Chapter 3

Tree Algorithms

In this chapter, we learn a few basic algorithms on trees, and how to construct trees in the first place so that we can run these (and other) algorithms. The good news is that these algorithms have many applications, the bad news is that this chapter is a bit on the simple side. But maybe that’s not really bad news?!

3.1 Broadcast

Definition 3.1 (Broadcast) A broadcast operation is initiated by a single processor, the source. The source wants to send a message to all other nodes in the system.

Definition 3.2 (Distance, Radius, Diameter) The distance between two nodes \( u \) and \( v \) in an undirected graph \( G \) is the number of hops of a minimum path between \( u \) and \( v \). The radius of a node \( u \) is the maximum distance between \( u \) and any other node in the graph. The radius of a graph is the minimum radius of any node in the graph. The diameter of a graph is the maximum distance between two arbitrary nodes.

Remarks:

- Clearly there is a close relation between the radius \( R \) and the diameter \( D \) of a graph, such as \( R \leq D \leq 2R \).

- The world is often fascinated by graphs with a small radius. For example, movie fanatics study the who-acted-with-whom-in-the-same-movie graph. For this graph it has long been believed that the actor Kevin Bacon has a particularly small radius. The number of hops from Bacon even got a name, the Bacon Number. In the meantime, however, it has been shown that there are “better” centers in the Hollywood universe, such as Sean Connery, Christopher Lee, Rod Steiger, Gene Hackman, or Michael Caine. The center of other social networks has also been explored, Paul Erdös for instance is well known in the math community.

Theorem 3.3 (Broadcast Lower Bound) The message complexity of broadcast is at least \( n-1 \). The source’s radius is a lower bound for the time complexity.
Proof: Every node must receive the message.

Remark:
- You can use a pre-computed spanning tree to do broadcast with tight message complexity. If the spanning tree is a breadth-first search spanning tree (for a given source), then the time complexity is tight as well.

**Definition 3.4 (Clean)** A graph (network) is **clean** if the nodes do not know the topology of the graph.

**Theorem 3.5 (Clean Broadcast Lower Bound)** For a clean network, the number of edges is a lower bound for the broadcast message complexity.

Proof: If you do not try every edge, you might miss a whole part of the graph behind it.

Remark:
- This lower bound proof directly brings us to the well known flooding algorithm.

**Algorithm 11 Flooding**

1. The source (root) sends the message to all neighbors.
2. Each other node \( v \) upon receiving the message the first time forwards the message to all (other) neighbors.
3. Upon later receiving the message again (over other edges), a node can discard the message.

Remarks:
- If node \( v \) receives the message first from node \( u \), then node \( v \) calls node \( u \) parent. This parent relation defines a spanning tree \( T \). If the flooding algorithm is executed in a synchronous system, then \( T \) is a breadth-first search spanning tree (with respect to the root).
- More interestingly, also in asynchronous systems the flooding algorithm terminates after \( R \) time units, \( R \) being the radius of the source. However, the constructed spanning tree may not be a breadth-first search spanning tree.

### 3.2 Convergecast

Convergecast is the same as broadcast, just reversed: Instead of a root sending a message to all other nodes, all other nodes send information to a root. The simplest convergecast algorithm is the echo algorithm:
3.3 BFS Tree Construction

In synchronous systems the flooding algorithm is a simple yet efficient method to construct a breadth-first search (BFS) spanning tree. However, in asynchronous systems the spanning tree constructed by the flooding algorithm may be far from BFS. In this section, we implement two classic BFS constructions—Dijkstra and Bellman-Ford—as asynchronous algorithms.

We start with the Dijkstra algorithm. The basic idea is to always add the “closest” node to the existing part of the BFS tree. We need to parallelize this idea by developing the BFS tree layer by layer:

Theorem 3.6 (Analysis of Algorithm 13) The time complexity of Algorithm 13 is $O(D^2)$, the message complexity is $O(m + nD)$, where $D$ is the diameter of the graph, $n$ the number of nodes, and $m$ the number of edges.

Proof: A broadcast/echo algorithm in $T_p$ needs at most time $2D$. Finding new neighbors at the leaves costs time 2. Since the BFS tree height is bounded by the diameter we have $D$ phases, giving a total time complexity of $O(D^2)$. 

Algorithm 12 Echo

Require: This algorithm is initiated at the leaves.
1: A leave sends a message to its parent.
2: If an inner node has received a message from each child, it sends a message to the parent.

Remarks:

- Usually the echo algorithm is paired with the flooding algorithm, which is used to let the leaves know that they should start the echo process; this is known as flooding/echo.
- One can use convergecast for termination detection, for example. If a root wants to know whether all nodes in the system have finished some task, it initiates a flooding/echo; the message in the echo algorithm then means “This subtree has finished the task.”
- Message complexity of the echo algorithm is $n - 1$, but together with flooding it is $O(m)$, where $m = |E|$ is the number of edges in the graph.
- The time complexity of the echo algorithm is determined by the radius of the spanning tree generated by the flooding algorithm.
- The flooding/echo algorithm can do much more than collecting acknowledgements from subtrees. One can for instance use it to compute the number of nodes in the system, or the maximum ID (for leader election??), or the sum of all values stored in the system, or a route-disjoint matching.
- Moreover, by combining results one can compute even fancier aggregations, e.g., with the number of nodes and the sum one can compute the average. With the average one can compute the standard deviation. And so on . . .
Algorithm 13 Dijkstra BFS

1. The algorithm proceeds in phases. In phase \( p \) the nodes with distance \( p \) to the root are detected. Let \( T_p \) be the tree in phase \( p \). We start with \( T_1 \) which is the root plus all direct neighbors of the root. We start with phase \( p = 1 \):

2. repeat

3. The root starts phase \( p \) by broadcasting “start \( p \)” within \( T_p \).

4. When receiving “start \( p \)” a leaf node \( u \) of \( T_p \) (that is, a node that was newly discovered in the last phase) sends a “join \( p + 1 \)” message to all quiet neighbors. (A neighbor \( v \) is quiet if \( u \) has not yet “talked” to \( v \).)

5. A node \( v \) receiving the first “join \( p + 1 \)” message replies with “ACK” and becomes a leave of the tree \( T_{p+1} \).

6. A node \( v \) receiving any further “join” message replies with “NACK”.

7. The leaves of \( T_p \) collect all the answers of their neighbors; then the leaves start an echo algorithm back to the root.

8. When the echo process terminates at the root, the root increments the phase

9. until there was no new node detected

Each node participating in broadcast/echo only receives (broadcast) at most 1 message and sends (echo) at most once. Since there are \( D \) phases, the cost is bounded by \( O(nD) \). On each edge there are at most 2 “join” messages. Replies to a “join” request are answered by 1 “ACK” or “NACK”, which means that we have at most 4 additional messages per edge. Therefore the message complexity is \( O(m + nD) \).

Remark:

- The time complexity is not very exciting, so let’s try Bellman-Ford!

The basic idea of Bellman-Ford is even simpler, and heavily used in the Internet, as it is a basic version of the omnipresent border gateway protocol (BGP). The idea is to simply keep the distance to the root accurate. If a neighbor has found a better route to the root, a node might also need to update its distance.

Algorithm 14 Bellman-Ford BFS

1. Each node \( u \) stores an integer \( d_u \) which corresponds to the distance from \( u \) to the root. Initially \( d_{\text{root}} = 0 \), and \( d_u = \infty \) for every other node \( u \).

2. The root starts the algorithm by sending “1” to all neighbors.

3. if a node \( u \) receives a message “\( y \)” with \( y < d_u \) from a neighbor \( v \) then

4. node \( u \) sets \( d_u := y \)

5. node \( u \) sends “\( y + 1 \)” to all neighbors (except \( v \))

6. end if

Theorem 3.7 (Analysis of Algorithm 14) The time complexity of Algorithm 14 is \( O(D) \), the message complexity is \( O(nm) \), where \( D, n, m \) are defined as in Theorem 3.6.
3.4. MST TREE CONSTRUCTION

Proof: We can prove the time complexity by induction. We claim that a node at distance $d$ from the root has received a message ‘$d$’ by time $d$. The root knows by time 0 that it is the root. A node $v$ at distance $d$ has a neighbor $u$ at distance $d - 1$. Node $u$ by induction sends a message ‘$d$’ to $v$ at time $d - 1$ or before, which is then received by $v$ at time $d$ or before. Message complexity is easier: A node can reduce its distance at most $n - 1$ times; each of these times it sends a message to all its neighbors. If all nodes do this we have $O(nm)$ messages.

Remarks:

• Algorithm 13 has the better message complexity and Algorithm 14 has the better time complexity. The currently best algorithm (optimizing both) needs $O(m + n \log^2 n)$ messages and $O(D \log^3 n)$ time. This “trade-off” algorithm is beyond the scope of this course.

3.4 MST Tree Construction

There are several types of spanning trees, each serving a different purpose. A particularly interesting spanning tree is the minimum spanning tree (MST). The MST only makes sense on weighted graphs, hence in this section we assume that each edge $e$ is assigned a weight $\omega_e$.

Definition 3.8 (MST) Given a weighted graph $G = (V, E, \omega)$. The MST of $G$ is a spanning tree $T$ minimizing $\omega(T)$, where $\omega(G') = \sum_{e \in G'} \omega_e$ for any subgraph $G' \subseteq G$.

Remarks:

• In the following we assume that no two edges of the graph have the same weight. This simplifies the problem as it makes the MST unique; however, this simplification is not essential as one can always break ties by adding the IDs of adjacent vertices to the weight.

• Obviously we are interested in computing the MST in a distributed way. For this we use a well-known lemma:

Lemma 3.9 For a given graph $G$ let $T$ be an MST, and $T' \subseteq T$ be a subgraph (also known as fragment) of the MST. Edge $e = (u, v)$ is an outgoing edge of $T'$ if $u \in T'$ and $v \notin T'$ (or vice versa). Let the minimum weight outgoing edge of $T'$ be the so-called blue edge $b(T')$. Then $T' \cup b(T') \subseteq T$.

Proof: We simplify the proof by assuming to have distinct edge weights, then the MST $T$ is unique. For the sake of contradiction, suppose that in the MST $T$ there is edge $e \neq b(T')$ connecting $T'$ with the remainder of $T$. Adding the blue edge $b(T')$ to the MST $T$ we get a cycle including both $e$ and $b(T')$. If we remove $e$ from this cycle we still have a spanning tree, and since by the definition of the blue edge $\omega_e > \omega_{b(T')}$, the weight of that new spanning tree is less than the weight of $T$. We have a contradiction.
Remarks:

• In other words, the blue edges seem to be the key to a distributed algorithm for the MST problem. Since every node itself is a fragment of the MST, every node directly has a blue edge! All we need to do is to grow these fragments! Essentially this is a distributed version of Kruskal’s sequential algorithm.

• At any given time the nodes of the graph are partitioned into fragments (rooted subtrees of the MST). Each fragment has a root, the ID of the fragment is the ID of its root. Each node knows its parent and its children in the fragment. The algorithm operates in phases. At the beginning of a phase, nodes know the IDs of the fragments of their neighbor nodes.

Algorithm 15 GHS

1. Initially each node is the root of its own fragment. We proceed in phases:
2. repeat
3. All nodes learn the fragment IDs of their neighbors.
4. The root of each fragment uses flooding/echo in its fragment to determine the blue edge \( b = (u, v) \) of the fragment.
5. The root sends a message to node \( u \); while forwarding the message on the path from the root to node \( u \) all parent-child relations are inverted (such that \( u \) is the new temporary root of the fragment)
6. node \( u \) sends a merge request over the blue edge \( b = (u, v) \).
7. if node \( v \) also sent a merge request over the same blue edge \( b = (v, u) \)
8. then
9. either \( u \) or \( v \) (whichever has the smaller ID) is the new fragment root
10. the blue edge \( b \) is directed accordingly
11. the newly elected root node informs all nodes in its fragment (again using flooding/echo) about its identity
12. else
13. node \( v \) is the new parent of node \( u \)
14. end if
15. until all nodes are in the same fragment

Remark:

• There are a few details that are not really synchronous. For instance, it may be that some fragments are much larger than others, and because of that some nodes may need to wait for others, e.g., if node \( u \) needs to find out whether neighbor \( v \) also wants to merge over the blue edge \( b = (u, v) \). The good news is that all this can be solved; indeed, the whole algorithm does not need to be synchronous in the first place and can be made fully asynchronous. The standard trick to do that is to add additional phase information to all messages, and only allow nodes to proceed to phase \( p + 1 \) if all neighbors are at least in phase \( p \). We will see better examples for this technique later.

Theorem 3.10 (Analysis of Algorithm 15) The time complexity of Algorithm 15 is \( O(n \log n) \), the message complexity is \( O(m \log n) \).
3.4. MST TREE CONSTRUCTION

Proof: Each phase mainly consists of two flooding/echo processes. In general, the cost of flooding/echo on a tree is $O(D)$ time and $O(n)$ messages. However, the diameter $D$ of the fragments may turn out to be not related to the diameter of the graph because the MST may meander, hence it really is $O(n)$ time. In addition in the first step of each phase, nodes need to learn the fragment ID of their neighbors; this can be done in 2 steps but costs $O(m)$ messages. There are a few more steps, but they are cheap. Altogether a phase costs $O(n)$ time and $O(m)$ messages. So we only have to figure out the number of phases: Initially all fragments are single nodes and hence have size 1. In a later phase, each fragment merges with at least one other fragment, that is, the size of the smallest fragment at least doubles. In other words, we have at most $\log n$ phases. The theorem follows directly.

Remarks:

- Algorithm 15 is called “GHS” after Gallager, Humblet, and Spira, three pioneers in distributed computing. Despite being quite simple the algorithm won the prestigious Edsger W. Dijkstra Prize in Distributed Computing in 2004, among other reasons because it was one of the first (1983) non-trivial asynchronous distributed algorithms. As such it can be seen as one of the seeds of this research area.

- We presented a simplified version of the original paper by Gallager et al. that featured an improved message complexity of $O(m + n \log n)$.

- In 1987, Awerbuch managed to further improve the GHS algorithm to get $O(n)$ time and $O(m + n \log n)$ message complexity, both asymptotically optimal.

- The GHS algorithm can be applied in different ways. GHS for instance directly solves leader election in general graphs: The leader is simply the last surviving root!