# Lecture notes for Jan 25, 2023 BFS and applications and DFS 

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## 1 Rooted trees

A rooted tree is a tree $T$ with a special vertex $r \in V(T)$. The vertex $r$ is called the root of $T$.

Let $T$ be a rooted tree with root $r$.

- For a vertex $v \in V(T)$, we say that a vertex $u \in V(T)$ is an ancestor (with respect to $(T, r)$ ) of $v$ if $u$ is in the unique path in $T$ from $r$ to $v$. Note that every vertex is an ancestor of itself.
- A proper ancestor (with respect to $(T, r)$ ) of a vertex $v$ is an ancestor (with respect to $(T, r)$ ) of $v$ distinct from $v$.
- A vertex $u$ is a descendant (with respect to $(T, r)$ ) of $v$ if $v$ is an ancestor (with respect to $(T, r)$ ) of $u$.
- A proper descendant (with respect to $(T, r)$ ) of a vertex $v$ is a descendant (with respect to $(T, r)$ ) of $v$ distinct from $v$.
- If $v$ is not the root, then the parent of $v$ is the unique proper ancestor adjacent to $v$.
- For a vertex $v$, every vertex that is a proper descendant adjacent to $v$ is called a child of $v$.
- Two vertices $u, v$ are incomparable (with respect to $(T, r)$ ) if none of $u, v$ is an ancestor of the other.

Let $G$ be a graph. Let $T$ be a spanning tree of $G$. Let $r$ be a vertex of $G$. When treating $T$ as a rooted tree with root $r$, we say that an edge $e \in E(G)-E(T)$ is

- a back edge if one end of $e$ is a proper ancestor of the other end of $e$, and
- a crossing edge if the ends of $e$ are incomparable with respect to $(T, r)$.


## 2 Distance and breadth-first-search

### 2.1 Distance in graphs

The length of a path is the number of edges of this path. For a graph $G$ and vertices $u, v$ of $G$, we say that a path $P$ in $G$ is a shortest path between $u$ and $v$ if it is a path between $u$ and $v$, and no path between $u$ and $v$ is shorter than $P$.

Proposition 1 Let $G$ be a graph. Let $u, v$ be vertices of $G$. If $P$ is a shortest path between $u$ and $v$, then for every vertex $w$ in $P$, the subpath of $P$ between $u$ and $w$ is a shortest path between $u$ and $w$.

Proof. If there exists a path $Q$ between $u$ and $w$ shorter than the subpath of $P$ between $u$ and $w$, then the union of $Q$ and the subpath of $P$ between $w$ and $v$ is a walk shorter than $P$, a contradiction.

Let $G$ be a graph. The distance between two vertices $u$ and $v$ of $G$, denoted by $d_{G}(u, v)$, is the length of a shortest path in $G$ between $u$ and $v$. (If no such a path exists, the distance is defined to be $\infty$.)

Proposition 2 If $H$ is a subgraph of $G$, then for any vertices $u, v \in V(H) \subseteq$ $V(G), d_{H}(u, v) \geq d_{G}(u, v)$.

Proof. Every path between $u, v$ in $H$ is a path in $G$. So a shortest path between $u, v$ in $H$ is a path in $G$ whose length is at least the length of a shortest path between $u, v$ in $G$.

Note that it is possible $d_{H}(u, v)=d_{G}(u, v)$, and it is possible that $d_{H}(u, v)>$ $d_{G}(u, v)$.

### 2.2 Breadth-first-search

Breadth-first-search is an algorithm that helps us compute the distance between two vertices.
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Breadth-First-Search (BFS)
Input: A connected graph $G$ and a vertex $r$ of $G$.
Output: A tree rooted at $r$.
Procedure:
Step 1: Label $r$ as $v_{1}$. Set $i=1$. Set $T$ to be the graph $(\{r\}, \emptyset)$. (That is, $T$ is the graph consisting of the single vertex $r$ and with no edge.)

Step 2: Repeatedly picking one edge $e$ of $G$ incident with $v_{i}$ and do Step 2-1, until all edges incident with $v_{i}$ have been seen.

Step 2-1: If $e$ has one end in $V(T)$ and one end not in $V(T)$, then adding $e$ into $T$ and label the end of $e$ not in $V(G)$ as $v_{j}$, where $j$ is least positive integer such that no vertex in $V(T)$ has been labelled as $v_{j}$.

Step 3: Set $i$ to be $i+1$. Do Step 2, unless $i>|V(G)|$.
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## Remark:

- It is easy to prove that $T$ is always a tree during the algorithm by induction on $i$. In particular, $T$ is a tree when the algorithm terminates. The final tree $T$ is called a breadth-first-search (BFS) tree rooted at $r$.
- Note that $T$ is a subgraph of $G$. Since $G$ is connected, $V(T)=V(G)$. Hence $T$ is a spanning tree of $G$.
- BFS runs in time $O(|E(G)|)$ since we visit every edge at most twice. When $G$ is disconnected, we can run BFS for each component of $G$ to obtain a spanning forest $F$ in time $O(|V(G)|+|E(G)|)$, where each component of $F$ is a BFS tree of a component of $G$.

Now we prove nice properties of a BFS tree. For simplicity, until the end of this subsection, $G, r, T$ denotes the graph $G$, vertex $r$ and tree $T$ mentioned in the BFS algorithm.

Note that during the algorithm, for each vertex $v \neq r$, when it is added into $T$, it must be added at Step 2-1 and the edge $v_{i} v$ is added into $T$ for some $i$, so $v_{i}$ is the parent of $v$ in $T$.

Lemma 3 If $v \neq r$ and $v$ is a child of $u$, then $d_{T}(r, v)=d_{T}(r, u)+1$.
Proof. Note that there exists a unique path $P$ in $T$ from $r$ to $v$. According to the algorithm, the edge $\{u, v\}$ is in $T$ and hence in $P$. So $P$ passes through $r, u, v$ in the order listed. So $d_{T}(r, v)=d_{T}(r, u)+1$.

Lemma 4 Let $x, y$ be integers with $2 \leq x<y$. If $v_{x}$ is a child of $v_{\alpha}$ and $v_{y}$ is a child of $v_{\beta}$, then $\alpha \leq \beta$.

Proof. If $\alpha>\beta$, then when we do step 2-1 for $i=\beta, v_{y}$ must be added into $T$, but $v_{x}$ is not added into $T$ until $i=\alpha>\beta$, contracting $x<y$.

Lemma $5 d_{T}\left(r, v_{1}\right) \leq d_{T}\left(r, v_{2}\right) \leq d_{T}\left(r, v_{3}\right) \leq \ldots \leq d_{T}\left(r, v_{n}\right)$.
Proof. We shall prove $d_{T}\left(r, v_{1}\right) \leq d_{T}\left(r, v_{2}\right) \leq d_{T}\left(r, v_{3}\right) \leq \ldots \leq d_{T}\left(r, v_{k}\right)$ by induction on $k$. When $k=1$, there is nothing to prove. When $k=2, v_{2}$ is a neighbor of $v_{1}=r$, so $d_{T}\left(r, v_{1}\right)=d_{T}(r, r)=0<1=d_{T}\left(r, v_{2}\right)$. So we may assume that $k \geq 3$ and $d_{T}\left(r, v_{1}\right) \leq d_{T}\left(r, v_{2}\right) \leq d_{T}\left(r, v_{3}\right) \leq \ldots \leq d_{T}\left(r, v_{k-1}\right)$.

Since $k \geq 3, k-1 \geq 2$. So $v_{k-1}$ is a child of $v_{\alpha}$ and $v_{k}$ is a child of $v_{\beta}$ for some $\alpha \leq \beta$ by Lemma 4. By the induction hypothesis, $d_{T}\left(r, v_{\alpha}\right) \leq d_{T}\left(r, v_{\beta}\right)$. By Lemma 3, $d_{T}\left(r, v_{k-1}\right)=d_{T}\left(r, v_{\alpha}\right)+1 \leq d_{T}\left(r, v_{\beta}\right)+1=d_{T}\left(r, v_{k}\right)$. This proves the lemma.

Lemma 6 For every nonnegative integer $j$, let $L_{j}=\{v \in V(G)=V(T)$ : $\left.d_{T}(v, r)=j\right\}$. Then no edge of $G$ has one end in $L_{\alpha}$ and one end in $L_{\beta}$ for some integers $\alpha, \beta$ with $\beta \geq \alpha+2$. In other words, if $x y$ is an edge of $G$, then $\left|d_{T}(r, x)-d_{T}(r, y)\right| \leq 1$. In particular, there exists no back edge.

Proof. Suppose there exists an edge $e$ of $G$ between a vertex $u$ in $L_{\alpha}$ and a vertex $v$ in $L_{\beta}$ for some $\beta \geq \alpha+2$. Since $v \in L_{\beta}, v$ is a child of a vertex $w$ in $L_{\beta-1}$ by Lemma 3. Since $\alpha<\beta-1$, we know $d_{T}(r, u)=\alpha<\beta-1=d_{T}(r, w)$. By Lemma 5, $u$ joins $T$ earlier than $w$. Since $u v \in E(G), v$ should be added into $T$ before all neighbors of $u$ are seen in the algorithm, a contradiction.

Theorem 7 For every vertex $v \in V(G), d_{G}(r, v)=d_{T}(r, v)$.
Proof. We shall prove this theorem by induction on $d_{G}(r, v)$.
When $d_{G}(r, v)=0, r=v$, so $d_{T}(r, v)=0=d_{G}(r, v)$. So we may assume that $d_{G}(r, v) \geq 1$ and $d_{G}\left(r, v^{\prime}\right)=d_{T}\left(r, v^{\prime}\right)$ for every vertex $v^{\prime}$ with $d_{G}\left(r, v^{\prime}\right)<d_{G}(r, v)$.

Note that since every path in $T$ is a path in $G, d_{G}(r, v) \leq d_{T}(r, v)$. So it suffices to prove that $d_{G}(r, v) \geq d_{T}(r, v)$.

Let $P$ be a shortest path in $G$ from $r$ to $v$. Note that $P$ contains at least 2 vertices since $r \neq v$. Let $u$ be the neighbor of $v$ in $P$. By Proposition 1, $d_{G}(r, v)=d_{G}(r, u)+1$. So $d_{G}(r, u)<d_{G}(r, v)$. By the induction hypothesis, $d_{G}(r, v)=d_{G}(r, u)+1=d_{T}(r, u)+1$. By Lemma $6,\left|d_{T}(r, u)-d_{T}(r, v)\right| \leq 1$, so $d_{T}(r, v) \leq d_{T}(r, u)+1=d_{G}(r, v)$. This proves the theorem.

### 2.3 Applications of BFS

Corollary 8 Let $G$ be a connected graph. Let $r$ be a vertex. Then in linear time, we can compute $d_{G}(r, v)$ and a shortest path in $G$ from $r$ to $v$ for all vertices $v$ of $G$ at once.

Proof. Find a BFS tree rooted at $r$. We can compute $d_{T}(r, v)$ for every $v \in V(G)$ during BFS. And $d_{G}(r, v)=d_{T}(r, v)$ by Theorem 7. Moreover, the unique path in $T$ from $r$ to $v$ is a shortest path in $G$ from $r$ to $v$.

The diameter of a graph $G$ is the largest distance between two vertices in $G$. That is, the diameter of $G$ equals $\sup _{u, v \in V(G)} d_{G}(u, v)$. Note that if $G$ is connected, then this supremum is actually a maximum; if $G$ is disconnected, then the diameter is infinite.

Corollary 9 The diameter of an input graph $G$ can be computed in time $O\left(|V(G)|^{2}+|V(G)||E(G)|\right)$.
Proof. For each vertex $r$ of $G$, we can compute $\sup _{v \in V(G)} d_{G}(r, v)$ in linear time by Corollary 8. So $\sup _{u, v \in V(G)} d_{G}(u, v)=\sup _{u \in V(G)} \sup _{v \in V(G)} d_{G}(u, v)$ can be computed in $O(|V(G)| \cdot(|V(G)|+|E(G)|))=O\left(|V(G)|^{2}+|V(G)||E(G)|\right)$.

A graph $G$ is bipartite if $V(G)$ can be partitioned into two (possibly empty) sets $A, B$ such that every edge of $G$ has one end in $A$ and one end in $B$. Such a partition $\{A, B\}$ is called a bipartition of $G$. Note that a graph is bipartite if and only if it is 2-colorable.

Corollary 10 Given the input graph $G$, we can either find a bipartition of $G$ or an odd cycle in $G$ in linear time.

Proof. Find a BFS tree $T$ rooted at an arbitrary vertex $r$. Let $A=\{v \in$ $V(G): d_{T}(r, v)$ is even $\}$ and let $B=\left\{v \in V(G): d_{T}(r, v)\right.$ is odd $\}$.

We can check whether there is an edge of $G$ with both ends in $A$ or with both ends in $B$ in linear time. If there exists no such an edge, then $\{A, B\}$ is a bipartition of $G$ and we output it.

So we may assume that there exists an edge $e$ with both ends in $C$ for some $C \in\{A, B\}$, say $e=u v$. For $x \in\{u, v\}$, let $P_{x}$ be the unique path in $T$ from $r$ to $x$. Note that $P_{u} \cap P_{v}$ is a path $R$ in $T$ from $r$ to a common ancestor $w$ of $u$ and $v$. And $P_{u} \cup P_{v}$ is the union of $R$ and the unique path $P$ in $T$ from $u$ to $v$, where $P$ and $R$ are edge-disjoint, so $|E(P)|=$ $\left|E\left(P_{u}\right)\right|+\left|E\left(P_{v}\right)\right|-2|E(R)|$. Since both $u, v$ are in $C,\left|E\left(P_{u}\right)\right|+\left|E\left(P_{v}\right)\right|$ is even, so $|E(P)|=\left|E\left(P_{u}\right)\right|+\left|E\left(P_{v}\right)\right|-2|E(R)|$ is even. Therefore, the cycle $P+u v$ is an odd cycle $O$, and we output $O$. Note that $P_{u}, P_{v}, w, R, P$ can be found in linear time. So $O$ can be found in linear time.

Corollary 11 A graph $G$ is bipartite if and only if $G$ does not contain an odd cycle (as a subgraph).

Proof. $(\Rightarrow)$ Since every odd cycle is not bipartite, no bipartite graph can contain an odd cycle.
$(\Leftarrow)$ If $G$ has no odd cycle, then the algorithm in Corollary 10 must find a bipartition of $G$, so $G$ is bipartite.

Corollary 12 Given a graph $G$, in linear time, we can find a bipartition of $G$ to correctly conclude that $G$ is bipartite if $G$ is bipartite, and output an odd cycle in $G$ to correctly conclude that $G$ is not bipartite if $G$ is non-bipartite. In particular, 2-COLORABILITY is in $\mathbf{P}$.

## 3 Depth-first-search

## Depth-First-Search (DFS)

Input: A connected graph $G$ and a vertex $r$ of $G$.
Output: A tree rooted at $r$.
Procedure:

Step 1: Set $T$ to be the rooted tree $(\{r\}, \emptyset)$ rooted at $r$. Set all edges of $G$ as "unmarked". Set $S$ to be the sequence $(r)$.

Step 2: Terminate the algorithm if $S$ has no entry. Say the last entry of $S$ is $v$. If there exists no unmarked edge of $G$ incident with $v$, then remove $v$ from $S$ and repeat Step 2. Otherwise, pick an unmarked edge $e$ of $G$ incident with $v$ and mark $e$. If $e$ is not a loop and the end $u$ of $e$ other than $v$ is not in $V(T)$, then add $e$ into $T$, add $u$ into $S$ as the last entry, and repeat Step 2; otherwise, just repeat Step 2.

## Remark:

- Clearly, $T$ is a tree during the entire process, and the final tree $T$ is a spanning tree of $G$ rooted at $r$. We call the final tree $T$ a depth-firstsearch (DFS) tree rooted at $r$.
- During the process, if a vertex $v$ is removed from $S$, then all edges incident with $v$ have been marked, so $v$ cannot be added into $S$ in the future.
- So Step 2 is executed at most $|V(G)|+|E(G)|+O(1)$ times. We can implement the algorithm so that finding an unmarked edge incident with $v$ can be done in $O(1)$ time. Hence the algorithm runs in time $O(|V(G)|+|E(G)|)$.
- Clearly, during the entire process, the entries of $S$ always form a path in $T$ from $r$ to the last entry of $S$ passing all entries in the order listed.

Like having no back edge is a key feature of every BFS tree, a key feature of every DFS tree is that it has no crossing edge.

Lemma 13 If $T$ is a DFS tree of a connected graph $G$, then there exists no crossing edge.

Proof. Suppose to the contrary that $e$ is a crossing edge. Let $x, y$ be the ends of $e$, where $x$ is added into $T$ earlier than $y$. When $x$ leaves $S, e$ is already marked. Let $z$ be the last entry of $S$ when we mark $e$. So $z \in\{x, y\}$ and $S$ contains both $x$ and $z$ at this moment. Hence $x$ is an ancestor of $z$ in $T$. If $z=x$, then $e$ must be added into the tree $T$, a contradiction. If $z=y$, then since $x$ is an ancestor of $z=y, e$ is a back edge, a contradiction.

