Distributed Triangle Detection via Graph Partitioning

Yi-Jun Chang (Michigan) Seth Pettie (Michigan) Hengjie Zhang (Tsinghua)

Distributed Message-passing Models

• LOCAL:

• CONGEST:

Can only communicate with neighbors. Unbounded message size.

Can only communicate with neighbors. $O(\log n)$ -bit message size.

• Congested Clique:

Allow all-to-all communication. $O(\log n)$ -bit message size.

Bandwidth constraint

Effect of Locality

How different are these two models?

• CONGEST:

Locality + Bandwidth constraint.

• Congested Clique:

Bandwidth constraint only.

Effect of Locality

Minimum Spanning Tree

• CONGEST:

 $\Theta(D + \sqrt{n})$ (well-known)

• Congested Clique:

Θ(1) (Jurdziński & Nowicki, SODA 2018)

Algorithms in Almost Mixing Time

Mixing time of lazy random walk

Minimum Spanning Tree

• CONGEST:

 $O(\tau_{\text{mix}}) \cdot n^{o(1)}$ (Ghaffari, Kuhn, and Su, PODC 2017)

• Congested Clique:

Θ(1) (Jurdziński & Nowicki, SODA 2018)

Algorithms in Almost Mixing Time

Mixing time of lazy random walk

Minimum Spanning Tree

• CONGEST:

 $O(\tau_{\text{mix}}) \cdot n^{o(1)}$ (Ghaffari, Kuhn, and Su, PODC 2017)

Congested clique algorithms can be simulated efficiently on graphs of low-mixing time.

Beyond Graphs of Low Mixing Time?

[Main question] Can we extend this technique to a broader graph class, or even general graphs?

Congested clique algorithms can be simulated efficiently on graphs of low-mixing time.

Graph Partition

Any graph can be decomposed into connected components of low mixing time after removing a small constant fraction of edges.

(Spielman and Teng, STOC 2004) (Trevisan, FOCS 2005) (Arora, Barak, and Steurer, FOCS 2010)

Our Results

• A variant of this graph decomposition can be constructed efficiently in CONGEST.

• Based on this decomposition, we obtain improved upper bounds for some problems in CONGEST.

Previous state-of-the-art (Izumi & Le Gall, PODC 2017)

Triangle detection:

Triangle listing:

$$O(n^{2/3})$$

 $O(n^{3/4})$

Our results

 $\overline{O(n^{1-1/\omega})} < \overline{O(n^{0.58})}$ $O(n^{2/3})$

