

Last lecture: showed the parallel LDD algorithm of MPX'13:

Computes a $(\beta, O(\frac{\log n}{\beta}))$ -decomposition for $0 < \beta < 1$ in

$O(n+m)$ expected work, $O(\log^2 n / \beta)$ depth whp

$\Rightarrow \beta m$ edges cut in expectation

\Rightarrow Strong diameter of $O(\frac{\log n}{\beta})$ whp.

This lecture: use LDD to solve other graph problems.

Graph Connectivity

Input: Undirected graph $G(V, E)$

Output: Connectivity labeling, L .

i.e. $L[u] = L[v]$ iff u, v in the same component in G
path between u and v

• Sequentially, easy to solve in $O(n+m)$ work using graph search (BFS/DFS).

[Can we get a work-efficient, poly-log depth connectivity alg.?

Lot of work on this problem in early days of parallel algorithms.

Culminated in a randomized work-efficient algorithm with $O(\log n)$ depth on PRAM using $O((m+n)/\log(n))$ processors, i.e. optimal PRAM alg. But the algorithm is extremely complicated and impractical. \Rightarrow not suitable to implement.

Several other work-efficient poly-log depth algorithms since then,

- Halperin/Zwick, Poon-Ramachandran, Cole-Klein-Tarjan

- some based on linear-work MST (will see this later).

Usually use sampling/filtering of edges for work-efficiency and need pretty complex arguments to prove work-efficiency.

This lecture: very simple LDD-based alg that is almost optimal.

LDD-Based Connectivity:

func CC(G, β):

$L = \text{LDD}(G, \beta)$

$G'(V', E') = \text{Contract}(G, L)$

// contract all vertices in an LDD cluster into a single vertex; only keep inter-piece edges and remove any duplicates.

if $|E'| = 0$ then return L

else

$L' = \text{CC}(G', \beta)$

// recurse

$L'' = \{L'[L[u]] \mid u \in V\}$

// propagate label information

Clear that the algorithm is correct.

Set β to a constant > 0 .

• # edges decreases from m to βm in expectation and the rate of reduction is independent across iterations $\Rightarrow O(\log_{1/\beta} m)$ whp.

• Each recursive call is $O(\log^2 n / \beta)$ depth, $O(m')$ work where $m' = \# \text{inter-component edges}$

$\Rightarrow \text{Depth} = O(\log_{1/\beta} m \log^2 n / \beta) = O(\log^3 n)$ whp

$\Rightarrow \text{Work} = \sum_{i=0}^{\infty} \beta^i m = O(m)$ in expectation

Details:

Consider one level of the alg: let m', n' be current m, n

LDD: $O(m' + n')$ work, $O(\log^2 n' / \beta)$ depth whp

Contract:

- filtering out intra-cluster: $O(m')$ work, $O(\log n')$ depth

- removing duplicates: $O(m')$ expected work, $O(\log n')$ depth whp using semisort

Relabeling: $O(n')$ work, $O(\log n')$ depth

Overall: $O(m' + n')$ expected work, $O(\log^2 n' / \beta)$ depth whp. (in a single level).

Is this algorithm practical? Yes!

⇒ Show/analyze code in WBS.

Spanner: sparse subgraphs that approximate distances in the original graph.

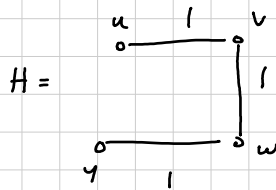
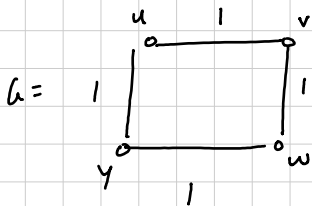
Let $G(V, E)$ be undirected, unweighted.

A subgraph H of G is a k -spanner if $\forall u, v \in V$

$$\text{dist}_H(u, v) \leq k \cdot \text{dist}_G(u, v)$$

k is called the "stretch" factor.

Example: the mesh graph M_n , eg. M_4 below:



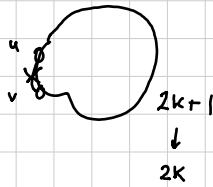
$$\text{Stretch}(H) = 3 \text{ since } \text{dist}_H(u, y) = 3 \leq 3 \cdot \text{dist}_G(u, y) = 3$$

To show a subgraph $H(V, E')$ is a spanner, sufficient to show stretch bounds for every edge $e \in E \setminus E'$.

Thorp-Zwick: $\forall k \geq 1$ any undirected graph on n vertices admits $(2k-1)$ -spanners with $O(n^{1+1/k})$ edges. This is essentially as good as possible. (conditional on ETH, below):

- $\gamma(n, k) = \max \# \text{ edges in } n\text{-vertex graph with } \text{girth} > k$
- Unweighted, undirected graphs of $\text{girth} > t+1$ do not have proper subgraphs that are t -spanners.

If one removes an edge in such a graph, the distance goes from $1 \rightarrow t'$
 $t' > t$
⇒ spanner prop. not satisfied



$$\text{dist}_H(u, v) \leq 2k \text{ dist}_G(u, v)$$

There is a $(4k+1)$ -spanner with $O(n^{1+1/k})$ edges that we can construct in $\text{--- work --- depth}$.

Spanner $(G(V, E), k)$

$$\beta \leftarrow \log\left(\frac{n}{2k}\right)$$

$$L \leftarrow \text{LDD}(G, \beta)$$

$$H \leftarrow \emptyset$$

- Add all BFS tree edges used in the LDD algorithm to H .
- For each vertex $v \in V$ where v is a boundary vertex, add one edge from v to each adjacent cluster, to H .

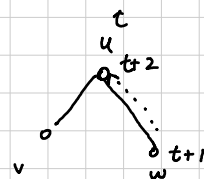
return H .

Recall our lemma from last class:

Lemma
 For any vertex or midpoint of an edge, the prob. that the smallest and k -th smallest value from $\{T_{v+1} d(v, u) : v \in V\}$ differ by $\leq a$ is $\leq (a\beta)^{k-1}$

Lemma: In $\text{LDD}(G, \beta = \frac{\log n}{2k})$, any vertex intersects $O(n^{1/k})$ other clusters in expectation.
 $B(v, 1)$

Consider $u \in V$. $u \xrightarrow{t \dots t+1} v$



$$(2\beta)^{k-1} = \left(\frac{\log n}{k}\right)^{k-1}$$

Need to set $a=2$. Plugging $\beta = \frac{\log n}{2k}$

Let $L = \#$ intersecting clusters. We have $E[L] = \sum_{k=1}^{\infty} \Pr[L \geq k]$

$$E[L] = \sum_{k=1}^{\infty} (1 - e^{-2\beta})^{k-1}$$

$$= \frac{1}{1 - (1 - e^{-2\beta})}$$

$$= e^{2\beta}$$

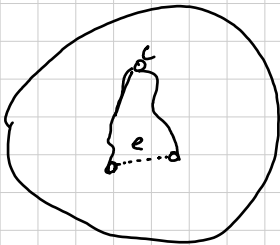
Plugging in $\beta = \frac{\log(n)}{2k}$; $e^{2\beta} = e^{\log(n)/k} = n^{1/k}$

\Rightarrow number of overall intersecting clusters is $O(n^{1+1/k})$ in expectation and $O(n^{1+1/k})$ boundary edges added in expectation.

Stretch?

Consider an edge e . There are a few cases:

(1) e is internal to a cluster.



\exists path through center of this cluster that has length $\leq 2r$ where r is radius of the cluster.

What is r ? $= O\left(\frac{\log n}{\beta}\right) = O(k)$ w.h.p.

$\frac{\log(n)}{\beta}$ in expectation \Rightarrow $2k$ in expectation.

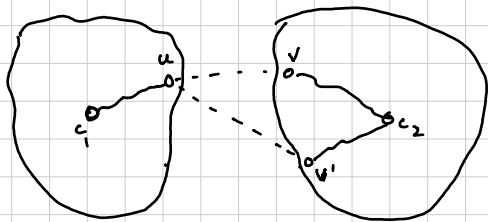
\Rightarrow stretch = $4k$ in expectation.

$X_i = \text{Exp}(\beta)$

$E[X_i] = \frac{1}{\beta}$

Expected value of max = $\ln(n)$

(2) e is inter-cluster



• if (u, v) added as the single inter-cluster edge between c_1 and c_2 , stretch is 1.

• Otherwise, \exists some other (u, v') edge that's added.

$$u \xrightarrow{1} v' \xrightarrow{r} c_2 \xrightarrow{r} v$$
 This path has length $4k+1$ in expectation
($r = 2k$ exp.)

Overall stretch is $4k+1$ in expectation.

Overall algorithm runs in $O(mn)$ work, $O(\log^2 n/p)$ depth whp

$O(k \log(n))$ depth whp.

