SOMETHING WITH CONNECTED COMPONENTS

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ABSTRACT

1. INTRODUCTION

Motivation.

Related work.

Connected components.

For an unordered graph $G = (V, E)$, the connected components are the ensemble of connected subgraphs, where connected means that for any two vertices, there exist a path along the edges connecting them. The straightforward algorithm to find them is to perform either a breath or depth first search from a starting random vertex in V, and give the same label to all the touched vertices. Then repeat the search from an unlabelled vertex until there are no more left. This has a cost in terms of memory accesses of $\Theta(|E|+|V|)$, which turns to be optimal [**?**].

2. PROPOSED ALGORITHM

Unfortunately this algorithm does not parallelize straightforwardly. Instead we firstly implemented an algorithm proposed by Uzi Vishkin [**?**] and later described in a class by Pavel Tvrdik [**?**]. This algorithm casts the problem in terms of the generation of a forest, where the vertices of the same connected component belong to the same three, and its root can be used as the representative. We define a star as a three of height one, a singleton a tree with a single element, and use the variables $n = |V|$ and $m = |E|$. His algorithm can be summarized as:

Algorithm 1 Pavel Tvrdik's Connected components

```
1:  procedure HOOK(i, j)
2:      p[p[i]] = p[j]
3:  end procedure
4:  procedure CONNECTEDCOMPONENTS(n, edges)
5:      p[i] = i   ∀i ∈ {1, ··· , n}.        ▷ Initialize a list of parents.
6:      while Elements of p are changed. do
7:          for ⟨i, j⟩ ∈ edges do            ▷ Execute in parallel.
8:              if i ≥ j then Hook(i, j)
9:              if isSingleton(i) then Hook(i, j)
10:         end for
11:         for ⟨i, j⟩ ∈ edges do            ▷ Execute in parallel.
12:             if isStar(i) and i ≠ j then Hook(i, j)
13:         end for
14:         p[i] = root(i)   ∀i ∈ {1, ··· , n}   ▷ Compress the forest in parallel.
15:     end while
16: end procedure
```

We defer to [**?**] for a proof of correctness.

After implementing this algorithm we found advantageous to remove the constraint that only singletons and stars can be hooked to another vertex, so that only a single pass through the edge list is required. Extra care is then required during parallel execution: as each vertex has only one outgoing connection, we need to avoid that a process overwrites a connection that has been formed by another one. We therefore need to grow our forest with the following rules:

1. A hook must originate from a vertex id higher than the destination.
2. All edges must generate a connection between the relative vertices, or vertices at an higher level in their three.
3. A hook must originate from a vertex that is currently the root of a tree.

The intuitive proof of correctness follows: rule 1 means that the graph gebnerated by the hooks generate a directed

graph with no cycles and with at most a single outgoing connection, therefore it must be a forest. Rule 2 and 3 enforce that after processing an edge between two nodes, they belong to the same tree, and rule 3 guarantees that this connection can not be broken by a different edge. At the end of the algorithm, by following the connections from each vertex to the root, we can find a representative for each connected component.

To implement rule 3 in a multithreaded environment, we use an atomic compare and swap. We compare the parent of the hook's origin with its id, if they match it means the vertex is still a root and we hook we hook it to its destination. It does not matter for correctness if the destination is a root, but we try withouth enforcing to hook to a root to minimize the three height. We found empirically that using `std::atomic_compare_exchange_weak`, compared to `std::atomic_compare_exchange_strong` offers better performance, as we anyway need to loop until a hook is successful.

In pseudocode our algorithm is:

Algorithm 2 Single pass connected component.

```
1:  procedure CONNECTEDCOMPONENTS(n, edges)
2:      p[i] = i   ∀i ∈ {1, ··· , n}.
3:      for ⟨i, j⟩ ∈ edges do            ▷ Execute in parallel.
4:          while hook is not successful. do
5:              from = max(root(i), root(j))
6:              to = mint(root(i), root(j))
7:              atomicHook(from, to)
8:          end while
9:          if !isRoot(i) then p[i] = root(i)
10:         if !isRoot(j) then p[j] = root(j)
11:     end for
12:     p[i] = root(i)   ∀i ∈ {1, ··· , n}    ▷ Compress the
        forest in parallel.
13: end procedure
```

While step 9 is not necessary for correctness, we found that reusing the already computed vertex's representative leads to a smaller tree height. This and the parallel compression works and was tested to be efficient only on architectures such as x86, where writes to 32 or 64-bits , used to store a vertex's id, are atomic.

We tried implementing the parallel execution of loop 3 with Boost fibers [**?**] whose execution is scheduled with a work stealing algorithm, and OpenMP with a dynamic scheduler. OpenMP performed better by a arge margin and will be used to acquire the data presented later.

The overall cost of the algorithm is $\Theta((n + m)\langle H\rangle)$, where $\langle H\rangle$ is the average tree height. Theefore $\langle H\rangle = \Theta(1)$ for a subcritical random graph, and on average (relatively to the execution order of the loop) $\langle H\rangle = \Theta(\log(n))$ for a supercritical one [**?**].

Multiple compute nodes. Algorithm works only on a single compute node with a shared memory model. Moreover it is efficient only when the graph is relatively sparse so that the chance of a collision between two processors trying to update the same parent is low.

We propose to extend our algorithm by distributing the list of egdes evenly among each MPI process, then each one of them computes a forest used only the subset of edges it received. This local computation is followed by a reduction step, where the list of representatives is sent to another process, which confront it with its own. If a discrepancy is detected, a hook is inserted between the two different parents, then the resulting forest is compressed again before the following reduction step.

Using p processes, the total execution time of this extension scales as $\Theta(\frac{(m+n)}{p}\langle H\rangle + n\log p)$.

On top of allowing to scale past a single compute node, this approach is advantageous on dense graphs: if the reduction cost is negligible, the scaling is the same as algorithm 2 executed on a single node, but we can avoid the cost of performing atomic hooks, if a single thread is used, or limit the number of failures if a few threads are used. Therefore a different mixture of MPI ranks and OpenMP threads per rank is advised depending on the density of the graph.

Distributed vertices. While the described approach works on generic graphs, it performs poorly on very sparse graphs using a large number of compute nodes. Morehover the full set of vertices' id must fit in memory, limiting the graph size to 8 billions vertices. If the connectivity of a graph the size of a human brain needs to be studied, we propose do distribute the representation of the vertices as well.

Often, real word graphs are embedded on a space with some metric, and connections are present much more frequently between vertices that are close together. For examples the pixel representing features of a picture, or the roads connecting cities with a known geographical position, posses this property.

We represent this type of graphs with a very simple model: a two dimensional lattice with random connections between nearest neighbours only. We split the lattice in as many square tiles as there are processes. Then each process applies algorithm 2 with the subset of edges connecting two vertices in their own tile. Finally we process the boundary edges, connecting vertices of different tiles, with MPI one sided communication. The list of ids of the local vertices is stored in an MPI window, so that the representative of a remote vertex can be obtained with `MPI_Get`, while an hook can be created with `MPI_Compare_and_swap`. Therefore only two global synchronization points are necessary: after all edges have been processed, and after the final compression of the forest.

Unfortunately, due to time constraint in developing, in our implementation each MPI operation is synchronized lo-

cally. This leads to good scaling results only on extreemly sparse graphs. Future work should consider batching several MPI requests before synchronization is required.

3. EXPERIMENTAL RESULTS

Results.

4. CONCLUSIONS

5. FURTHER COMMENTS

Here we provide some further tips.