## Distributed Bipartite Matching

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## Bipartite Matching



- Matching (M) is set of edges such that $E(u, v)$
- Vertices incident to only one edge in M


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## DMBM: Kernel Cont.



- Partition Graph based on one set of vertices
- Distribute vertices and associated edge lists to processes
- Better partitioning can be created but at greater cost

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## DMBM: Kernel


$|M|=4$

- Updates must be when vertex is matched
- Similar to Push-Relabel
- Lots of communication required


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## DMBM: Kernel Cont.

Given a graph $G\left(V(u, v), E\left(u_{i}, v_{j}\right)\right)$ :

```
bool augment path(uint uid) {
    visited[uid] = true;
    for (uint i = 0; i < graph[uid].size(); i++) {
        uint neighbour = graph[uid][i];
        if (visited[neighbour]) {
            continue;
        }
        // Base-case. We've reached a node at the end of an alternating path that
        // ends in a freenode
        if (matched[neighbour] == UNMATCHED) {
            matched[uid] = neighbour;
            matched[neighbour] = uid;
            return true;
        } else if (matched[neighbour] != uid) {
            // This is not your standard DFS. Because we're DFSing along an
            // alternating path, when we choose the next vertex to visit, we MUST
            // then go along its matching edge. So we say we've visited the neighbour
            // trivially and then recursing on matched[neighbour].
            visited[neighbour] = true;
            if (augment path(matched[neighbour])) {
                matched[uid] = neighbour;
                    matched[neighbour] = uid;
                    return true;
            }
        }
    }
    return false;
```


## DMBM: Kernel Cont.

Given a graph $G\left(V(u, v), E\left(u_{i}, v_{j}\right)\right)$, and process count $P$ :
distribute vertex and edge list assignments
omp for $\mathrm{i}<\mathrm{u}_{\mathrm{p}}$.size() do
augmentPath( $u_{i}$ )
if $E\left(v_{j}\right)$ contains $u_{i}$ st $u_{i} \notin u_{p}$ notify $P_{k}$ assigned $u_{i}$ of $v_{j}$ visitation
if notified ( $v_{j}$ )

$$
\text { augmentPath }\left(v_{j}\right)
$$

gather matchings
end

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## Problem 1: Partitioning



- If Scale Free graphs are used, the partitioning can become highly skewed
- Imbalance causes communication hotspots
- Concerned with vertex count AND edge count


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## Problem 2: Communication



- Communication at core of compute phase
- Message volume and interconnect becomes dominant factor
- Does NOT scale well!!


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## DMBM: Time Complexity

- Ford-Fulkerson: $\mathrm{O}\left(\mathrm{VE}^{2}\right)$
- Hopcroft-Karp: O(|E| $\sqrt{ }(\mathrm{V}))$
- Distributed $\mathrm{MBM}: \mathrm{O}\left(\mathrm{O}\left(\mathrm{VE}^{2}\right)+\mathrm{Vlg}(\mathrm{P})\right)$
- not $100 \%$ on this...

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## DMBM: Data Sets

Suite Sparse Matrix Collection https://sparse.tamu.edu

Largest undirected biparite graph:

- $12,471 \times 872,622$ (885,093 total vertices)
- 22,624,727 edges

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## DMBM: Present and Future

## Presently:

- It works!
- Performance is abysmal

Future:

- Implement Hopcroft-Karp to see if communication is reduced
- HavoqGT vertex-centric framework
- Communication may be unavoidable

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## DMBM: Ray of Hope



## Ariful Azad, Aydin Buluc (LBNL)

- Sparse algebra based Distributed MCM
- SpMV plays significant role


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## DMBM: Ray of Hope Cont.






Fig. 5: Runtime breakdown of MCM-DIST for four representative graphs using 12 threads per MPI process on Edison.


Fig. 6: Strong scaling of MCM-DIST when computing maximum matching on three classes of randomly generated graphs with five different scales on Edison.

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