & \infty \\ \infty & 3 & 7 & \infty \\ 0 & -1 & \infty & \infty \\ 7 & \infty & 4 & \infty \end{array}
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Recall, $(A \otimes B)[i,j] = |\{k : A[i,k] \leq B[k,j]\}|$.

4. Compute $A(b) \otimes B$ for each bucket b.

$$A(2) \otimes A = \begin{pmatrix} 10 & \infty & 5.1 & \infty \\ \infty & 3 & 7 & \infty \\ 0 & -1 & \infty & \infty \\ 7 & \infty & 4 & \infty \end{pmatrix} \otimes \begin{pmatrix} 10 & -1.1 & 5.1 & 3.2 \\ 2 & 3 & 7 & 1 \\ 0 & -1 & -2 & -3 \\ 7 & 2.1 & 4 & 2.1 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 2 & 1 & 2 & 2 \\ 1 & 0 & 0 & 0 \end{pmatrix}$$

This tells us for every bucket b and each i, j, the number of coords k such that A[i, k] is in bucket b and $A[i, k] \leq B[k, j]$.

This step takes $O(sn^{\frac{3+\omega}{2}})$.

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- 6. The overall runtime is maximized for $s=n^{\frac{3-\omega}{4}}$ and the runtime is then $O(n^{\frac{9+\omega}{4}})=O(n^{2.85}).$
- 7. You can do slightly better by using sparse dominance $\rightarrow O(n^{2.79})$.

Sparse dominance

Theorem: Let A and B be $n\times n$ matrices with entries from a totally ordered set. Let $S\subseteq [n]\times [n]$ such that $|S|=m\geq n$. Let C be the matrix such that

$$C[i,j] = |\{k \mid (i,k) \in S \text{ and } A[i,k] \le B[k,j]\}|.$$

There is an algorithm that, given A, B, and S, outputs C in $O(\sqrt{m} \cdot n^{\frac{1+\omega}{2}})$ time.

Intuition: The set S of coordinate pairs contains all entries of A we care about. Comparisons between entries of A not in S and entries of B are ignored.

Recall the matrices A(b):

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A(b) has $O(n^2/s)$ finite entries. Each of the s dominance products thus takes $O(n^{\frac{3+\omega}{2}}/\sqrt{s})$, and the running time for the entire algorithm is: $O(n^3/s + \sqrt{s}n^{\frac{3+\omega}{2}})$, minimized for $s = n^{1-\omega/3}$.

Corollary: The MaxMin product of $n\times n$ matrices A and B, and hence APBP can be computed in $O(n^{2+\frac{\omega}{3}})=O(n^{2.79})$.

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Open Problems

- 1. dominance product, MaxMin product in $O(n^{\omega})$?
- 2. truly subcubic distance product using dominance product?

Thank You!